

TURBULENT CHEMICAL REACTIONS: APPLICATION
TO ATMOSPHERIC CHEMISTRY

Thesis by

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*This thesis is dedicated
to my mother*

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ABSTRACT

The general problem of predicting the rates of chemical reactions in turbulent fluid has been studied. Particular attention has been given to bimolecular (second-order) chemical reactions occurring isothermally. The problem of predicting the rates of turbulent chemical reactions has been formulated in both Eulerian and Lagrangian frameworks. The Eulerian formulation leads to a closure problem in the equations for the moments of concentrations of the reacting species. The Lagrangian formulation, while circumventing the usual closure problem, is based on a transition probability density function of particle displacements which can only be predicted in the simplest of cases.

The principal result of the work is the development of a new closure method for second-order chemical reactions in turbulence. The method is referred to as the *diffusion zone model*. The essential idea of the model is that one defines the diffusion zone as that region of the fluid in which the reactive species would coexist if they were inert. Then, chemical reaction is presumed to occur only within the diffusion zone; outside the diffusion zone the concentrations of the reactive species are identical to those if the species were inert. The diffusion zone model is tested extensively on available laboratory data and gives good agreement.

The diffusion zone closure method is adapted for predicting the rates of atmospheric chemical reactions occurring in plumes. The effect of inhomogeneous mixing on the reaction rates is explicitly accounted

for in the formulation. In assessing this effect, a numerical planetary boundary layer velocity field is employed as a source of the necessary turbulence data. The model is applied to predicting the rates of nitric oxide oxidation in a plume from the Morgantown power plant in Maryland. Predictions are in good qualitative agreement with available measurements on the plume.

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I. INTRODUCTION

Chemical reactions in turbulent fluids occur in situations of both industrial and natural significance. The main problem in the design of industrial and in the analysis of natural systems is the prediction of the rate of the chemical reactions. Apart from the laboratory ideals of the perfectly mixed vessel and pure plug flow, chemical reactions in turbulent flow will be influenced by the degree of turbulence and mixing in the system. To illustrate the variety of reactive flows in which turbulence is important, Table 1.1 lists some recent studies.

One of the major problems in the analysis of the dynamic behavior of air pollutants is predicting the rates of chemical reactions occurring among atmospheric species. The atmosphere is, of course, in a turbulent state, and since reactive pollutants are always emitted from highly localized sources, there is a period during which the pollutants are inhomogeneously mixed and reactions are taking place. During this period, depending on the speed of the reactions, the speed and extent of mixing may have a significant effect on the reaction rates. The objective of this work is twofold. First, we wish to study the general problem of predicting the rates of chemical reactions occurring in turbulent fluids. Second, we wish to study particularly the effect of atmospheric turbulence on the rates of chemical reactions involving air pollutants. We shall concentrate here on isothermal second-order reactions in turbulent flow.

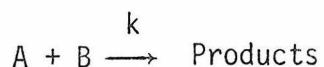
Table 1.1 Examples of Chemical Reactions in Turbulent Flow*

| Description of Physical Phenomena | References |
|--|--|
| 1. Tubular plug-flow reactor | Toor and Singh (1973) |
| 2. Continuous flow stirred-tank reactor | Evangelista et al. (1969) |
| 3. Nitric oxide formation in combustors | Flagan and Appleton (1974) |
| 4. Turbulent diffusion flame structure | Chung (1971, 1972) |
| 5. Chemical kinetics measurements | Glassman and Eberstein (1963) |
| 6. Turbulent jets and wakes | Borghini (1974), Shea (1976) |
| 7. Photochemical smog formation | Donaldson and Hilst (1972ab) |
| 8. Pollutant dispersion from smokestacks | Davis et al. (1974), Davis and Klauber (1975) Romberg (1974) |

*Hill (1976) has a more extensive but somewhat different list.

1.1 The Importance of Concentration Fluctuations

To introduce the problem with which we will be dealing, let us consider the isothermal, bimolecular chemical reaction



occurring between species A and B with rate constant k. The local rate of change of the concentrations of A and B in a turbulent system are

$$\frac{\partial A}{\partial t} = \frac{\partial B}{\partial t} = -kAB \quad (1.1)$$

where k is the kinetic rate constant, and A and B are the instantaneous values of the concentrations of species A and B, respectively. The instantaneous concentrations can be expressed as the sum of ensemble average* and fluctuating concentrations:

$$A = \langle A \rangle + A' \quad (1.2a)$$

$$B = \langle B \rangle + B' \quad (1.2b)$$

Substituting (1.2) into (1.1), and ensemble averaging, we find that the mean rate of reaction is given by

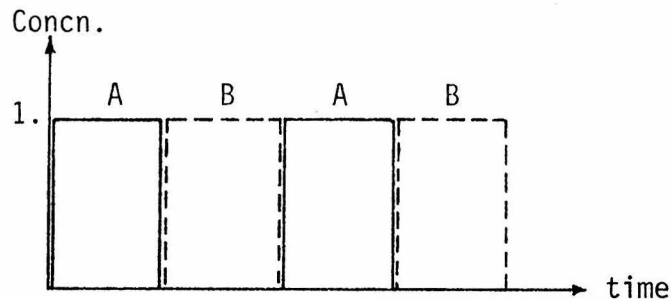
$$\frac{\partial \langle A \rangle}{\partial t} = \frac{\partial \langle B \rangle}{\partial t} = -k(\langle A \rangle \langle B \rangle + \langle A'B' \rangle) \quad (1.3)$$

*Ensemble average, or statistical average, of a certain random variable is the average value of the ensemble of an infinite number of statistically independent measurements of the variable. If not otherwise mentioned we shall use 'mean' or 'average' to refer to ensemble average in the following texts.

One is generally interested in predicting the ensemble average rate of reaction. The approximation often used in computing this rate is to replace the instantaneous, local expressions (1.1) which are derived from the chemical kinetics alone, by the corresponding expressions involving the ensemble mean concentrations, i.e.

$$\frac{\partial \langle A \rangle}{\partial t} = \frac{\partial \langle B \rangle}{\partial t} = -k \langle A \rangle \langle B \rangle \quad (1.4)$$

We see, however, from (1.3) that only when the correlation of the fluctuations $\langle A'B' \rangle$ is identically zero is (1.4) an exact relation. If $\langle A'B' \rangle > 0$, the mean reaction rate is enhanced over that predicted by (1.4); and if $\langle A'B' \rangle < 0$, the rate is suppressed. Thus, in order to use (1.4), it is necessary to demonstrate that $\langle A \rangle \langle B \rangle \gg |\langle A'B' \rangle|$ for the particular system of interest. Donaldson and Hilst (1972b) have suggested a hypothetical example in which the reaction rate predicted by (1.4) is totally incorrect. Consider a situation in which the flow of material by a given point is such that alternate patches of A and B pass the point. Let us suppose that half the time the flow is all A and half the time it is completely B as depicted below. If this pattern keeps repeating



and the patches of A and B all have a concentration of unity, the average concentrations are obviously $\langle A \rangle = \langle B \rangle = 1/2$. When the flow is all A, $A' = 1/2$ and $B' = -1/2$, but when the flow consists of B only, $A' = -1/2$ and $B' = 1/2$. Thus for this flow

$$\langle A \rangle \langle B \rangle = 1/4$$

$$\langle A' B' \rangle = -1/4$$

and by (1.3) the correct reaction rate is zero; but according to (1.4) the rate is $k/4$.

The second order correlation $\langle A' B' \rangle$ arises as a result of inhomogeneous mixing in the system. The next question that naturally follows is - Is it possible to predict $\langle A' B' \rangle$ for a system of interest and thereby utilize (1.3) instead of (1.4) for predictions? Diffusion of one substance through another is usually approached by considering a differential material balance in a fixed volume of space. The resulting differential equation is called the equation of mass continuity and can be solved as a boundary value problem describing the behavior of the concentration of the material of interest. Let us consider the equations of continuity for A and B in a fluid,

$$\frac{\partial A}{\partial t} + u_k \frac{\partial A}{\partial x_k} = \mathcal{D}_A \frac{\partial^2 A}{\partial x_k \partial x_k} - k_{AB} \quad (1.5)$$

$$\frac{\partial B}{\partial t} + u_k \frac{\partial B}{\partial x_k} = \mathcal{D}_B \frac{\partial^2 B}{\partial x_k \partial x_k} - k_{AB} \quad (1.6)$$

where u_k is the fluid velocity in coordinate direction k ($k = 1,2,3$), and \mathcal{D}_A and \mathcal{D}_B are the molecular diffusivities of A and B in the fluid. The summation convention has been employed in (1.5) and (1.6) in which a repeated subscript in a term indicates summation over the three components of that term, e.g.

$$u_k \frac{\partial A}{\partial x_k} \equiv \sum_{k=1}^3 u_k \frac{\partial A}{\partial x_k} \quad (1.7)$$

In a turbulent fluid, the velocities are customarily decomposed into mean and fluctuating components,

$$u_k = \langle u_k \rangle + u'_k \quad (1.8)$$

Introducing (1.2), and (1.8) into (1.5) and (1.6) and ensemble-averaging the resulting equations leads to

$$\frac{\partial \langle A \rangle}{\partial t} + \langle u_k \rangle \frac{\partial \langle A \rangle}{\partial x_k} + \frac{\partial}{\partial x_k} \langle u'_k A' \rangle = \mathcal{D}_A \frac{\partial^2 \langle A \rangle}{\partial x_k^2} - k(\langle A \rangle \langle B \rangle + \langle A' B' \rangle) \quad (1.9)$$

$$\frac{\partial \langle B \rangle}{\partial t} + \langle u_k \rangle \frac{\partial \langle B \rangle}{\partial x_k} + \frac{\partial}{\partial x_k} \langle u'_k B' \rangle = \mathcal{D}_B \frac{\partial^2 \langle B \rangle}{\partial x_k^2} - k(\langle A \rangle \langle B \rangle + \langle A' B' \rangle) \quad (1.10)$$

where in obtaining the third term of the left-hand side we have employed the total continuity equation for incompressible flow

$$\frac{\partial u'_k}{\partial x_k} = 0 \quad (1.11)$$

We see that (1.9) and (1.10) contain the seven new dependent variables, $\langle u'_k A' \rangle$, $\langle u'_k B' \rangle$ ($k = 1, 2, 3$), and $\langle A' B' \rangle$. The last terms on the right hand sides of (1.9) and (1.10) are, of course, the local rates of reaction discussed earlier.

Perhaps the first inclination in dealing with the seven new dependent variables is to derive the equations which govern their dynamics. Doing so, we obtain

$$\begin{aligned} \frac{\partial}{\partial t} \langle u'_k A' \rangle + \langle u'_i \rangle \frac{\partial}{\partial x_i} \langle u'_k A' \rangle + \langle u'_k u'_i \rangle \frac{\partial \langle A \rangle}{\partial x_i} + \frac{\partial}{\partial x_i} \langle u'_k u'_i A' \rangle \\ = \mathcal{D}_A \frac{\partial^2 \langle u'_k A' \rangle}{\partial x_i \partial x_i} - k [\langle u'_k A' \rangle \langle B \rangle + \langle u'_k B' \rangle \langle A \rangle + \langle u'_k A' B' \rangle] \end{aligned} \quad (1.12)$$

$$\begin{aligned} \frac{\partial}{\partial t} \langle u'_k B' \rangle + \langle u'_i \rangle \frac{\partial}{\partial x_i} \langle u'_k B' \rangle + \langle u'_k u'_i \rangle \frac{\partial \langle B \rangle}{\partial x_i} + \frac{\partial}{\partial x_i} \langle u'_k u'_i B' \rangle \\ = \mathcal{D}_B \frac{\partial^2 \langle u'_k B' \rangle}{\partial x_i \partial x_i} - k [\langle u'_k B' \rangle \langle A \rangle + \langle u'_k A' \rangle \langle B \rangle + \langle u'_k A' B' \rangle] \end{aligned} \quad (1.13)$$

$$\frac{\partial}{\partial t} \langle A' B' \rangle + \langle u'_k \rangle \frac{\partial}{\partial x_k} \langle A' B' \rangle = - \langle u'_k B' \rangle \frac{\partial \langle A \rangle}{\partial x_k} - \langle u'_k A' \rangle \frac{\partial \langle B \rangle}{\partial x_k} \quad (1) \quad (2)$$

$$- \frac{\partial}{\partial x_k} \langle u'_k A' B' \rangle + \mathcal{D} \frac{\partial^2 \langle A' B' \rangle}{\partial x_k \partial x_k} - 2\mathcal{D} \left\langle \left(\frac{\partial A'}{\partial x_k} \right) \left(\frac{\partial B'}{\partial x_k} \right) \right\rangle \quad (3) \quad (4) \quad (5)$$

$$- k [\langle A' B' \rangle (\langle A \rangle + \langle B \rangle) + \langle A' B'^2 \rangle + \langle A'^2 B' \rangle + \langle B'^2 \rangle \langle A \rangle + \langle A'^2 \rangle \langle B \rangle] \quad (6)$$

(1.14)

where in (1.14) we have assumed $\mathcal{D}_A = \mathcal{D}_B = \mathcal{D}$ for simplicity. The numbered terms in (1.14) have the following significance:

- (1) = convection of $\langle A'B' \rangle$ by mean velocity
- (2) = generation of $\langle A'B' \rangle$ by mean gradients in $\langle A \rangle$ and $\langle B \rangle$
- (3) = transport of $\langle A'B' \rangle$ by turbulent velocity fluctuations
- (4) = molecular diffusion of $\langle A'B' \rangle$
- (5) = dissipation of $\langle A'B' \rangle$ by molecular diffusion
- (6) = production and decay of $\langle A'B' \rangle$ due to chemical reaction

Although we can ascribe a significance to the terms in (1.12-14) these equations cannot be solved in closed form because of the appearance of still more unknowns. This phenomenon in turbulence is known as the closure problem. Some work has been done on proposing methods* to close the hierarchy of moment equations for turbulent chemistry, and we shall review this work shortly. First, however, qualitative conditions can be developed for when terms of the form $\langle A'B' \rangle$ can be expected to be significant (Donaldson and Hilst, 1972ab). In the absence of mean gradients, two competing processes influence the behavior of $\langle A'B' \rangle$:

*Closure methods are hypothesized based either on physical insight or strictly analytical approximations. The most common closure methods that have been proposed for the turbulent fluxes $\langle u'_k c' \rangle$, for example, is Prandtl's mixing length theory (Hinze, 1959) in which

$$\langle u'_k c' \rangle = -K \frac{\partial \langle C \rangle}{\partial x_k}$$

where K is a so-called eddy diffusivity.

molecular diffusion and chemical reaction. When two species react to form another species at a certain point, the concentrations of these two reactants instantly become smaller and as a result concentration fluctuations are produced. These fluctuations are dissipated by molecular diffusion (as expressed by term (5) in Eq. (1.14)); but if the rate of dissipation is significantly slower than that of the chemical reaction, then the reactants tend to become segregated and the overall rate of the reaction is perturbed. The ratio of the magnitudes of the reaction and dissipation processes is expressed by the quotients of terms (5) and (6) in Eq. (1.14). Expressing the dissipation in terms of a dissipation length scale λ (Corrsin, 1958), i.e.

$$2\mathcal{D} \left\langle \frac{\partial A'}{\partial x_k} \frac{\partial B'}{\partial x_k} \right\rangle \sim \frac{2\mathcal{D} \langle A'B' \rangle}{\lambda^2} \quad , \quad (1.15)$$

and assuming that the reaction rate term (6) in (1.14) is of the same order of magnitude as

$$k \langle A+B \rangle \langle A'B' \rangle \quad ,$$

we can write

$$N = \frac{\tau_D}{\tau_R} = \frac{k \langle A+B \rangle \lambda^2}{2\mathcal{D}} \quad (1.16)$$

where τ_D and τ_R are the time scales of the dissipation and reaction processes, respectively. The dimensionless quantity N is called the Damkohler number. When $N \ll 1$, the time scale of the reaction is much larger than that of the dissipation, and concentration fluctuations are eradicated before they had time to affect the chemistry. In this

case the mean reaction rate can be adequately predicted by $-k\langle A\rangle\langle B\rangle$. On the other hand, when $N \gg 1$, the characteristic time for chemical reaction is short compared to that for molecular dissipation. In that case, $\langle A'B'\rangle$ will tend to $-\langle A\rangle\langle B\rangle$, and the two species will be poorly mixed. The rate of reaction between A and B will be governed not by the kinetics but by the rate of molecular diffusion.

A characteristic value of λ in the atmosphere is 10 cm. Using a typical value of $0.17 \text{ cm}^2/\text{sec}$ for \mathcal{D} and 0.2 ppm for $(\langle A\rangle + \langle B\rangle)$, we calculated values of N for several reactions occurring in photochemical smog and listed them in Table 1.2. Although it is not strictly correct to consider reactions individually which occur in a complicated sequence, we do obtain, nevertheless, an indication that many atmospheric reactions might be "diffusion-limited." The extent of the importance of this effect will be examined later in this work. We shall first review previous work relating to the study of this effect.

Table 1.2 Estimated Ratios of Characteristic Diffusion Time to Characteristic Reaction Time for Important Atmospheric Reactions

| Reaction | k^* (ppm-min) ⁻¹ | $N = \frac{\tau_D}{\tau_R}$ |
|-----------------------------|----------------------------------|-----------------------------|
| $O_3 + NO = NO_2 + O_2$ | 2.9×10^1 | 28.4 |
| $NO_2 + O_3 = NO_3 + O_2$ | 4.6×10^{-2} | 0.045 |
| $NO_3 + NO = 2NO_2$ | 1.5×10^4 | 14705.8 |
| $NO + HO_2^* = NO_2 + OH^*$ | 7.0×10^2 | 686.3 |

* Seinfeld (1975)

1.2 Previous Work

Perhaps the first work on the statistical theory of turbulent chemical reactions was that of Corrsin (1958), who considered decay of a single species undergoing an isothermal irreversible reaction in isochoric, isotropic turbulence. He concluded that first-order reactions possess the same general character as those of pure mixing without reaction, but for second-order reactions, the nonlinearity of the kinetics introduces additional unknowns and makes the problem decidedly different from the pure mixing one. Considerable interest has since been focused on the spectral analysis of reactive mixtures in turbulent fluids (Corrsin, 1961, 1962, 1964; O'Brien and Francis, 1962; Pao, 1964). These calculated spectra, however, have not yet received experimental support. (Hill, 1976). Later reported work concentrates more on the prediction of statistical moments of concentrations of reactive species as shall be reviewed below.

All approaches to turbulent phenomena may be grouped into two broad categories: Eulerian and Lagrangian. The former begins with the mass continuity equation (1.5-6) and at some point introduces a closure hypothesis to achieve a solvable set of equations. The Lagrangian approach arrives at a closed set of model equations by postulating functional forms for certain probability density functions of the displacements of fluid particles*. In the remainder of this section we review

*Monin and Yaglom (1971) define a fluid particle as a lump of fluid that is large compared to the mean free path of molecules of the fluid but small enough so that within the framework of continuum mechanics it may be identified as a point moving in the fluid.

previous attempts of both the Eulerian and the Lagrangian types to analyze the problem of isothermal reactions* in turbulent flows.

Eulerian Approaches (Closure Methods)

There have been several studies of the simple Eulerian equation

$$\frac{dA(t)}{dt} = -kA^2(t) \quad (1.17)$$

with stochastic initial conditions $A(0)$. O'Brien (1966) compared expressions for $\langle A(t) \rangle$ and $\langle A'^2(t) \rangle$ obtained by averaging the exact solution of Eq. (1.17) with the predictions of these quantities using the third moment discard, quasinormal, and direct interaction approximations. None of the approximations behaved satisfactorily when the relative amplitude of the initial fluctuations, viz $\langle A'^2(0) \rangle / \langle A(0) \rangle^2$, was large. Later O'Brien (1968) recognized that since A is a nonnegative random variable, the moments of A must satisfy Liapunov's inequality (Uspensky, 1937):

$$\langle A^b \rangle^{a-c} \leq \langle A^c \rangle^{a-b} \langle A^a \rangle^{b-c}, \quad a > b > c \geq 0 \quad (1.18)$$

where a , b and c are constants. As an illustration, the third moment of A' must satisfy

$$\langle A'^3 \rangle \geq \langle A \rangle \langle A'^2 \rangle \left(\frac{\langle A'^2 \rangle}{\langle A \rangle^2} - 1 \right) \quad (1.19)$$

*For treatments of nonisothermal reactions in turbulent flows, see Libby (1972), Dopazo and O'Brien (1973), Flagan and Appleton (1974).

It is obvious that if we adopt, say, the third moment discard approximation, $\langle A'^2 \rangle / \langle A \rangle^2$ must be ≤ 1 at all times. To avoid unrealistic restrictions on the initial fluctuations, O'Brien (1968) proposed a so-called inequality preserving closure approximation (IPCA) for $\langle A'^3 \rangle$ such that Liapunov's inequality is always satisfied. O'Brien again compared the predictions of his IPCA with the exact solutions of (1.17) and found satisfactory agreement. O'Brien and Eng (1970) generalized the closure for reaction of order one to three, and O'Brien and Lin (1972) used a different IPCA for two-species reaction with spatial variables. Finally, in a recent study Lee (1973) presented a generalized direct interaction approximation to (1.17). All of these approximations appeared to behave satisfactorily when tested against (1.17). Since each of these closure schemes was developed and tested on a simple system, namely (1.17), there is no assurance that they will hold when applied to the full continuity equations (1.5-6).

One of the few closure schemes that have been applied to the full continuity equations is that of McCarthy (1970). By discarding fifth order cumulants, he developed a hierarchy of 78 differential equations for single-point concentration moments and microscales. Later Lin and O'Brien (1972) presented a closure theory which incorporates Lin's (1971) third order IPCA for the reaction terms and Lee's (1966) modification of the quasinormal approximation for the convective terms. Computations of the decay of moments and spectra of A were carried out for various conditions. The decay of $\langle A \rangle$ and $\langle A'^2 \rangle$ was found to

depend primarily on the Damkohler number defined as in (1.16). More recently, Hilst et al. (1973) combined a third order IPCA (different from that of Lin, 1971) for the reaction term $A'B'$ with an "invariant model" (Donaldson and Rosenbaum, 1969) for the convection term. Hilst et al. then applied the model to the reaction of O_3 and NO emanating from four cross-wind freeway line sources and solved the resulting 12 coupled differential equations numerically.

Due to their mathematical complexity and the lack of experimental support, none of the closure schemes mentioned above is particularly useful in practical applications to chemical engineering and atmospheric pollution. Closure schemes more suited to these needs have been developed by Toor, Brodkey, and their students.

Toor (1962) noticed that by subtracting (1.6) from (1.5) the equation governing $C = A - B$ is identical to the transport equation of an inert species:

$$\frac{\partial C}{\partial t} + u_k \frac{\partial C}{\partial x_k} = \mathcal{D} \frac{\partial^2 C}{\partial x_k \partial x_k} \quad (1.20)$$

provided that $\mathcal{D}_A = \mathcal{D}_B = \mathcal{D}$. Since for very rapid reactions A and B do not coexist, C is equivalent to A when $C \geq 0$ and it is equivalent to B when $C \leq 0$. The probability density functions for A and B are thus the same as for C in the proper range, and

$$\langle A \rangle = \langle B \rangle = \int_0^{\infty} C p(C) dC \quad (1.21)$$

Following the work of Hawthorne et al. (1949), Toor assumed $p(C)$ to be Gaussian, and an explicit result was obtained in terms of stoichiometric feed ratio β and the mean and variance of C . For stoichiometric feed ($\beta=1$) the result is

$$\langle A(x,t) \rangle = \frac{\langle C'^2(x,t) \rangle^{1/2}}{\sqrt{2\pi}} \quad (1.22)$$

at any point x in the fluid. In a one-dimensional tubular reactor, a more useful expression is

$$F_{\infty}(t) \equiv \frac{\langle A(t) \rangle}{\langle A(0) \rangle} = \frac{\langle B(t) \rangle}{\langle B(0) \rangle} = \frac{\langle C'^2(t) \rangle^{1/2}}{\langle C'^2(0) \rangle^{1/2}} \equiv I_S^{1/2} \quad (1.23)$$

The quantity I_S can be measured by an identical but nonreacting system. Toor (1969) further showed that for the case of initially unmixed reactants I_S is equivalent to the intensity of segregation:

$$I_S = \frac{\langle A_I B_I(t) \rangle}{\langle A_I B_I(0) \rangle} \quad (1.24)$$

where A_I, B_I are concentrations that species A and B would have if they were inert. Toor's theory has since received extensive experimental support (Vassilatos and Toor, 1965; Keeler et al. 1965; Mao, 1969; Torrest and Ranz, 1970; Miyairi et al. 1971; Miyawaki et al. 1974). Although his theory was based on the assumption of $\mathcal{D}_A = \mathcal{D}_B$, a generalization to nonequal diffusivities has been reported recently (O'Brien, 1971). We shall discuss Toor's method in more detail in Appendix A.

For reaction of intermediate speed, Toor's model does not apply because reactants do coexist. To handle these cases Toor (1969) proposed for reactions of any speed with stoichiometric feed ($\beta=1$) that

$$\langle A'B' \rangle = \langle A_I' B_I' \rangle \quad (1.25)$$

This postulate was based on the observation that (1.25) is valid in the limit of both very slow and very rapid stoichiometric reactions. (The latter follows from (1.23-24) and the fact that $\langle A'B' \rangle \rightarrow -\langle A \rangle \langle B \rangle$.) Yieh (1970) provided additional theoretical support for (1.25) by assuming statistical independence of the product and reactant concentrations (although a negative correlation would be expected). Mao and Toor (1971) and Ajmera et al. (1975) tested (1.25) in both liquid and gas phase reactions in an idealized one-dimensional reactor* and found it to be reasonably accurate. Based on these findings Toor and Singh (1973) later proposed scale-up criteria for tubular reactors. In addition to this model, several others have suggested schemes for describing reactions of intermediate speeds, (e.g., Kattan and Adler, 1967; Mao and Toor, 1970; Rao and Dunn, 1970; Evangelista et al. 1969; Kattan and Adler, 1972). Some of their results are discussed in Section II.3.

* More facts about this reactor and the experiments and models associated with it are given in Section II.3. and Appendix A.

Lagrangian Approaches

Lagrangian treatments of turbulence phenomena are relatively few in number because the particle statistics required by this approach are generally extremely difficult to determine.

In a series of papers, Chung (1969, 1970, 1972) used probability density functions (pdf) of a model velocity field which obeys a Langevin equation to study reacting turbulent shear flow. The resulting Fokker-Planck equations were solved by the method of moments, assuming Maxwellian distribution. Chung's formulation has been questioned by Bush and Fendell (1974) and O'Brien (1974). Different Lagrangian formulations were proposed by Hill (1970) and Riley (1973) for one-point concentration moments, but solutions were not given. Recently, Lamb (1972b) derived an expression which relates reactive decay of a species to the probability density function that describes the paths of inert fluid particles. Lamb's approach will be discussed further in Chapter IV.

Brief Summary

Despite the number of previous studies, there does not exist a generally valid means of predicting the rates of chemical reactions in turbulent fluids. Toor's (1962) theory for very rapid reactions is perhaps the most useful, but it has not been tested in systems other than chemical reactors. Schemes that appear to be more general, such as the IPCA's, are very difficult to apply due to their mathematical complexity.

I.3 Scope of this Thesis

As noted earlier, there are two approaches available for turbulent phenomena. The Eulerian approach starts from the familiar species continuity equation, but it is inevitably frustrated by the closure problem. Although a number of approximate methods can be invoked for closing the hierarchy of equations, an exact solution cannot be obtained. However, an important issue is the development of simple, yet accurate, closure approximations for terms of the form $\langle A'B' \rangle$ which arise when chemical reactions take place. The development of such approximations would enable the use of the Eulerian continuity equations in calculating turbulent, reacting flows. More importantly, perhaps, such a development would enable one to assess when predictions based on mean-value kinetics alone will not suffice. Chapter II is devoted to the development of a new closure approximation for turbulent, reacting flows. The closure model is then tested against laboratory data. The application of this model to atmospheric chemistry is the subject of Chapter III. A new formulation will be presented to predict the reactive decay of the species emitted from smokestacks. To assess the effect of turbulence on atmospheric reaction rates, we employ the numerical planetary boundary layer flowfield of Deardorff (1974ab) as a source of the necessary turbulence data. The resulting prediction is then compared with some atmospheric observations.

The Lagrangian approach to fluid mechanical problems describes phenomena with respect to specific fluid particles. In the case of the

turbulent diffusion of inert species, the Lagrangian formulation circumvents the closure problem which arises in the Eulerian description (Lamb, 1972a). However, the key variable in the Lagrangian formulation, the transition probability density function (pdf), is very difficult to predict theoretically or measure experimentally. When species may react chemically, the Lagrangian formulation requires that individual particle trajectories be known in order to account for "collisions" of reactive species. A numerical technique of measuring the transition pdf is demonstrated in Chapter III, while Chapter IV is devoted to a Lagrangian analysis of turbulent chemical reactions. Although the analysis is quite complicated, results of interest are obtained in some special cases.

Conclusions and recommendations for future work are given in Chapter V.

II. A DIFFUSION ZONE MODEL FOR TURBULENT CHEMICAL REACTION

The object of this chapter is to develop a closure scheme for second-order turbulent reaction systems. A so-called diffusion-zone model will be presented based largely on physical insight rather than on mathematical assumptions. Applications of this scheme to simulate reactor data are carried out, and agreement with data is shown to be very good. A comparison with other related work is also given at the end of the chapter.

II.1 Development of the Diffusion Zone Model

There are three phenomena influencing the concentration statistics of reactive species in a turbulent fluid: turbulent mixing, molecular diffusion, and chemical kinetics. The relative roles of these three phenomena are illustrated in Figure 2.1, in which we depict the turbulent mixing of two fluid elements containing reactive species A and B. Turbulent mixing serves to stretch the patches of A and B into thin sheets and filaments as shown in Figure 2.1 (a). However, the turbulence is not capable of mixing the patches at the very smallest scales. The region of small-scale inhomogeneity can be estimated to have a length of the order of $(\nu \mathcal{D}^2 / \epsilon)^{1/4}$, where ν is the kinematic viscosity of the fluid, \mathcal{D} the molecular diffusivity of the species (for the sake of qualitative argument, assume $\mathcal{D}_A = \mathcal{D}_B = \mathcal{D}$), and ϵ the rate of energy dissipation of

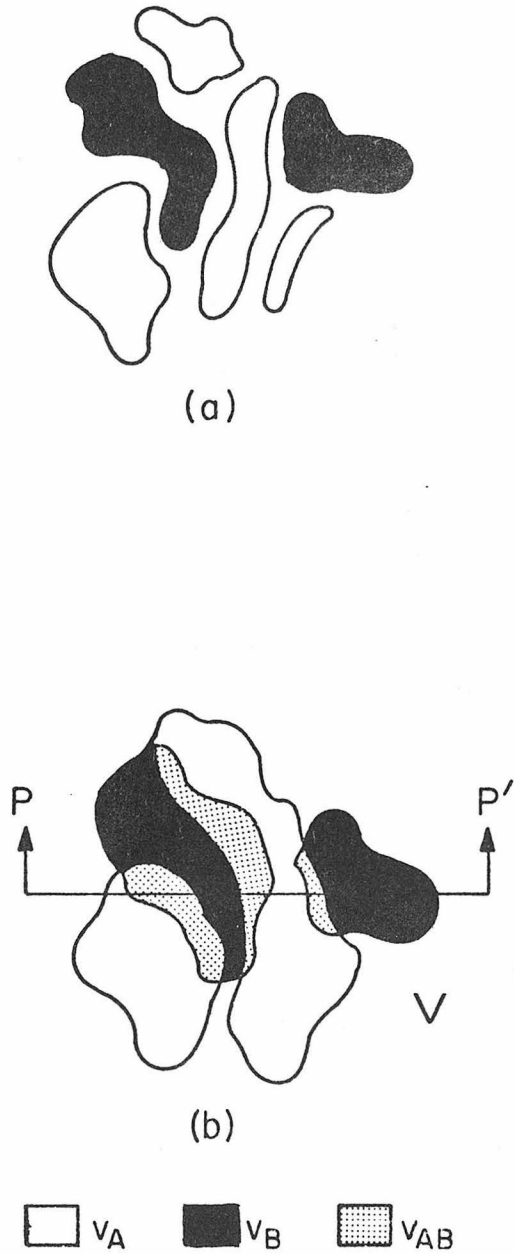


Figure 2.1 Schematics of turbulent mixing (a) without molecular diffusion and (b) with molecular diffusion.

the turbulence (Tennekes and Lumley, 1972, p. 240). This length scale can be referred to as the *contaminant microscale*, first introduced by Batchelor. For chemical reaction to occur, it is necessary that the patches of A and B mix. Such mixing, on scales smaller than the contaminant microscale, can only be achieved by molecular diffusion. Let us call the region in which this diffusion takes place the *diffusion zone* (Figure 2.1 (b)). If A and B are inert, the diffusion zone is simply the region v_{AB} in which A and B coexist. When A and B react with each other, the region in which they coexist is often called the reaction zone. It is normally smaller than v_{AB} due to consumption of the species by reaction.

For simplicity we develop the model considering only a single second-order chemical reaction of the form $A + B \rightarrow \text{products}$ and only a single spatial dimension. Subsequently, the model will be generalized to three space dimensions (Section II.4)

Consider a volume V of a turbulent fluid into which species A and B are released. At any given point in the fluid, the rates of change of the concentrations of A and B due to chemical reaction are given by

$$\frac{dA}{dt} = \frac{dB}{dt} = -k_{AB}. \quad (2.1)$$

We now define the volume averaged quantities

$$\bar{A} = \frac{1}{V} \int_V A dv, \quad (2.2a)$$

$$\bar{B} = \frac{1}{V} \int_V B dv, \quad (2.2b)$$

$$\overline{AB} = \frac{1}{V} \int_V AB dv \quad (2.2c)$$

where A and B denote the concentrations of species A and B, respectively.

Taking the volume average of (2.1) we have

$$\frac{d\bar{A}}{dt} = \frac{d\bar{B}}{dt} = -k\overline{AB}. \quad (2.3)$$

If the volume V is sufficiently large and if the joint mean values of A and B are homogeneous (independent of location) and meet certain other weak requirements*, then according to ergodic theory the volume averages defined by (2.2) are equivalent to ensemble averages and hence the former are invariant from one realization of the turbulent-chemical process to another. In the analyses that follow, we shall assume that all averages over the volume V possess this invariance property.

Suppose that in the places of the reactive species A and B inert materials A_I and B_I are released instead. In this case we define v_{AB} to be the total volume of fluid within V in which A_I and B_I are found together. In practice, one might consider the edge of a particular portion of v_{AB} to be the point where a probe indicates that the concentration of A_I or B_I (or both) is vanishingly small. In any event, v_{AB} is defined only for the sake of convenience and it does not enter explicitly into the final mathematical form of the closure model equations. In a similar manner we define v_A and v_B to be the volumes within V

* Such as the spatial correlation $\langle A(x+\Delta x)B(x) \rangle$ goes to zero as Δx goes to infinity. See Monin and Yaglom (1971) for more details.

in which the species A_I is found alone and B_I is found alone, respectively. (See Figure 2.1b.) It is important to keep in mind that v_A , v_B , and v_{AB} are defined in terms of the inert rather than the reactive scalar species.

Since $A_I(B_I)$ exists only in v_{AB} and v_A (v_B), the volume averages analogous to (2.2) can be expressed in the following forms:

$$\bar{A}_I = \frac{1}{V} (v_A \hat{A}_I + v_{AB} \hat{a}_I) \quad (2.4a)$$

$$\bar{B}_I = \frac{1}{V} (v_B \hat{B}_I + v_{AB} \hat{b}_I) \quad (2.4b)$$

$$\overline{A_I B_I} = \frac{1}{V} \mu_I \hat{a}_I \hat{b}_I v_{AB} \quad (2.4c)$$

where

$$\hat{A}_I = \frac{1}{v_A} \int_{v_A} A_I dv \quad , \quad \hat{B}_I = \frac{1}{v_B} \int_{v_B} B_I dv \quad , \quad (2.5)$$

$$\hat{a}_I = \frac{1}{v_{AB}} \int_{v_{AB}} A_I dv \quad , \quad \hat{b}_I = \frac{1}{v_{AB}} \int_{v_{AB}} B_I dv \quad , \quad (2.6)$$

and

$$\mu_I = \frac{1}{\hat{a}_I \hat{b}_I} \left[\frac{1}{v_{AB}} \int_{v_{AB}} A_I B_I dv \right] \quad . \quad (2.7)$$

Under the assumptions invoked earlier that γ is large and that the statistics of the scalar fields satisfy the conditions of the ergodic theorem, the volume averages \bar{A}_I , \bar{B}_I , and $\overline{A_I B_I}$ are invariant from one realization of the turbulence to another. Hence, according to

Eqs. (2.4), in general the quantities defined in (2.5)-(2.7) also possess this invariance property.

Returning to the reactive species, we postulate that the mean values given by Equation (2.2) can be expressed in the following forms:

$$\bar{A} = \frac{1}{V} (v_A \hat{A}_I + v_{AB} \hat{a}) \quad , \quad (2.8a)$$

$$\bar{B} = \frac{1}{V} (v_B \hat{B}_I + v_{AB} \hat{b}) \quad , \quad (2.8b)$$

$$\bar{AB} = \frac{1}{V} \mu \hat{a} \hat{b} v_{AB} \quad , \quad (2.8c)$$

where

$$\hat{a} = \frac{1}{v_{AB}} \int A dv \quad , \quad \hat{b} = \frac{1}{v_{AB}} \int B dv \quad , \quad (2.9)$$

$$\mu = \frac{1}{\hat{a} \hat{b}} \left[\frac{1}{v_{AB}} \int AB dv \right] \quad , \quad (2.10)$$

and where v_A , v_B , v_{AB} , \hat{A}_I , and \hat{B}_I have the same meanings as before. The integrals in Eqs. (2.9) and (2.10) denote integrations over the region v_{AB} where, all things being constant, A and B would be found together if they were inert. Due to the depletion of both species A and B by the reaction, the total volume of fluid (within V) in which the reactive species are found together is always less than v_{AB} .

Substituting (2.8a) and (2.8b) into (2.8c) and combining the result with (2.3), we obtain the rate equation

$$\frac{d\bar{A}}{dt} = \frac{d\bar{B}}{dt} = - k_{\mu} \left[\bar{A} - \frac{v_A \hat{A}_I}{V} \right] \left[\bar{B} - \frac{v_B \hat{B}_I}{V} \right] \frac{v_{AB}}{V} \quad . \quad (2.11)$$

On making use of Eqs. (2.4a,b) we can reduce (2.11) to the final form

$$\frac{d\bar{A}}{dt} = \frac{d\bar{B}}{dt} = \frac{-k\lambda\Gamma_{AB}}{\zeta_A\zeta_B} \left[\bar{A} - (1 - \zeta_A)\bar{A}_I \right] \left[\bar{B} - (1 - \zeta_B)\bar{B}_I \right] \quad (2.12)$$

where

$$\Gamma_{AB} = \frac{\overline{A_I B_I}}{\overline{A_I} \overline{B_I}} \quad , \quad (2.13)$$

$$\zeta_A = \frac{\hat{a}_I v_{AB}}{\overline{A_I} V} \quad , \quad \zeta_B = \frac{\hat{b}_I v_{AB}}{\overline{B_I} V} \quad (2.14)$$

$$\lambda = \mu/\mu_I \quad . \quad (2.15)$$

The parameters ζ_A and ζ_B , which we call the *mixing parameters*, are the ratios of the quantities of A_I and B_I in the domain v_{AB} to the total quantities of these species present in V . Thus, ζ_A and ζ_B are strictly determined by the properties of the turbulence and molecular diffusion because they are defined in terms of the inert materials A_I and B_I . The parameter λ , which we call the *reaction parameter*, is the ratio of the correlation coefficient of the concentrations of the reactive species in the domain v_{AB} to that of the inert materials A_I and B_I in the same region. In contrast to ζ_A and ζ_B , the chemistry plays a role in determining the functional form of λ .

Although Γ_{AB} , $\overline{A_I}$, and $\overline{B_I}$ are quantities that are readily measurable, the parameters λ , ζ_A , and ζ_B are not and thus Eq. (2.12) is unsolvable until appropriate expressions for these parameters have been found. It is apparent, therefore, that we have transformed the closure

problem associated with Eq. (2.3) into one of finding expressions for the mixing and reaction parameters ζ_A , ζ_B , and λ . The latter task would seem to be more tractable because as it is formulated, the effects of the chemistry and those of the turbulence are to a large extent segregated. This permits a much easier use of physical insight in developing expressions for the unknown parameters and it facilitates the use of the resulting rate equations in applied problems. In the next section we discuss in detail the determination of ζ_A , ζ_B , and λ .

First, we note some special cases. If the reactants are fully premixed, $\zeta_A = \zeta_B = 1$. Therefore (2.12) simplifies to

$$\frac{d\bar{A}}{dt} = -k\lambda\Gamma_{AB}\bar{A}\bar{B} \quad (2.16)$$

Eq. (2.16) is also the asymptotic form of (2.12) in the case $t \rightarrow \infty$.

In practice, premixed reactants occur primarily in two cases. Firstly, when one species interacts with itself, e.g. $A + A \rightarrow$ products, (2.16) becomes

$$\frac{d\bar{A}}{dt} = -k\lambda\Gamma_A\bar{A}^2 \quad (2.17)$$

where $\Gamma_A = \overline{A^2}/\bar{A}^2$. Secondly, when A and B are mixed under a condition unfavorable to reaction such as lack of radiation energy (in photochemical reactions), low temperature in an endothermal reaction, etc. the governing equation is again of the form of (2.16). In both cases, the macroscopic mixing is considered complete. If the microscopic mixing is also complete, then the reaction is much slower than the mixing process, and

r_{AB} , ζ_A , and ζ_B approach unity. In addition, we shall see subsequently that λ also approaches unity. Thus, in this case

$$\frac{d\bar{A}}{dt} = -k\bar{A}\bar{B} \quad (2.18)$$

which is the familiar expression generally used for turbulent second-order reaction.

II.2 Estimation of Parameters

Since both the mixing and reaction parameters are defined in terms of characteristics of the diffusion zone v_{AB} , an understanding of the processes that control the generation and fate of this region is essential to the formulation of meaningful expressions for ζ_A , ζ_B , and λ . Of course, the precise details of the evolution of the diffusion zone are not known, but sufficient knowledge of a heuristic nature is available (due mainly to the work of Batchelor (1959) and others) that a model of v_{AB} adequate for the purposes of estimating ζ_A , ζ_B , and λ can be developed. The heuristic knowledge of which we speak is best described using an example.

Suppose that marked fluid of molecular diffusivity \mathcal{D} is injected into a turbulent fluid of kinematic viscosity ν in which turbulent energy is being dissipated at a rate ϵ . Velocity fluctuations of spatial scales that are large compared to $(\nu^3/\epsilon)^{1/4}$ serve to distort the initial cloud of marked fluid by pulling and squeezing it into convoluted sheets and ribbons of ever decreasing thickness. At scales small

compared to $(\nu^3/\epsilon)^{1/4}$, the velocity field appears as a pure straining motion and consequently, sheets and ribbons of marked fluid produced by larger scale velocity perturbations are squeezed still further. If the molecular diffusivity of the marked fluid were zero, the stretching and squeezing of the sheets of marked fluid would continue indefinitely and mixing would occur only in the limit of infinitely large times. However, if \mathcal{D} is finite, then according to Batchelor (1959), once the sheet widths approach a thickness comparable to $(\nu\mathcal{D}^2/\epsilon)^{1/4}$ the concentration gradients normal to the sheet are sufficiently large that molecular fluxes of marked fluid just balance the compression of the sheet by the straining motion. At points in the fluid where this occurs, thorough mixing of the marked and convecting fluids results, and hence, by our definition, portions of the diffusion zone v_{AB} are created. At scales larger compared to $(\nu\mathcal{D}^2/\epsilon)^{1/4}$ concentration gradients are too weak to cause significant mixing and thus the generation of v_{AB} is confined primarily to those portions of the marked fluid where the sheet thickness is comparable to $(\nu\mathcal{D}^2/\epsilon)^{1/4}$. Based on these observations, we conclude that the evolution of v_{AB} can be modeled in one dimension using concentration pulses of widths and separations of the order of $(\nu\mathcal{D}^2/\epsilon)^{1/4}$. We shall subsequently derive expressions for ζ_A , ζ_B , and λ based on such a model.

II.2.1 The Mixing Parameters

By definition ζ_A is the ratio of the quantity of inert A brought into v_{AB} by turbulent mixing and molecular diffusion to the total amount of A present. Hence, it is a measure of the mixing efficiency. In principle, ζ_A and ζ_B can be determined by measurements of the mixing of inert species A_I and B_I in the turbulent fluid. When this type of measurement is not available, an approximation may be developed. If Γ_{AB} is known, Γ_A and Γ_B in general are also known (see Appendix A). Hence, Γ_A and Γ_B may be used as additional available information.

Let us define

$$\Gamma_A = \frac{\overline{A_I^2}}{\overline{A_I^2}} = \frac{1}{V} (\mu_A v_A \hat{A}_I^2 + \mu_a v_{AB} \hat{a}_I^2) / \overline{A_I^2} \quad (2.19)$$

where

$$\mu_A = \frac{1}{\hat{A}_I^2} \frac{1}{v_A} \int_{v_A} A_I^2 dv \quad (2.20)$$

$$\mu_a = \frac{1}{\hat{a}_I^2} \frac{1}{v_{AB}} \int_{v_{AB}} a_I^2 dv \quad (2.21)$$

with similar definitions for Γ_B , μ_B , and μ_B . Substituting (2.4ab) and (2.14) into (2.19), we obtain

$$\Gamma_A = \zeta_A \frac{V}{v_{AB}} \left[\frac{\mu_A \hat{A}_I}{\hat{a}_I} (1 - \zeta_A) + \mu_a \zeta_A \right] \quad (2.22)$$

Since $\Gamma_{AB} = \mu_I \zeta_A \zeta_B V / v_{AB}$, we can eliminate V / v_{AB} from (2.22) to obtain

$$\frac{\Gamma_{AB}}{\Gamma_A} = \zeta_B \frac{\mu_I/\mu_a}{\frac{\mu_A \hat{A}_I}{\mu_a \hat{a}_I} (1-\zeta_A) + \zeta_A} \quad (2.23)$$

with a similar expression for Γ_{AB}/Γ_B .

Although (2.23) gives ζ_A in terms of the measurable quantity Γ_{AB}/Γ_A , the expression involves μ_I , μ_A , and μ_a , which are unknown. (The closure problem has, of course, not been resolved; it is implicit in the definitions of the second-order correlations μ_I , μ_a , etc.) It is at this point that an approximation must be made to eliminate the closure problem. This is achieved here by considering some special cases.

Let us say that in the first case each fluid element has a flat concentration distribution. Imagine that we measure the concentrations of A and B along the plane PP' in Figure 2.1(b) and that we observe a distribution as shown in case 1 of Table 2.1 (p.36) In reality, molecular diffusion will obliterate sharp discontinuities in concentrations. However, because ζ_A and ζ_B represent the ratios of the amounts of inert A and B in v_{AB} to those in the entire system, the details of the concentration profiles within v_{AB} are not of prime importance. We therefore assert that this approximation of concentration profiles is good, particularly for systems where $v \gg \mathcal{D}$. In this case, $\mu_a = \mu_A = \mu_b = \mu_B = \mu_I = 1$, and $\hat{a}_I = \hat{A}_I$ and $\hat{b}_I = \hat{B}_I$. Upon using these approximations in (2.23) we obtain

$$\zeta_A = \frac{\Gamma_{AB}}{\Gamma_B} \qquad \zeta_B = \frac{\Gamma_{AB}}{\Gamma_A} \qquad (2.24)$$

When the reactants are non-premixed, ζ_A and ζ_B start with a value of zero. Γ_{AB} in this case is also zero. Hence (2.24) gives the right prediction. In the limit of $t \rightarrow \infty$, ζ_A and ζ_B go to unity if the reaction occurs in a confined environment. Γ_{AB} , Γ_A and Γ_B in this limit also go to one, hence (2.24) also gives the correct prediction. However, we expect that (2.24) slightly underestimates the true ζ 's. This is because when all the reactants are brought into the diffusion zones ($\zeta_A = \zeta_B = 1$), there still can be concentration fluctuations. Therefore, Γ_{AB} approaches unity slower than ζ does, and Γ_{AB}/Γ_B (or Γ_{AB}/Γ_A) may be smaller than the true ζ_A (or ζ_B). When the reaction is sufficiently fast, this estimation may predict a reaction rate slower than the true value. This will be seen later.

Another important case one often encounters is that in which one of the reactants is emitted into an environment containing a uniform concentration of the other. Let us say that concentration B_I has a uniform distribution as depicted in case 2 of Table 2.1. We have, in this case, $\overline{A_I B_I} = B_I \overline{A_I}$. Consequently, $\Gamma_B = \Gamma_{AB} = 1$, and $\mu_B = \mu_b = \mu_I = 1$. Using these relationships in (2.23), we obtain

$$\zeta_A = 1 \quad , \quad \zeta_B = \frac{\mu_a}{\Gamma_A} \qquad (2.25)$$

The correlation coefficient μ_a remains to be determined.

We note that μ_a is a function of concentration statistics within the diffusion zone v_{AB} . Since molecular diffusion is the most effective means of transport in the diffusion zone, μ_a is largely determined by molecular diffusion. We therefore propose that μ_a can be estimated to good approximation using the solution of the molecular diffusion equation

$$\frac{\partial a_I}{\partial t} = D_A \frac{\partial^2 a_I}{\partial x^2} \quad (2.26)$$

with initial condition

$$a_I(x,0) = \begin{cases} a_0, & \text{if } -\frac{1}{2} \delta \leq x \leq \frac{1}{2} \delta ; \\ 0, & \text{otherwise.} \end{cases} \quad (2.27)$$

The width δ is chosen to be the basic length scale of diffusion zones; i.e., $(\nu D^2/\epsilon)^{1/4}$. (This point is discussed further in the next subsection where we use essentially the same technique to estimate the reaction parameter λ .) The solution of (2.26) and (2.27) is (see for example, Carslaw and Jaeger, 1959).

$$a_I(x,t) = \frac{a_0}{2} \left[\operatorname{erf} \left(\frac{x + \frac{1}{2} \delta}{2\sqrt{D_A t}} \right) - \operatorname{erf} \left(\frac{x - \frac{1}{2} \delta}{2\sqrt{D_A t}} \right) \right] \quad (2.28)$$

Figure 2.2 shows a plot of μ_a as a function of nondimensionalized time $t D_A/\delta^2$ derived from the definition (2.21) of μ_a and (2.28). According to these results, for times larger than about $10^4(\delta^2/D_A)$ ζ_A and ζ_B are given by

$$\zeta_A = 1, \quad \zeta_B = \frac{1}{\Gamma_A} ; \quad t \geq 10^4 \delta^2/D_A . \quad (2.29)$$

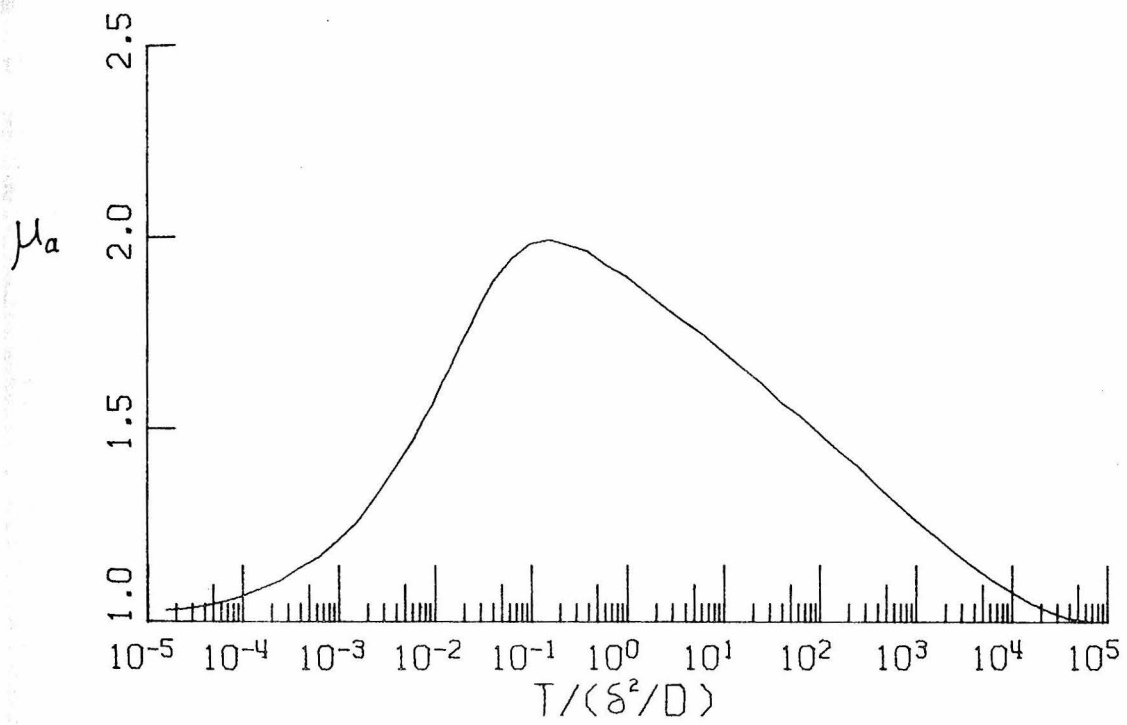


Figure 2.2 Temporal behavior of auto-correlation coefficient μ_a in a diffusion zone.

It turns out that these expressions are applicable to a wide range of problems of applied interest.

The approximate formulas for ζ_A and ζ_B developed in this section and their possible applications are summarized in Table 2.1. As discussed before, these approximations may result in slight underestimates of reaction rates. It is hoped that a more universal formula for the mixing parameters will be developed in the future. For practical interest, however, these approximations appear to be useful and reasonably accurate (cf. Sections II.3 and III.4).

Table 2.1 Summary of Approximation Formulas for the Mixing Parameters

| Model Concentration Profile | | Approximation Formulas | Possible Applications |
|---|-------------|---|--|
| $t = 0$ | $t > 0$ | | |
| <p>Legend: — a_I, - - - - - b_I</p> <p>(1a)</p> | <p>(1b)</p> | $\zeta_A = \frac{\Gamma_{AB}}{\Gamma_B}, \quad \zeta_B = \frac{\Gamma_{AB}}{\Gamma_A}$ | <p>Liquid-phase reactors with non-premixed feeds. Better when $v/D \gg 1$.</p> |
| <p>(2a)</p> | <p>(2b)</p> | $\zeta_A = 1, \quad \zeta_B = \frac{\mu_a}{\Gamma_A}$ <p>with μ_a given in Fig. 2.2 or taken as unity.</p> | <p>One reactant is introduced into a uniform mixture of another reactant. In atmosphere, μ_a may be taken as 1.</p> |

II.2.2 The Reaction Parameter

The reaction parameter λ is defined as the ratio of the concentration correlation coefficient in the reactive case to that in the inert case and depends on both molecular diffusion and chemical reaction. Thus, λ cannot be determined solely from inert mixing experiments and must be predicted. We note that λ is a function only of concentration statistics within the diffusion zone v_{AB} . Since turbulence can only bring A and B within a distance of the order of $(\nu D^2/\epsilon)^{1/4}$ * of each other, only molecular diffusion is capable of producing the intimate mixing required for reaction.

Let us first examine the limiting behavior of λ . When the rate of reaction is much slower than that of molecular diffusion, the reactant concentrations do not differ appreciably from A_I and B_I . In this case, λ is approximately unity. On the other hand, if the reaction rate is much faster than the mixing rate, reaction always tends to exhaust the reactants as soon as they are brought into the diffusion zone. Then, the concentrations of either A or B or both will be zero in v_{AB} . In this limit $\lambda \rightarrow 0$.

Regardless of how fast the reaction is, there always will exist a period of time, $0 \leq t \leq \tau_R$, where τ_R is an initial reaction time scale, during which the concentrations of the reactants will not have

*This length scale is $(\nu D^2/\epsilon)^{1/4}$ if $\nu \gg D$ and $(D^3/\epsilon)^{1/4}$ if $\nu \ll D$.
(Batchelor, 1959)

changed appreciably. In this period $A \cong A_I$ and $B \cong B_I$ and $\lambda = 1$. As the reaction and the diffusion proceed to completion, concentration gradients diminish and both μ and μ_I approach unity. Thus, as $t \rightarrow \infty$, λ should again approach unity. In the range $\tau_R < t < \infty$ we expect $\lambda < 1$.

For the purpose of estimating λ we propose to describe the concentration distributions in the diffusion zone by the following idealized representation (see Figure 2.3). Thin sheets of fluid containing A and B are initially separated by a distance of one sheet width. The sheet width δ is taken as the closest distance turbulence is capable of bringing two fluid elements together, i.e. $(\nu D^2/\epsilon)^{1/4}$ (assuming $D = 1/2(D_A + D_B)$ if the diffusivities are not equal). Let a , b , a_I and b_I denote the instantaneous concentrations of A and B in the reactive and inert cases, respectively. The equations governing these concentrations are

$$\frac{\partial a}{\partial t} = D_A \frac{\partial^2 a}{\partial x^2} - kab \quad (2.25)$$

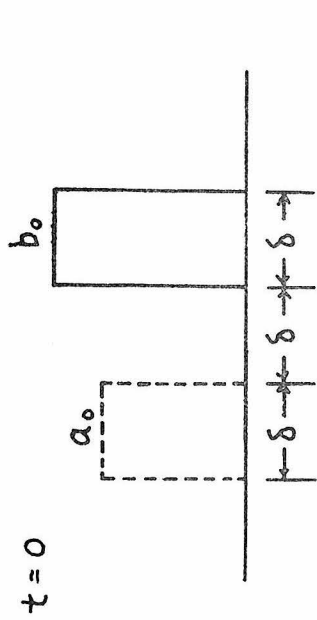
$$\frac{\partial b}{\partial t} = D_B \frac{\partial^2 b}{\partial x^2} - kab \quad (2.26)$$

$$\frac{\partial a_I}{\partial t} = D_A \frac{\partial^2 a_I}{\partial x^2} \quad (2.27)$$

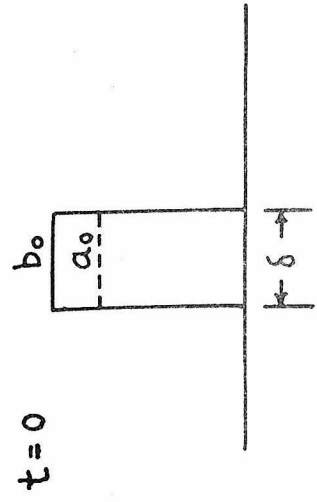
$$\frac{\partial b_I}{\partial t} = D_B \frac{\partial^2 b_I}{\partial x^2} \quad (2.28)$$

with initial and boundary conditions

$$a(x,0) = a_I(x,0) = \begin{cases} a_0 & -3/2 \delta \leq x \leq -1/2 \delta \\ 0, & \text{otherwise} \end{cases} \quad (2.29)$$



(a) Non-Premixed Reactants



(b) Premixed Reactants

$\beta = 1$ $\beta \neq 1$

Figure 2.3 Schematic concentration profiles in estimation of λ

$$b(x,0) = b_I(x,0) = \begin{cases} b_0, & 1/2 \delta \leq x \leq 3/2 \delta \\ 0, & \text{otherwise} \end{cases} \quad (2.30)$$

$$a(x,t) = b(x,t) = 0 \quad x \rightarrow \pm \infty \quad (2.31)$$

Defining the following dimensionless parameters,

$$\begin{aligned} a^* &= \frac{a}{a_0} & b^* &= \frac{b}{b_0} & a_I^* &= \frac{a_I}{a_0} & b_I^* &= \frac{b_I}{b_0} \\ x^* &= \frac{x}{\delta} & t^* &= \frac{\mathcal{D}_A t}{\delta^2} \\ \alpha &= \frac{k\delta^2 a_0}{\mathcal{D}_A} & \beta &= \frac{b_0}{a_0} & \gamma &= \frac{\mathcal{D}_B}{\mathcal{D}_A} \end{aligned}$$

(2.25)-(2.31) become

$$\frac{\partial a^*}{\partial t^*} = \frac{\partial^2 a^*}{\partial x^{*2}} - \alpha \beta a^* b^* \quad (2.32)$$

$$\frac{\partial b^*}{\partial t^*} = \gamma \frac{\partial^2 b^*}{\partial x^{*2}} - \alpha a^* b^* \quad (2.33)$$

$$\frac{\partial a_I^*}{\partial t^*} = \frac{\partial^2 a_I^*}{\partial x^{*2}} \quad (2.34)$$

$$\frac{\partial b_I^*}{\partial t^*} = \gamma \frac{\partial^2 b_I^*}{\partial x^{*2}} \quad (2.35)$$

$$a^*(x^*,0) = a_I^*(x^*,0) = \begin{cases} 1 & -\frac{3}{2} \leq x^* \leq -\frac{1}{2} \\ 0 & \text{otherwise} \end{cases} \quad (2.36)$$

$$b^*(x^*,0) = b_I^*(x^*,0) = \begin{cases} 1 & \frac{1}{2} \leq x^* \leq \frac{3}{2} \\ 0 & \text{otherwise} \end{cases} \quad (2.37)$$

$$a^*(x^*,t^*) = b^*(x^*,t^*) = 0 \quad x^* \rightarrow \pm \infty \quad (2.38)$$

The solutions for a_1^* and b_1^* are

$$a_1^*(x^*, t^*) = \frac{1}{2} \left[\operatorname{erf} \left(\frac{x^*+3/2}{2\sqrt{t^*}} \right) - \operatorname{erf} \left(\frac{x^*+1/2}{2\sqrt{t^*}} \right) \right] \quad (2.39)$$

$$b_1^*(x^*, t^*) = \frac{1}{2} \left[\operatorname{erf} \left(\frac{x^*-1/2}{2\sqrt{\gamma t^*}} \right) - \operatorname{erf} \left(\frac{x^*-3/2}{2\sqrt{\gamma t^*}} \right) \right] \quad (2.40)$$

Determination of a^* and b^* requires numerical solution of (2.32) and (2.33), the details of which are given in Appendix B. Once $a^*(x^*, t^*)$ and $b^*(x^*, t^*)$ have been determined, λ is computed as follows.

$$\lambda = \frac{\int_{L_{AB}} a^* b^* dx^* \int_{L_{AB}} a_1^* dx^* \int_{L_{AB}} b_1^* dx^*}{\int_{L_{AB}} a_1^* b_1^* dx^* \int_{L_{AB}} a^* dx^* \int_{L_{AB}} b^* dx^*} \quad (2.41)$$

The boundaries of the one-dimensional diffusion zone L_{AB} are chosen as the points where a_1^* or b_1^* has a value of 0.001. (See Appendix B for details.)

The behavior of λ as a function of t^* is shown in Figures 2.4-2.6 for the combinations of the parameters α , β , and γ summarized in

Table 2.2.

Table 2.2 Combinations of Parameters for λ vs. t^*

| Figure No. | α | β | γ |
|------------|------------|---------|----------|
| 2.4 | $0.1-10^4$ | 1 | 1 |
| 2.5 | 10^3 | 0.1-10 | 1 |
| 2.6 | 10^3 | 1 | 0.1-10 |

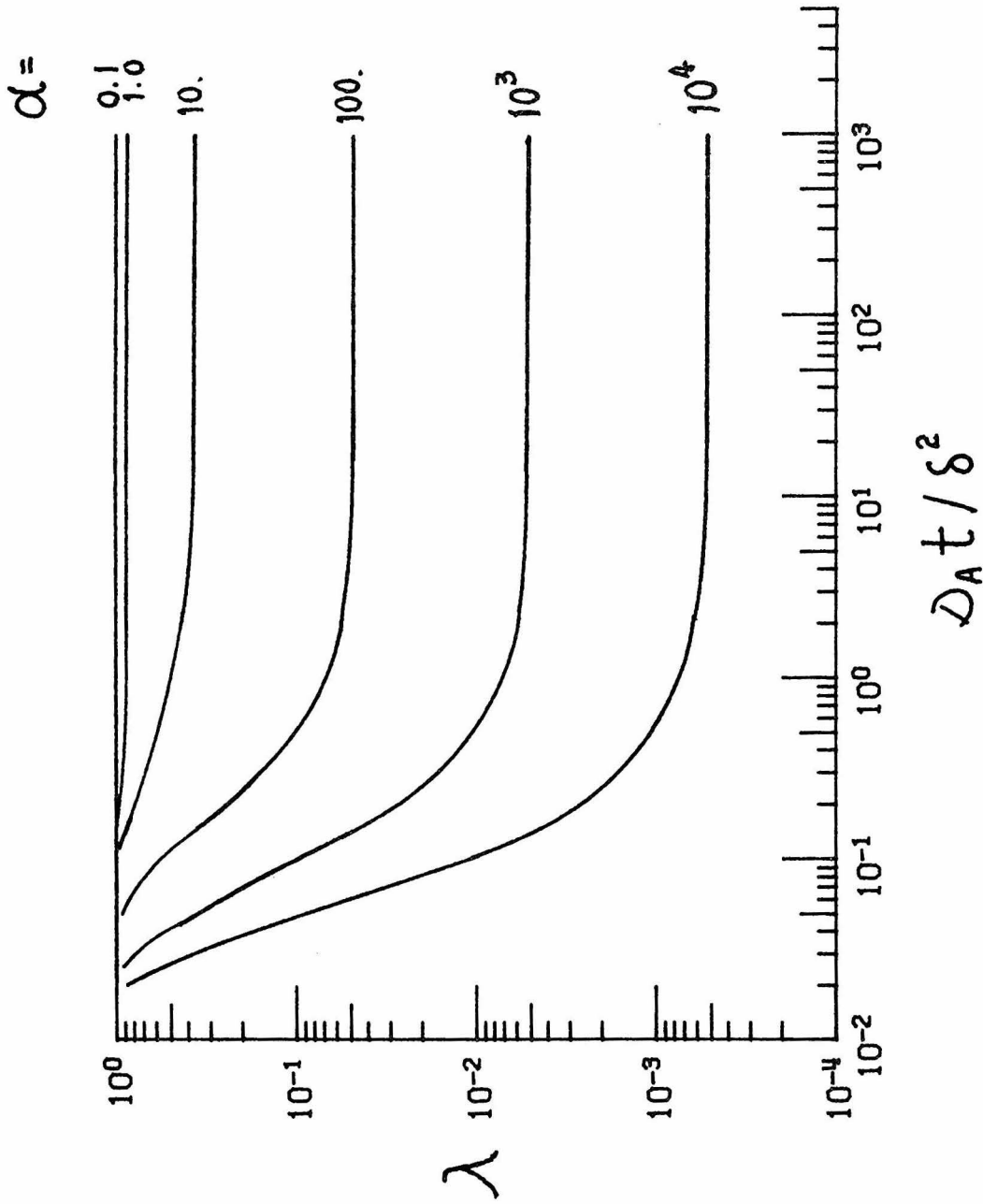


Figure 2.4 Temporal behavior of λ for various values of α at $\beta = \gamma = 1$.

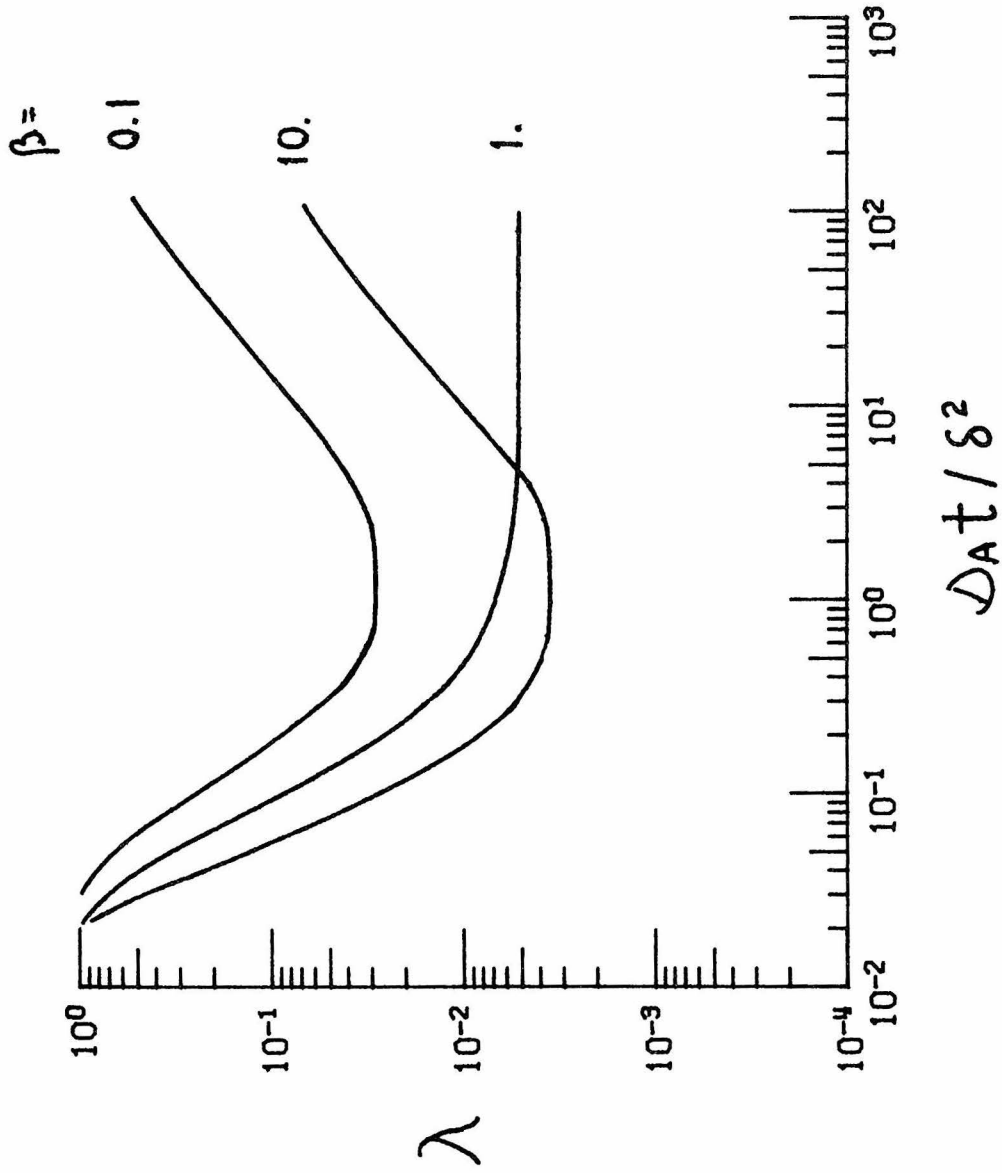


Figure 2.5 Temporal behavior of λ for various values of β at $\alpha = 1000$ and $\gamma = 1.$

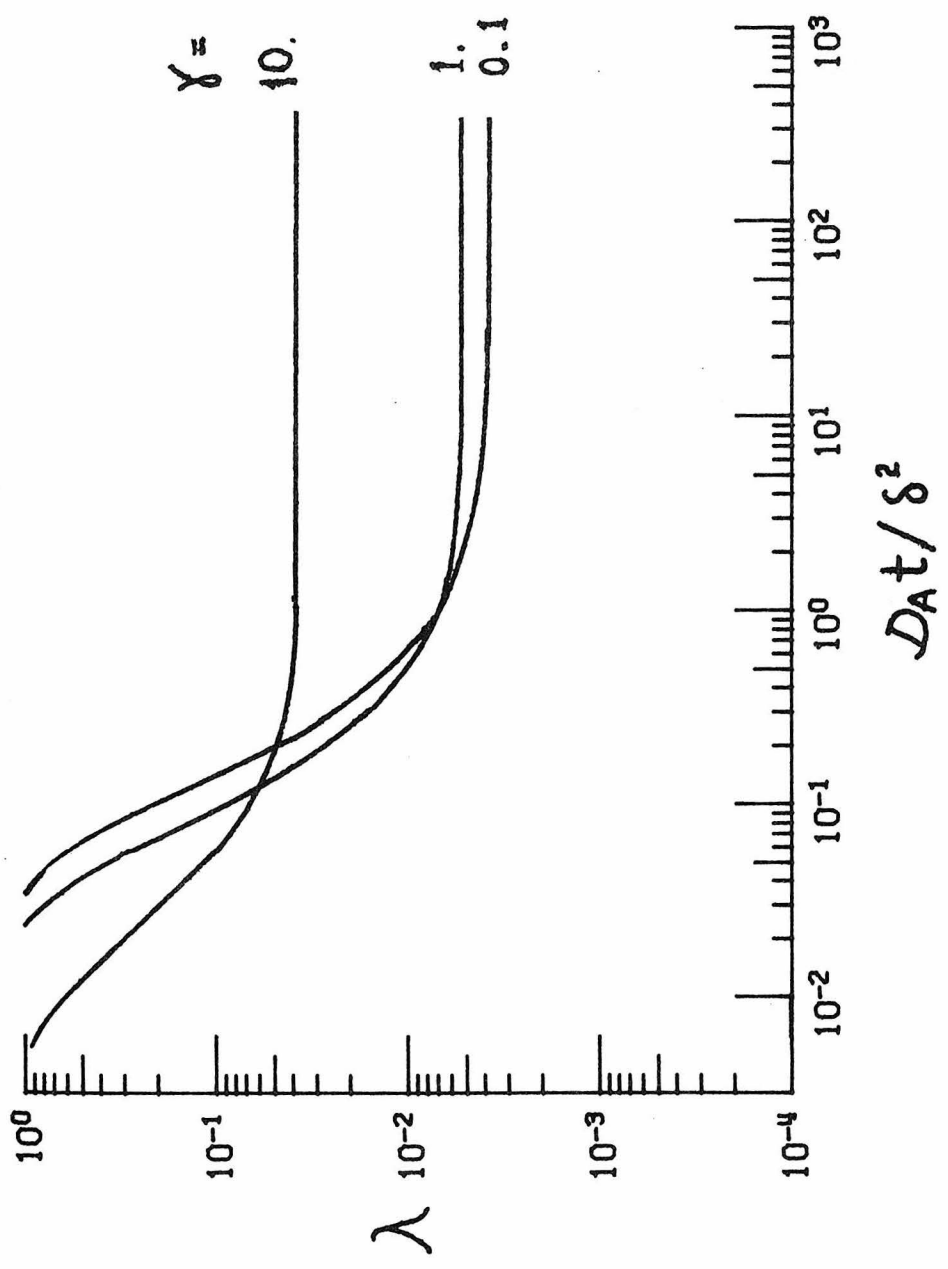


Figure 2.6 Temporal behavior of λ for various values of γ at $\alpha = 1000$ and $\beta = 1$.

The temporal behavior of λ is consistent with expectations, i.e. at $t^* = 0$, $\lambda = 1$; for $t^* > 0$, λ decreases; and as t^* becomes large, λ again approaches 1. This limiting behavior is not shown in Figures 2.4 and 2.6 because it happened much later in the case $\beta = 1$. In addition, the variation of λ vs. t^* with the parameters α , β , and γ can be discussed on a physical basis.

For $\alpha \leq 1$, $\lambda = 1$ for all t^* . When $\alpha \leq 1$, reaction is slow compared to molecular diffusion, the concentration profiles differ little from the inert profiles, and λ is always close to unity. As α increases beyond order unity, the reaction speed increases and λ decreases from its initial value of 1. λ is found to reach a minimum value over some period of time, the minimum value decreasing as α increases. Physically, as α increases, the reaction rate increases relative to the molecular diffusion rate, reactant concentrations within the diffusion zone decrease and become appreciably smaller than inert concentrations, and λ decreases. (Figure 2.4)

When $\beta = 1$ (stoichiometric feed), λ remains at its minimum value for a period of time of the order of $10^4(\delta^2/\mathcal{D}_A)$ (not shown in the figures). For $\beta \neq 1$, the period of time over which λ is at its minimum value is shorter than that for $\beta = 1$. This behavior occurs because in the case of non-stoichiometric feed, the reaction goes to completion faster than when $\beta = 1$. The λ_{\min} (the minimum value of λ) remains essentially the same in the case of $\beta > 1$ and the case of $\beta = 1$. For the

case when $\beta < 1$, λ_{\min} differs considerably from the value at $\beta = 1$. Closer examination reveals that λ_{\min} in the case of $\beta < 1$ has the same value as the λ_{\min} for the case where $\alpha = \frac{k\delta^2 b_0}{D_A}$, $\beta \geq 1$. This is because when $\beta < 1$, reactant B is present in a lesser amount and the reaction is limited by the time scale $(kb_0)^{-1}$ rather than $(ka_0)^{-1}$. (Compare the λ_{\min} in the curve of $\alpha = 1000$, $\beta = 0.1$ in Figure 2.5 with that in the curve of $\alpha = 100$, $\beta = 1$ in Figure 2.4.)

The effect of γ is shown in Figure 2.6. When $\gamma < 1$ ($D_B < D_A$) the behavior of λ does not deviate appreciably from that for the case when $\gamma = 1$. But when $\gamma > 1$ ($D_B > D_A$), λ reaches a minimum value which would be reached for the case where $\alpha = \frac{k\delta^2 a_0}{D_B}$. (Compare the curve for $\alpha = 1000$, $\gamma = 10$ in Figure 2.6 with the curve for $\alpha = 100$, $\gamma = 1$ in Figure 2.4.) It therefore becomes obvious that the controlling time scale for molecular diffusion is δ^2/D_B if $D_B > D_A$ and is δ^2/D_A if $D_A > D_B$. (Note that λ decays to its minimum in about $0.1 (\delta^2/D_A) = \delta^2/D_B$ in the case of $\gamma = 10$, rather than δ^2/D_A as in the case of $\gamma = 1$.)

Of most interest in the temporal behavior of λ is that region of time when λ is at its minimum value. The early decay of λ from 1 to λ_{\min} occurs on a time scale of the order of δ^2/D_A (if $D_A > D_B$). For a fast reaction (diffusion-limited) the reaction does not substantially commence until a time the order of δ^2/D_A , since this is the time required to bring the reactants into contact. For a slow reaction, the time scale δ^2/D_A required for λ to decay to λ_{\min} is short compared to the reaction time scale. In addition, in the basic equation (2.13), λ and Γ_{AB} occurs

as a product and since Γ_{AB} starts at a small value, the prediction should be insensitive to λ at small times. On the other hand, when λ starts increasing from λ_{\min} toward 1 as $t^* \rightarrow \infty$, the reaction has essentially gone to completion, and this region is of little concern. Thus the value of λ of most importance is the minimum value λ_{\min} . Based on the foregoing arguments, we will take λ to be a constant, λ_{\min} , for use in predicting the rate of decay of the reactive species.

Based on the values of λ_{\min} in Figure 2.4 ($\beta = \gamma = 1$), it is found that λ_{\min} can be described by a simple expression as a function of α :

$$\lambda = \lambda_{\min} = \frac{1}{1+0.16\alpha} \quad (2.41)$$

As we have already seen, the effect of β and γ on λ_{\min} is minor for $\beta \geq 1$ and $\gamma \leq 1$. Therefore we shall define α as the ratio of the shortest diffusion time scale to the longest reaction time scale. That is

$$\alpha = \frac{k\delta^2 \min[a_0, b_0]}{\max[\mathcal{D}_A, \mathcal{D}_B]} \quad (2.42)$$

where a_0, b_0 are feed concentrations of the reactive species A and B, respectively. Physically such a definition is justified because the diffusion process is dominated by the species with the larger diffusivity, while a second-order reaction is limited by the species with the smaller concentration.

If reaction is much faster than diffusion, i.e. $\alpha \rightarrow \infty$, $\alpha\lambda = 1/0.16$. Thus, in this case (2.13) becomes

$$\frac{d\bar{A}}{dt} = - \frac{D_{A^{\Gamma}AB}}{0.16a_0 \delta^2 \zeta_A \zeta_B} [\bar{A} - (1 - \zeta_A)\bar{A}_I][\bar{B} - (1 - \zeta_B)\bar{B}_I] \quad (2.43)$$

which, as might be expected, is independent of the rate constant k . Physically, in this case, the reaction rate is solely controlled by the rate of turbulent mixing and molecular diffusion. Only inert mixing data is required in (2.43). A similar result was obtained by Toor (1962) (see Appendix A). In that model it is predicted that the rate of conversion of a very rapid reaction in a tubular reactor is dependent on inert concentration statistics only. In the next section we shall compare the predictions of Toor's and other models with the present one.

It should be pointed out that in turbulence, quantities such as δ are often known only in terms of their orders of magnitude. Thus one might wonder how sensitive the prediction of the model is to the value of α . It will be shown later (Figures 2.13 to 2.18) that an estimate of α accurate to within one order of magnitude is sufficient to predict laboratory data within experimental error.

For the case of premixed reactants, λ is calculated by solving (2.25-28) with the set of initial conditions depicted in Figure 2.3b, p.39. The temporal behavior of λ in this case shows two distinct patterns. For the case of $\beta = 1$, λ stays at unity for all times; but for cases where $\beta \neq 1$, λ decays to the same minimum value as its non-premixed counterpart. Physically these behaviors can be understood with the aid of Figure 2.3(b). In the case of $\beta = 1$, which is equivalent to the case

of $A + A \rightarrow$ products, both μ and μ_I are autocorrelations of Gaussian type distributions. Consequently, $\mu = \mu_I$ and $\lambda = 1$. When the feed ratio is not stoichiometric, the reactant in excess concentration depletes the other reactant and creates a valley in its concentration profile (Figure 2.3b). The concentration correlation therefore becomes similar to the non-premixed case, and λ shares the same minimum value.

II.3 Application of the Diffusion Zone Model to Laboratory Data

An extensive experimental program has been conducted over a long period in the laboratories of Toor (Vassilatos and Toor, 1965; Mao, 1969) and Brodkey (McKelvey, 1968; Zakanycz, 1971) on chemical reactions occurring in turbulent mixing. Both groups have employed a multi-jet reactor originally described by Vassilatos and Toor (1965). The reactor and the nature of the experiments are described in Appendix A. In essence, in the reactor alternate jets of acid and base are ejected from a plate containing numerous small holes into a tube, creating efficient turbulent mixing. Conditions are such that the mean concentrations of A and B vary only with distance from the entrance. (In the inert case \bar{A}_I and \bar{B}_I are constant throughout the reactor.) The reactor affords the opportunity to measure the rate of reaction between two species in the presence of rather well-characterized turbulent mixing. In this section we apply the diffusion zone model to several experiments reported on the two multijet reactors in order to calculate the degree of conversion with distance down the reactors.

Since \bar{A}_I and \bar{B}_I are constant in the reactor, it is convenient to let $F_A = \bar{A}/\bar{A}_I$ and $F_B = \bar{B}/\bar{B}_I$, in which case (2.12) becomes

$$\frac{dF_A}{dt} = - \frac{k\lambda\Gamma_{AB}\bar{B}_I}{\zeta_A\zeta_B} [F_A - 1 + \zeta_A][F_B - 1 + \zeta_B] \quad (2.44)$$

Since the reactions studied in the multijet reactor were of the general form $A + nB \rightarrow$ products, we can write

$$\bar{B} = \bar{B}_I + n(\bar{A} - \bar{A}_I) \quad (2.45)$$

or

$$F_B = 1 + \frac{n\bar{A}_I}{\bar{B}_I} (F_A - 1) \quad (2.46)$$

Substituting (2.46) into (2.44) gives

$$\frac{dF_A}{dx} = - \frac{k\lambda\Gamma_{AB}n\bar{A}_I}{U(x)\zeta_A\zeta_B} [F_A - 1 + \zeta_A][F_A - 1 + \beta\zeta_B] \quad (2.47)$$

where $\beta = \bar{B}_I/n\bar{A}_I$ and the relationship $dx/dt = U(x)$ has been used to convert time to distance x down the reactor. $U(x)$ is the mean axial velocity at position x and is discussed in Appendix A.

Values of Γ_{AB} , Γ_A , and Γ_B required to evaluate ζ_A and ζ_B by (2.24) were obtained from conversion data for very rapid reactions in the case of $\beta = 1$. (see Appendix A.) Equation (2.41) was used to calculate λ . Table 2.3 lists the experiments that we will consider. Basically two types of reactions were studied. The first are those of

Table 2.3 Experimental Conditions for Data Used to Evaluate Diffusion Zone Model

| Species | | k | a _O | n | β | α [#] | Reference |
|------------------------|-----------------|---------------------|--|---|--------------------------------------|--------------------------------------|--|
| A | B | ℓ/g-mole-sec | g-mole/ℓ | | | | |
| CO ₂ | NaOH | 12400 | .00699 .00625 .00655 .00658 .00724 | 2 | 1.26 1.39 1.71 2.53 3.88 | .075 .067 .070 .071 .078 | Vassilatos and Toor (1965) Re = 3700* |
| CO ₂ | NaOH | 8320 | .03331 .03405 .03387 | 2 | 1.0 1.5 3.0 | .241 .246 .245 | Mao (1969) Re = 1460 |
| CO ₂ | NaOH | 8320 | .03357 .03363 .03335 | 2 | 1.0 1.5 3.0 | .243 .243 .241 | Mao (1969) Re = 2430 |
| Nitrilo-triacetic Acid | OH ⁻ | 1.4x10 ⁷ | .02910 .02830 .02720 | 1 | 1.0 1.5 3.0 | 177.1 172.3 165.5 | Mao (1969) Re = 2430 |

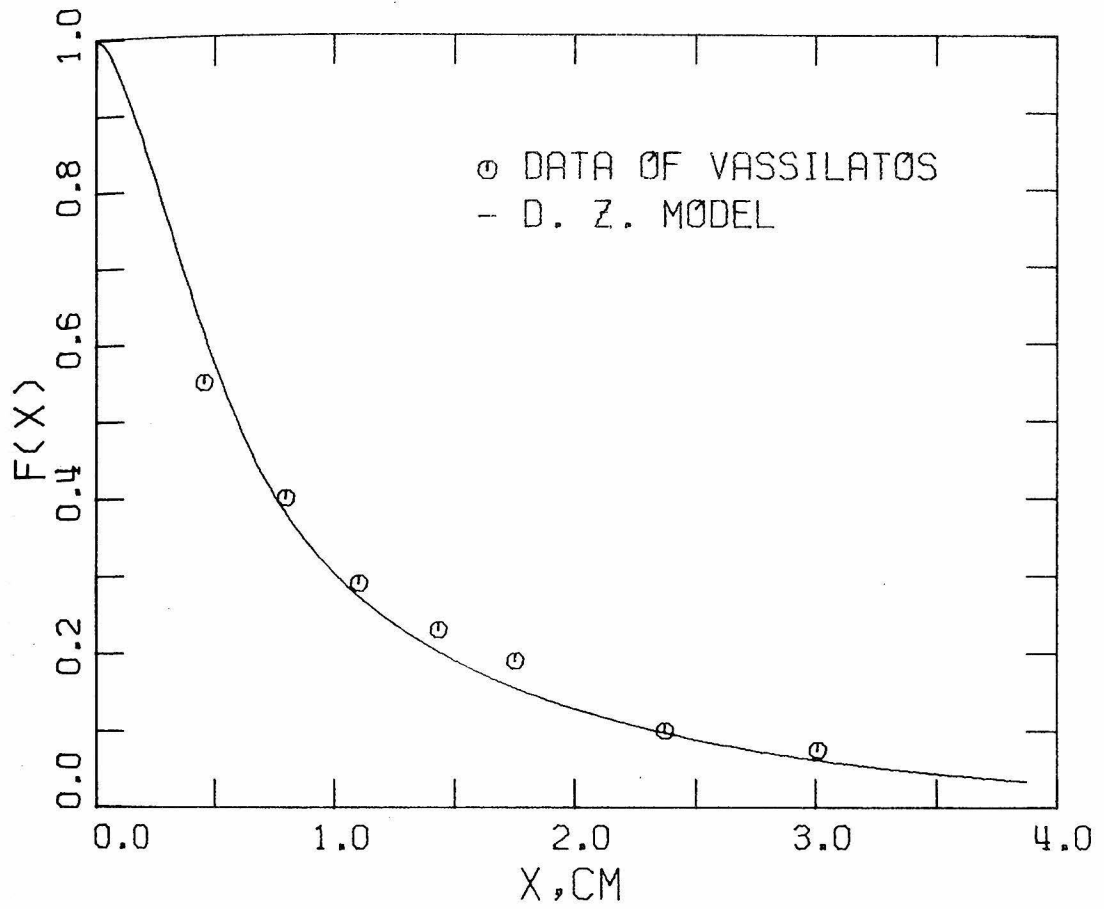
$\alpha = k n a_0 \delta^2 / D$, where $D = D_A = D_B = 2.3 \times 10^{-5}$ cm/sec, $\delta \approx 10^{-4}$ cm from Yieh (1970)

* Reynolds number is based on nozzle measurements.

so-called intermediate rates, having a value of α of 0.01-1. The second type are very fast reactions ($\alpha \gg 1$). The laboratory data reported of this type are independent of the rate constant k (diffusion-limited), and some of them have been used to compute Γ_{AB} etc. Slow reactions ($\alpha \ll 1$) are not of interest to us because they can be represented by the homogeneous law (eq. (1.16)).

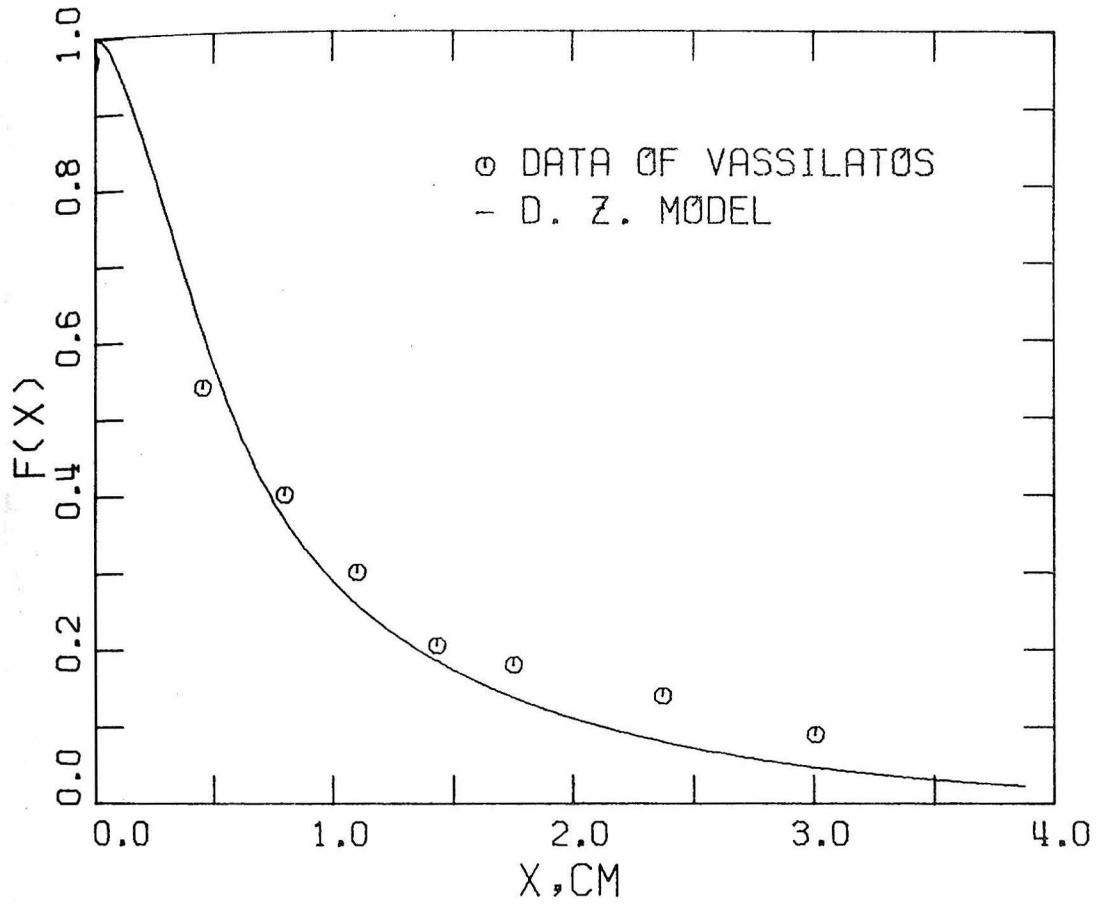
In the calculation (2.46) was solved numerically to predict F_A as a function of distance from the reactor inlet. The results of the calculations for the experiments listed in Table 2.3 are given in Figures 2.7-2.20. In each case, information on inert concentration correlations was obtained using Toor's model, as described in Appendix A. Hence, any errors inherent in Γ_{AB} , Γ_A , and Γ_B will be carried into our calculation. In general, the predictions agree more closely with the data of Vassilatos and Toor (1965), Figures 2.7-2.11, than with those of Mao (1969), Figures 2.12-2.20. In most cases it was found that the predicted concentration of \bar{A} was somewhat higher than the observed for the data of Mao (1969). Since negative values of $F(x)$ were reported in several sets of Mao's data, it is possible that his measurements were in error (Figures 2.14, 17, 19, 20). In addition, as we have indicated, the approximation (2.24) will generally give a slower decay of the species and may have been the source of error.

One of the most important features of our diffusion zone (DZ) model revealed in these calculations is the effect of inhomogeneous mixing on the rate of reaction. Comparison with the predictions of the



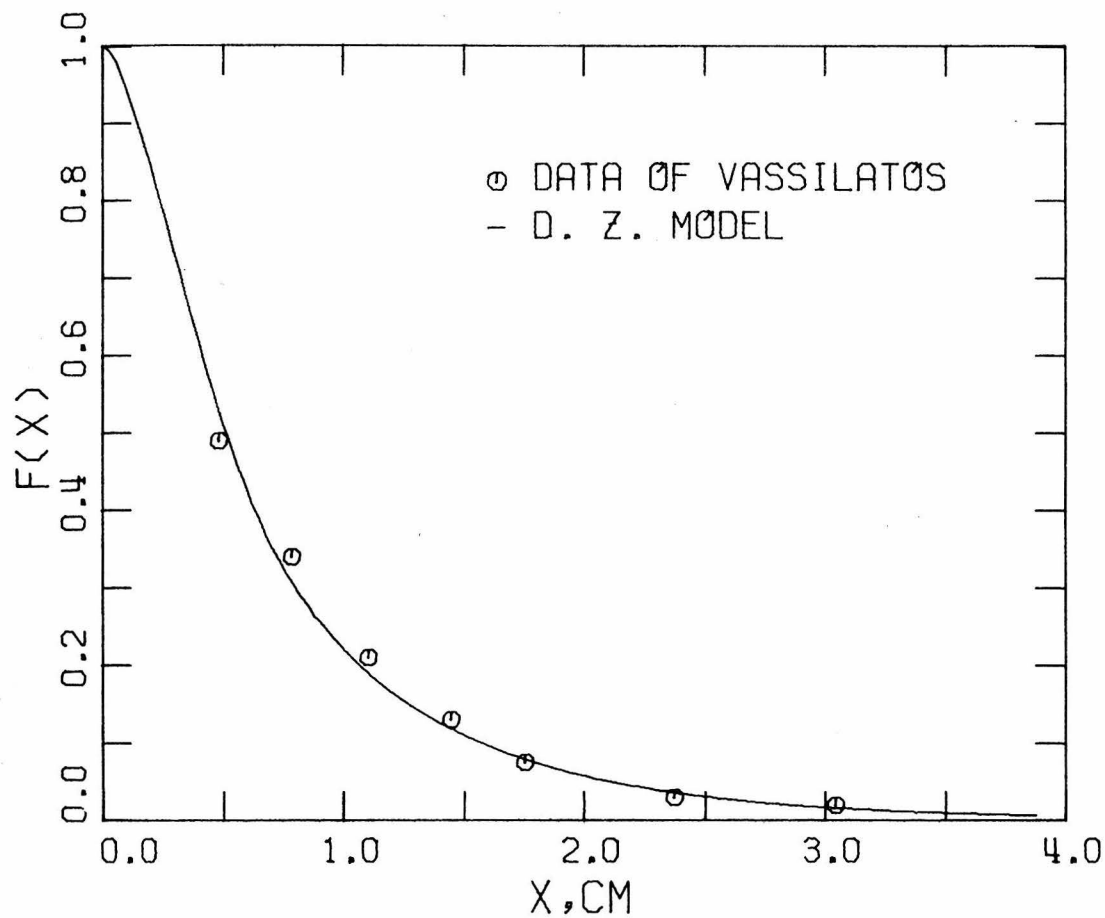
K = 0.124E 05 LIT/MØL-SEC
A0 = 0.00699 MØL/LIT
BETA = 1.26

Figure 2.7 Comparison of DZ model prediction with reactor data.



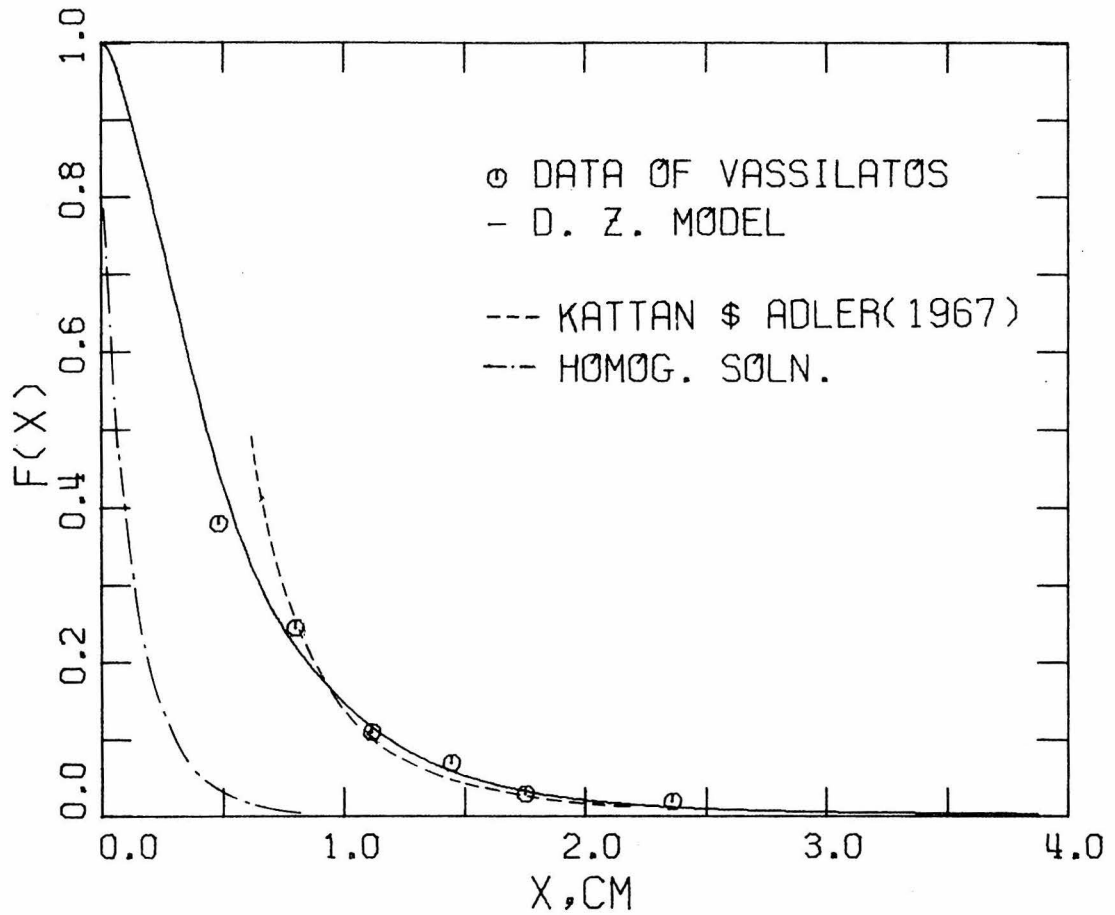
K = 0.124E 05 LIT/MØL-SEC
A0 = 0.00625 MØL/LIT
BETA = 1.39

Figure 2.8 Comparison of DZ model prediction with reactor data.



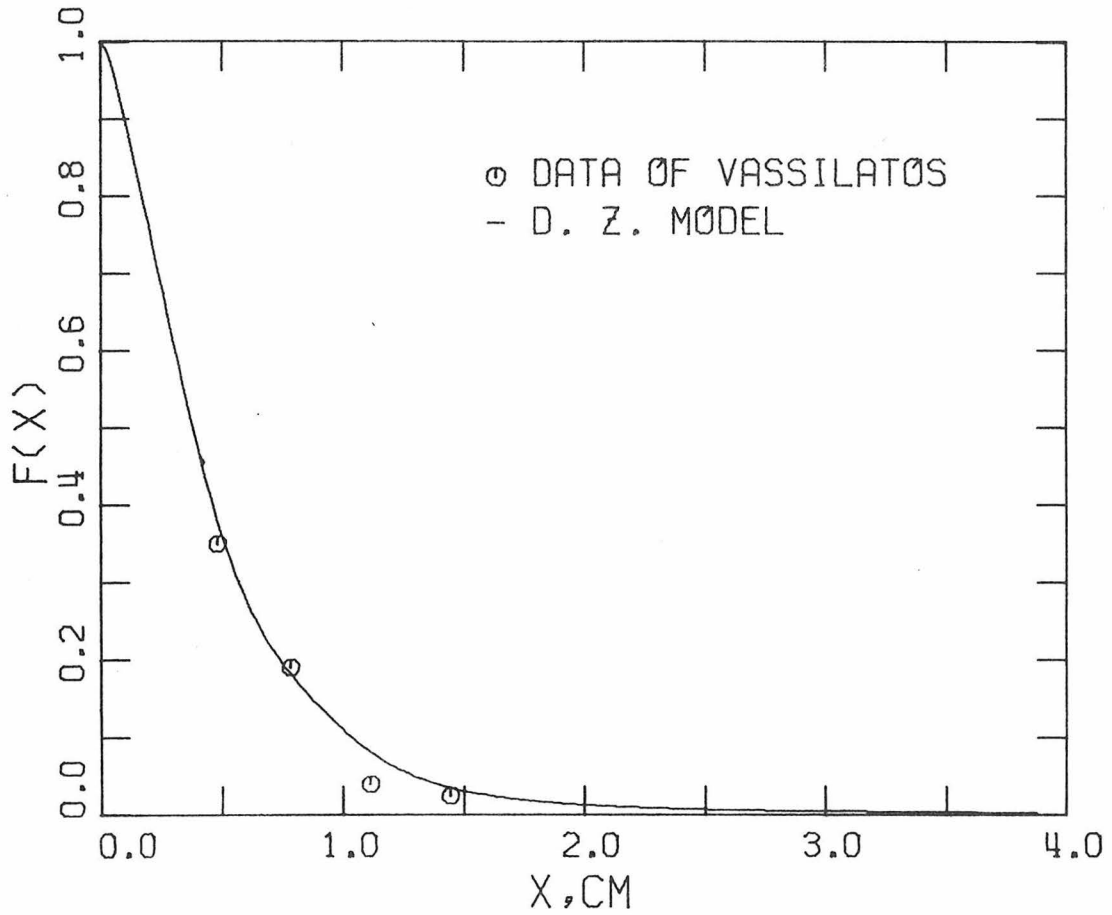
K = 0.124E 05 LIT/MOL-SEC
A0 = 0.00655 MOL/LIT
BETA = 1.71

Figure 2.9 Comparison of DZ model prediction with reactor data.



K = 0.124E 05 LIT/MØL-SEC
AO = 0.00658 MØL/LIT
BETA = 2.52

Figure 2.10 Comparison of DZ model prediction with reactor data and other models.



K = 0.124E 05 LIT/MØL-SEC
AO = 0.00724 MØL/LIT
BETA = 3.88

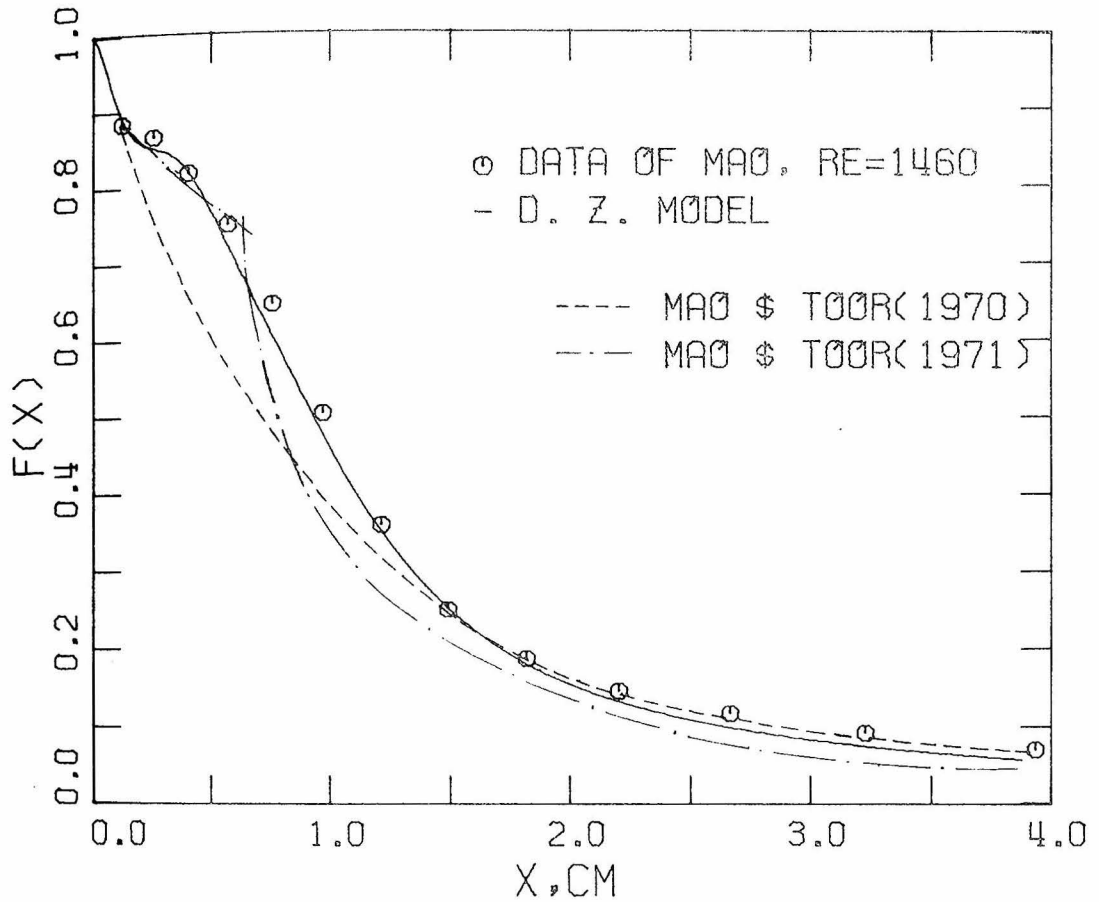
Figure 2.11 Comparison of DZ model prediction with reactor data.

homogeneous law, i.e.

$$U(x) \frac{d\bar{A}}{dx} = -k\bar{A}\bar{B} \quad (2.48)$$

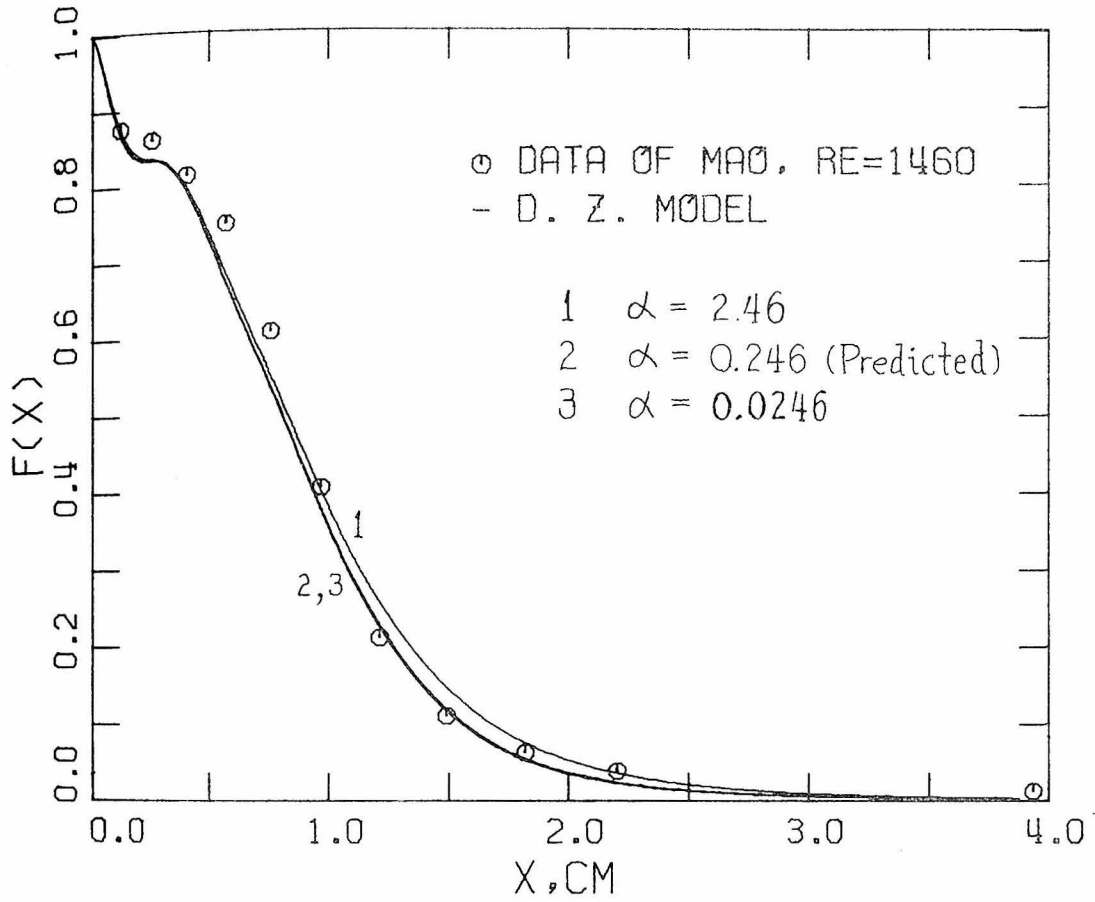
are shown in several cases. (Figures 2.10, 15-16). It is seen that the inhomogeneity in mixing has inhibited the reaction considerably. Furthermore, the DZ-model predictions are very sensitive to the mixing condition in the reactor. This is seen particularly in Figures 2.12-14 where a break in calculated F_A occurs around $x = 0.2$ cm. Comparing these results with the data used for Γ_{AB} , Γ_A , etc (Figure A.3), we conclude that this break is solely a reflection of the mixing condition. Physically, this breaking point is where the jets from reactor head meet. Considerable backflow is induced here and the mixing is temporarily slowed. This effect is not shown in Vassilatos' data because data were not taken close enough to the head region. With these data it was necessary to extrapolate the data back into the region of the reactor head (see Fig. A.2). The data gathered by Mao (1969) from the laminar jets ($Re = 1460$) (Fig. A.3) reveal a much slower rate of mixing than the data from the turbulent jets ($Re = 2430$) (Fig. A.4). All these characteristics in the mixing process are reflected in the DZ model predictions.

The sensitivity of the model prediction to estimates of parameters such as α is shown in Figures 2.13 and 2.18. In the case of intermediate rate reactions, Figure 2.13 indicates that variations of an order of magnitude in the value of α result in negligible changes in the



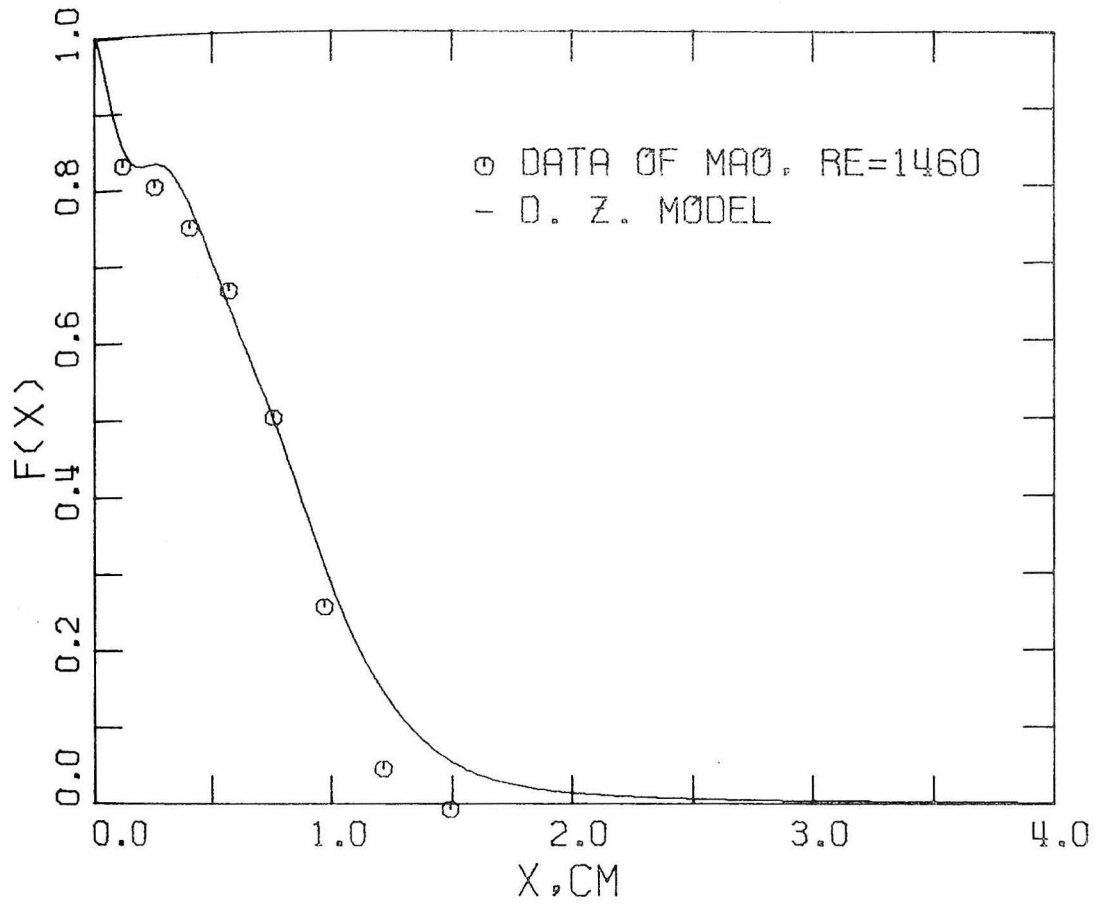
K = 0.832E 04 LIT/MOL-SEC
AO = 0.03331 MOL/LIT
BETA = 1.00

Figure 2.12 Comparison of DZ model prediction with reactor data and other models.



K = 0.832E 04 LIT/MOL-SEC
AO = 0.03405 MOL/LIT
BETA = 1.50

Figure 2.13 Comparison of DZ model prediction with reactor data. The model is not very sensitive to estimates of its parameter α .



K = 0.832E 04 LIT/MOL-SEC
AO = 0.03387 MOL/LIT
BETA = 3.00

Figure 2.14 Comparison of DZ model prediction with reactor data.

predicted mean concentration. The prediction of λ in the case of very fast reactions is more sensitive to the value of α . However, in this limit the reaction is diffusion-controlled and the influence of the reaction parameter λ is small. We therefore see that again variations of one order of magnitude in α change the conversion rate by only a small factor (Fig. 2.18).

Some previous work is discussed here. Kattan and Adler (1967) developed a so-called stochastic mixing model in which fluid elements containing reactive species are assumed to interact through a random process of coagulation and redispersion. The rate of this coagulation-redispersion process is adjusted to match the inert mixing data. Chemical reaction is incorporated into the model by changing the concentration of the species in each fluid element in accordance with the governing kinetics. Kattan and Adler (1967) compared the prediction of their model with Vassilatos' data and obtained results such as those plotted in Figure 2.10. It is seen that their model cannot describe the entire range of the chemical conversion process.

The model of Mao and Toor (1970) (see also Mao, 1969) divides the fluid into slabs of thickness δ which are alternately arranged with reactants A and B at constant initial concentrations. Since only molecular diffusion is allowed between the slabs, the set of model equations is very similar to (2.25-28). The thickness δ is first adjusted so that the solution of the system fits the data of a very rapid reaction in the same turbulence. The resulting value of δ is then used to

predict conversions of other reactions in the same turbulence. Rao and Edwards (1971) have shown that the model of Mao and Toor (1970) is in essence the same as the model of Kattan and Adler (1967). One of the calculations of Mao and Toor (1970) is given in Figure 2.12 in comparison with our DZ-model. The deviation of their model from the observed data is quite obvious. Mao and Toor (1971) later proposed a second model in which the reactor is divided into two regions. The first region extends from the head of the reactor to the point where the jets meet. Because of the considerable backflow in this region, they treat this domain as a homogeneous mixer, (2.48). The second region covers the remainder of the reactor and is modeled using the hypothesis of Toor (1969) $\overline{A'B'} = \overline{A'_1B'_1}$. Under this hypothesis,

$$U \frac{d\bar{A}}{dx} = -k(\bar{A}\bar{B} + \overline{A'_1B'_1}) \quad (2.49)$$

where $\overline{A'_1B'_1}$ is derived from very rapid reaction data as in Appendix A, and U is the constant bulk average velocity of the flow. One of the calculations of Mao and Toor (1971) is given in Figure 2.12. The prediction is not in good agreement with experimental data. Moreover, since the division between the two regions of the reactor cannot be determined a priori, this model is awkward for many purposes.

Yieh (1970) discovered that the model of Mao and Toor (1971), eq. (2.49), is improved if U is measured as a function of x rather than as a constant bulk velocity. Yieh assumed that the product concentration in a multijet reactor is uniformly distributed across the

reactor. That is, if C is the instantaneous concentration of the product,

$$C' \equiv C - \bar{C} = 0 \quad (2.50)$$

Since mass conservation requires that

$$A + B + C = 0,$$

we must have

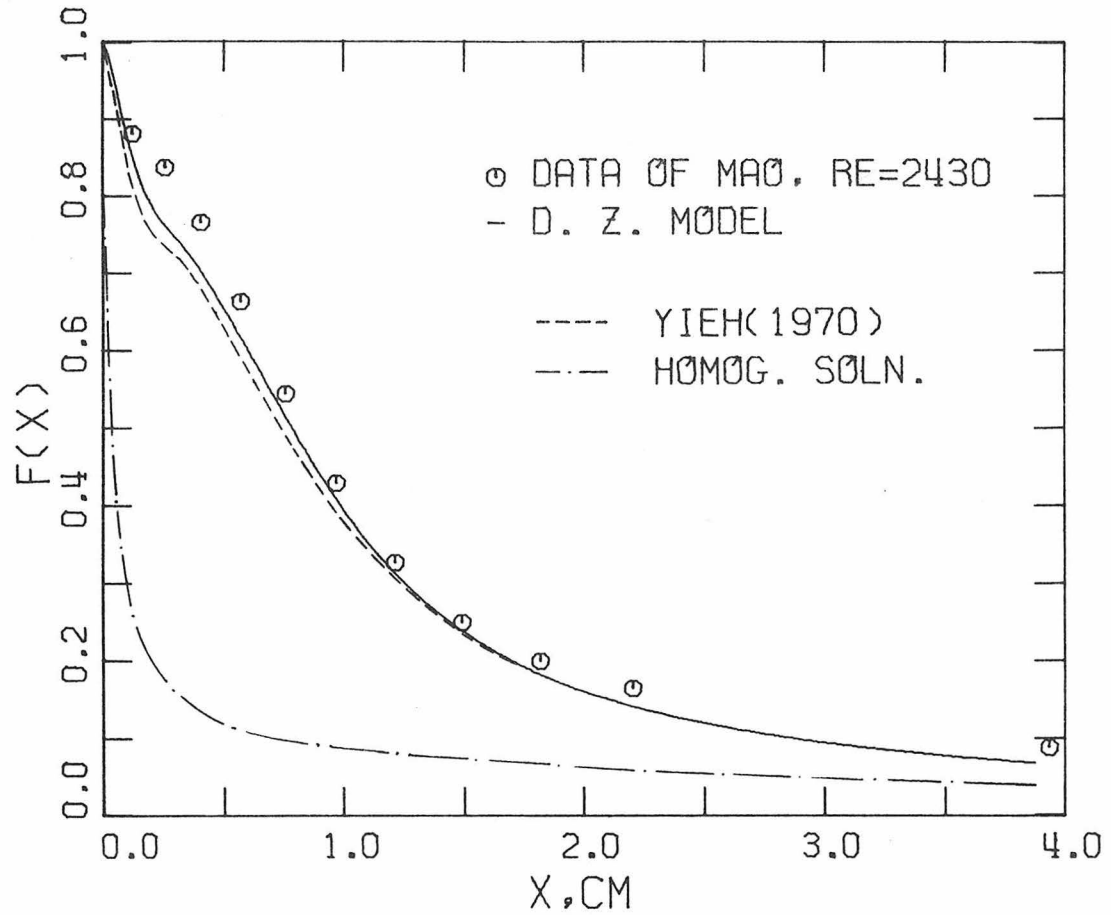
$$A' + B' = -C' = 0.$$

This result causes the equation governing $\overline{A'B'}$ (eq. (1.15)) to reduce to a form that is independent of the reaction kinetics. Yieh (1970) thus concluded that

$$\overline{A'B'} = \overline{A'_I B'_I}$$

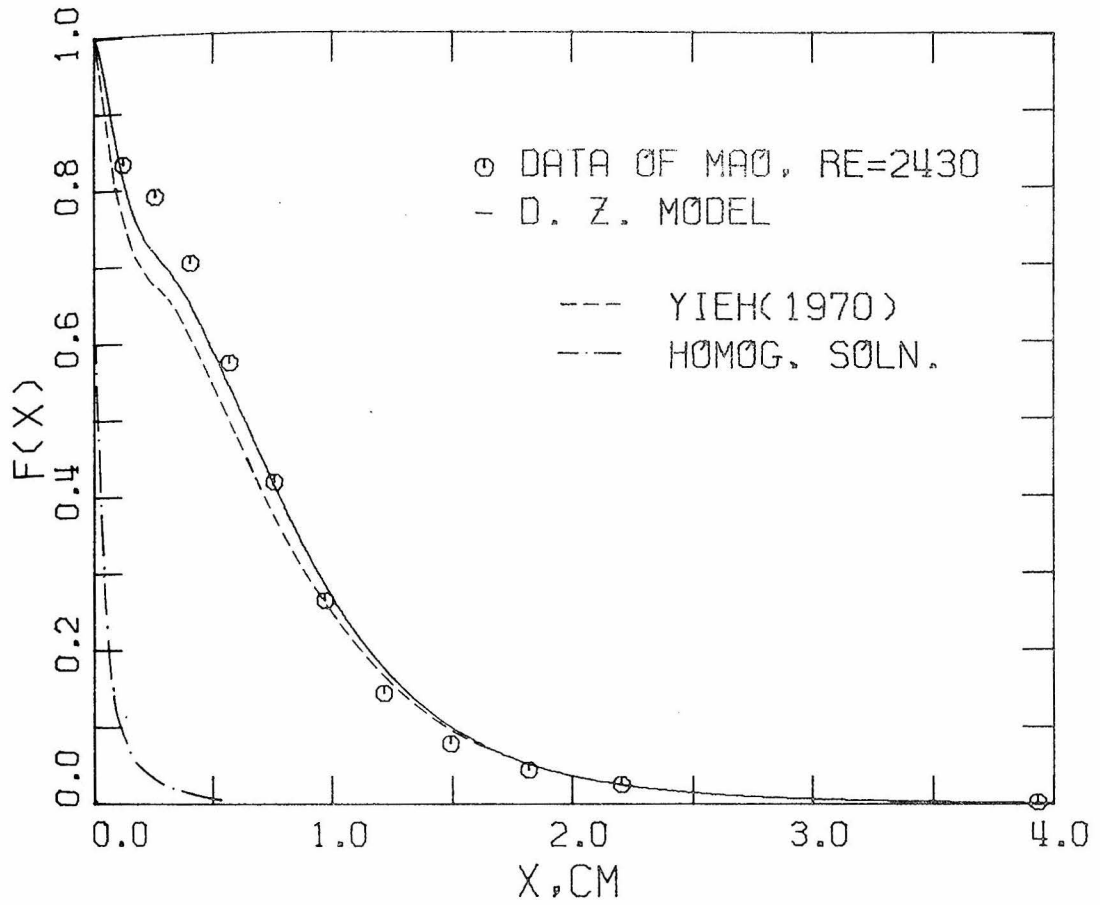
and arrived at (2.49). Yieh's predictions based on the same velocity profile $U(x)$ as we used give good agreement with observations. (Figures 2.15-16). However, assumption (2.50) is not generally valid and therefore Yieh's model is greatly limited in its scope of applications.

Applications of the diffusion zone model to very rapid reactions ($\alpha \gg 1$) are shown in Figures 2.18-20. Relatively good agreement between the predicted and observed conversion rates results in all three cases. Toor (1962) derived a model for very rapid reactions in



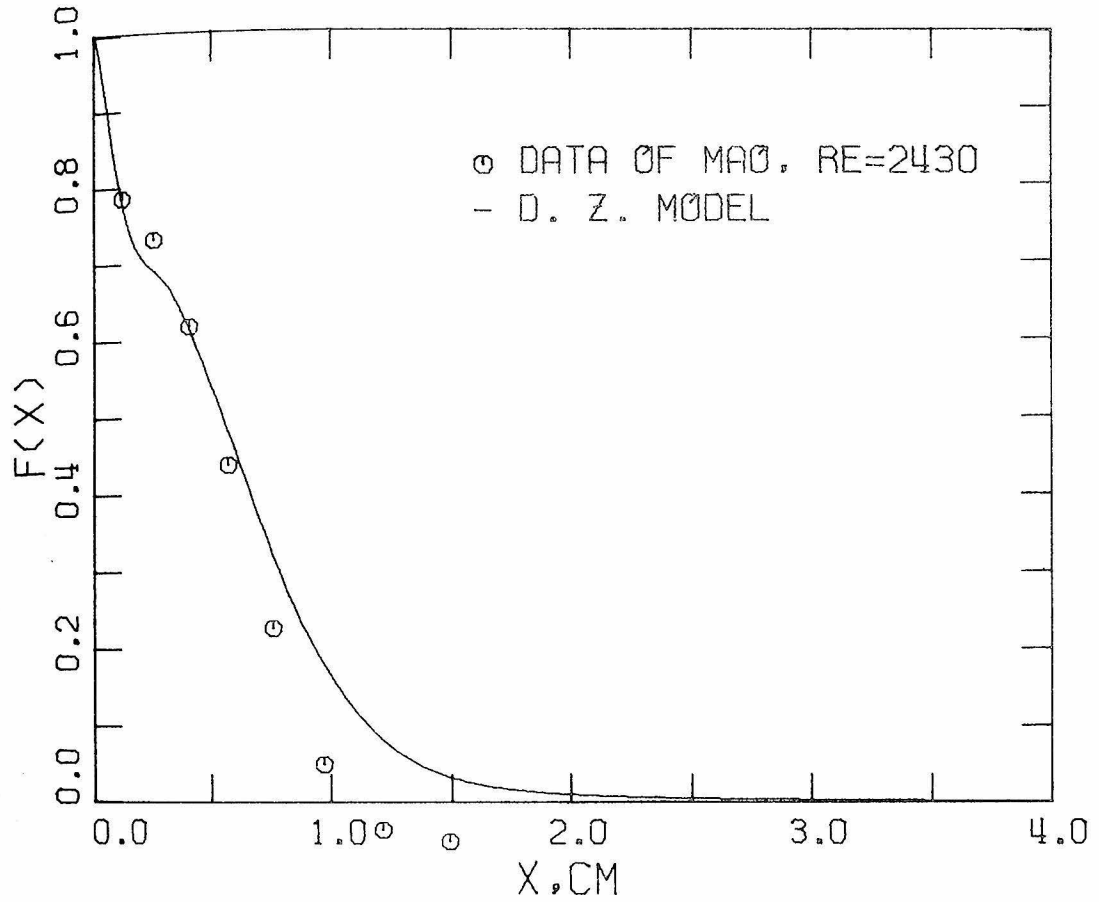
K = 0.832E 04 LIT/MOL-SEC
AO = 0.03357 MOL/LIT
BETA = 1.00

Figure 2.15 Comparison of DZ model prediction with reactor data and other models.



K = 0.832E 04 LIT/MOL-SEC
AO = 0.03363 MOL/LIT
BETA = 1.50

Figure 2.16 Comparison of DZ model prediction with reactor data and other models.



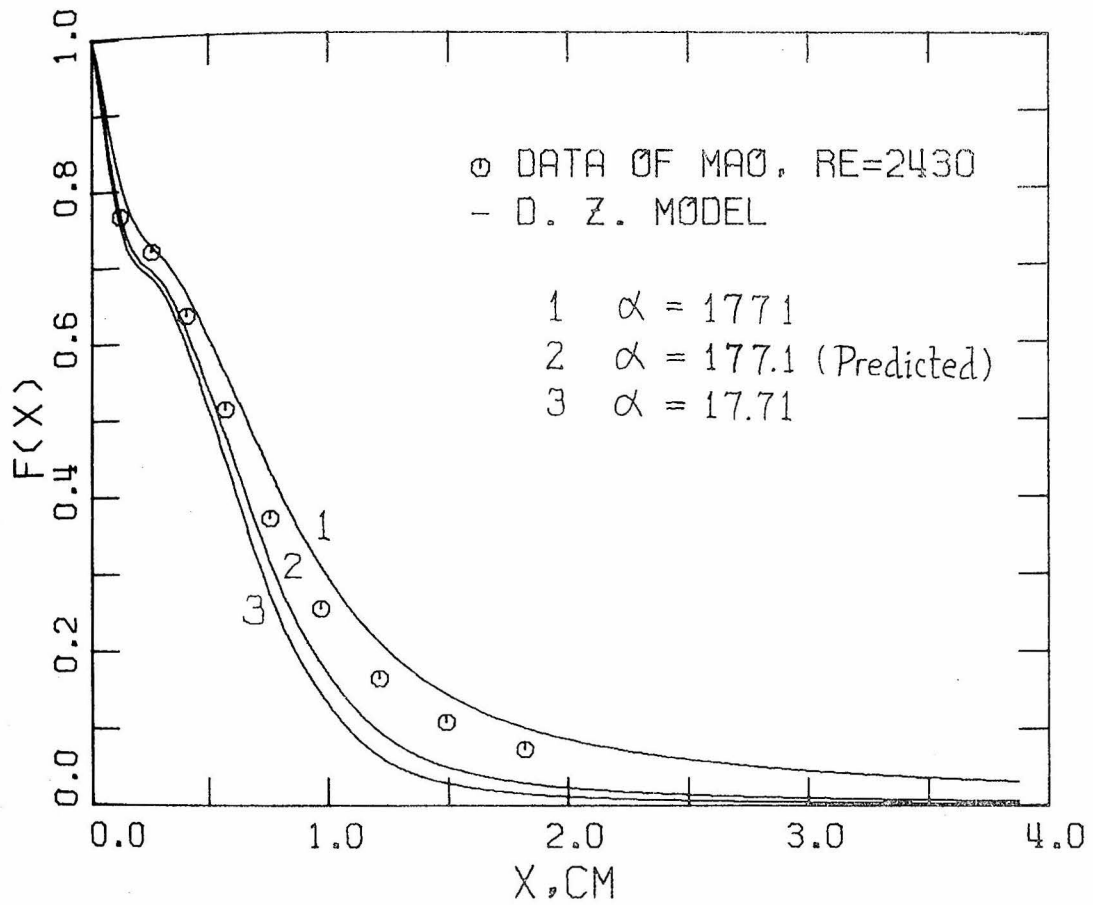
K = 0.832E 04 LIT/MOL-SEC
AO = 0.03335 MOL/LIT
BETA = 3.00

Figure 2.17 Comparison of DZ model prediction with reactor data.

which the conversion depends only on inert concentration statistics (Appendix A). Since we have used the experimental data in Figure 2.18 and Toor's model to estimate the values of Γ_{AB} etc. required in the DZ model, a comparison with Toor's model for this case ($\beta = 1$) is not possible. However, the comparison of predictions of both models in the case of non-stoichiometric feed are shown in Figures 2.19-20. The calculations of Toor's model were done by Mao (1969). The difference between the two models seems to be minor.

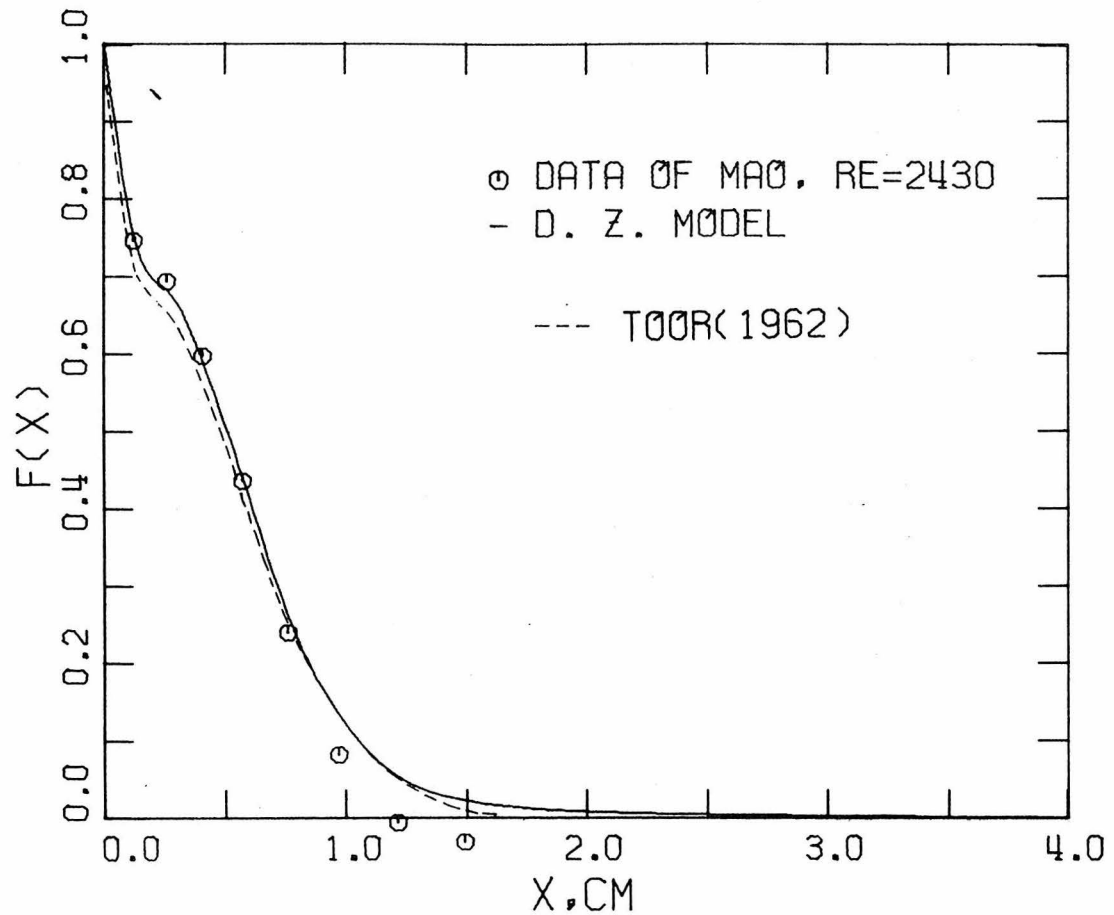
We have therefore demonstrated that the diffusion-zone model provides a realistic description of the rate of decay of chemical species reacting in a given turbulence. This model applies to systems of any feed concentrations, any reaction speed and any turbulence, provided that the inert concentration statistics in that turbulence are given. The DZ model does not require adjustment of any of its parameters and estimate of its parameters can be obtained from turbulence data such as rate of energy dissipation ϵ , etc, that are usually available. We also note that the solutions of the model equations are simple to obtain numerically, in contrast to some currently available models.

Having demonstrated the applicability of the DZ model to reactor data, we consider in the next chapter applications to nonlinear chemical processes in the atmospheric boundary layer.



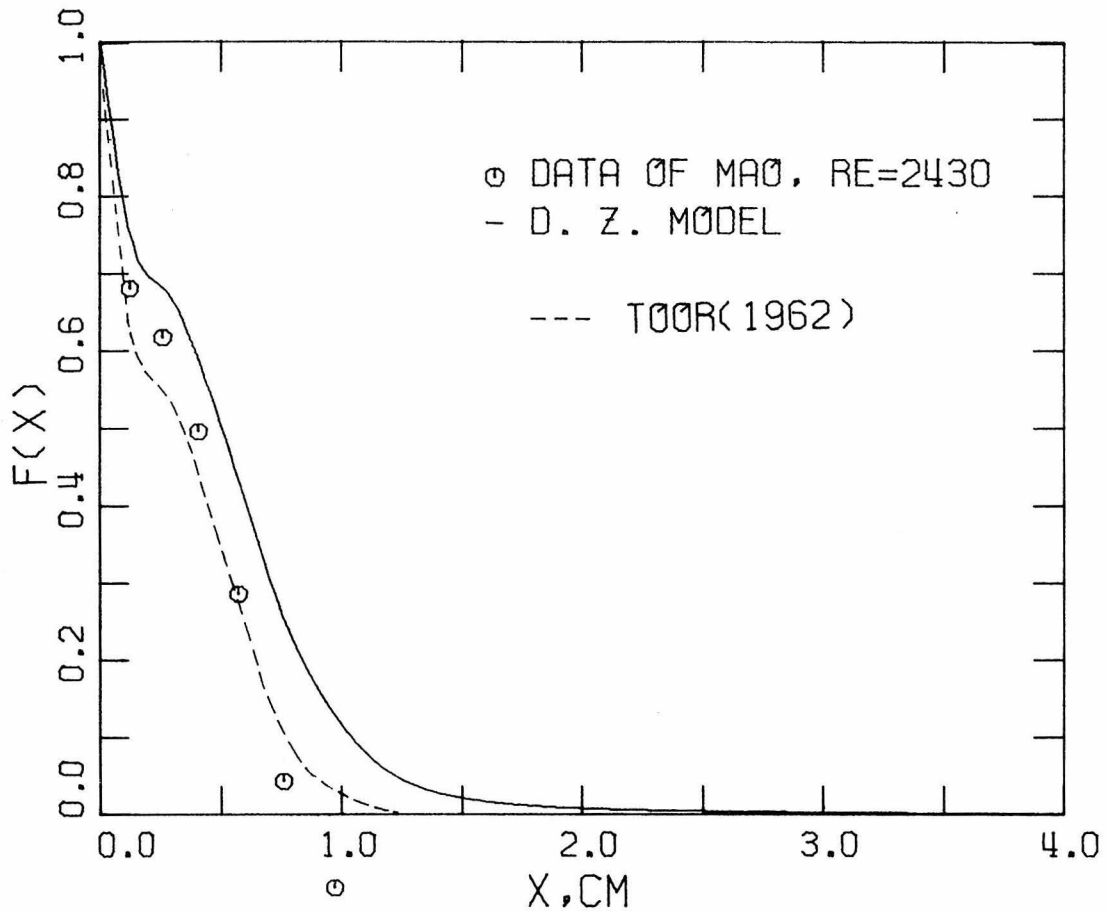
K = 0.140E 08 LIT/MOL-SEC
AO = 0.02910 MOL/LIT
BETA = 1.00

Figure 2.18 Comparison of DZ model prediction with reactor data in the case of very rapid reactions. The model is not very sensitive to estimates of its parameter α .



K = 0.140E 08 LIT/MOL-SEC
AO = 0.02830 MOL/LIT
BETA = 1.50

Figure 2.19 Comparison of DZ model prediction with reactor data of a very rapid reaction. Toor's model was calculated by Mao(1969).



K = 0.140E 08 LIT/MOL-SEC
AO = 0.02720 MOL/LIT
BETA = 3.00

Figure 2.20 Comparison of DZ model prediction with reactor data of a very rapid reaction. Toor's model was calculated by Mao(1969).

II.4 Generalization of the Diffusion Zone Model to Three Spatial Dimensions

For conceptual simplicity, the diffusion zone model was developed for the case of a spatially homogeneous system. In this section the model is generalized to these dimensions. The basic equation governing the ensemble mean concentration of a species A in a turbulent fluid is (see Chapter I)

$$\frac{\partial \langle A \rangle}{\partial t} + \langle u_i \rangle \frac{\partial \langle A \rangle}{\partial x_i} = - \frac{\partial}{\partial x} \langle u_i' A' \rangle + D_A \frac{\partial^2 \langle A \rangle}{\partial x_i^2} - k \langle AB \rangle \quad (1.9)$$

Assume the process is stationary so that the ensemble average can be replaced by a time average, \bar{A} . If K-theory is used to represent the turbulent mass flux and molecular diffusion flux is neglected, we obtain

$$\frac{\partial \bar{A}}{\partial t} + \bar{u}_i \frac{\partial \bar{A}}{\partial x_i} = \frac{\partial}{\partial x_i} \left(K_i \frac{\partial \bar{A}}{\partial x_i} \right) - k \bar{A} \bar{B} \quad (2.51)$$

where at a given location \tilde{x} , the time average is defined as

$$\bar{A}(\tilde{x}) = \lim_{T \rightarrow \infty} \frac{1}{T} \int_{t_0}^{t_0+T} A(\tilde{x}, t) dt \quad (2.52)$$

Now let us define τ_{AB} , similar to ν_{AB} , as the sum of periods of time within the sample period T when A and B are found together at \tilde{x} . Similarly, τ_A and τ_B are the sums of the time periods in T when A and B, respectively, exist alone at \tilde{x} . All these quantities are defined by inert measurements. Thus, we can decompose (2.52) into

$$\bar{A}(x) = \frac{1}{T} (\hat{A}_I \tau_A + \hat{a} \tau_{AB})$$

where

$$\hat{A}_I = \frac{1}{\tau_A} \int_{\tau_A} A dt \quad \hat{a} = \frac{1}{\tau_{AB}} \int_{\tau_{AB}} A dt$$

with similar definitions for \bar{B} , \hat{B} , and \hat{b} . Also, we let

$$\bar{AB} = \frac{1}{T} \mu \hat{a} \hat{b} \tau_{AB}$$

where

$$\mu = \left[\frac{1}{\tau_{AB}} \int_{\tau_{AB}} AB dt \right] \frac{1}{\hat{a} \hat{b}}$$

Likewise we can define the corresponding inert quantities:

$$\bar{A}_I = \frac{1}{T} (\hat{A}_I \tau_A + \hat{a}_I \tau_{AB})$$

$$\bar{\Gamma}_{AB} = \frac{1}{T} \mu_I \hat{a}_I \hat{b}_I \tau_{AB}$$

where

$$\hat{a}_I = \frac{1}{\tau_{AB}} \int_{\tau_{AB}} A_I dt$$

$$\mu_I = \left[\frac{1}{\tau_{AB}} \int_{\tau_{AB}} A_I B_I dt \right] \frac{1}{\hat{a}_I \hat{b}_I}$$

Substituting the foregoing expressions into (2.51), we see that the local expression for \bar{AB} shows the same form as its spatially homogeneous counterpart,

$$\overline{AB} = \frac{k\lambda\Gamma_{AB}}{\zeta_A\zeta_B} [\bar{A} - (1 - \zeta_A)\bar{A}_I][\bar{B} - (1 - \zeta_B)\bar{B}_I] \quad (2.53)$$

where

$$\zeta_A = \frac{\hat{a}_I \tau_{AB}}{\bar{A}_I \Gamma} \quad \zeta_B = \frac{\hat{b}_I \tau_{AB}}{\bar{B}_I \Gamma} \quad (2.54)$$

$$\lambda = \frac{\mu}{\mu_I} \quad (2.55)$$

Combining (2.51) and (2.53) yields the three-dimensional form of the diffusion zone model,

$$\frac{\partial \bar{A}}{\partial t} + \bar{u}_i \frac{\partial \bar{A}}{\partial x_i} = \frac{\partial}{\partial x_i} \left(K_i \frac{\partial \bar{A}}{\partial x_i} \right) - \frac{k\lambda\Gamma_{AB}}{\zeta_A\zeta_B} [\bar{A} - (1-\zeta_A)\bar{A}_I][\bar{B} - (1-\zeta_B)\bar{B}_I] \quad (2.56)$$

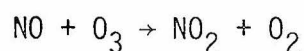
Similar expressions can easily be obtained for \bar{B} .

The parameters ζ_A , ζ_B and λ defined in (2.54-55) bear the similar significances as their one-dimensional counterparts (2.14-15). Intuitively they can be estimated by the same method, even the same formulas, given in Section II.2. We shall not pursue this matter further because the full equation (2.56) is rather difficult to apply. Application of (2.56) to predict concentration decay of reactive species in a three-dimensional turbulent motion can be achieved if the inert concentration statistics are given in a three-dimensional form. Since inert concentration statistics as such are not always available and since K-theory is not generally valid, we consider in the next chapter an alternative method of applying the DZ model to three-dimensional problems.

III. APPLICATION OF THE DIFFUSION ZONE MODEL TO CHEMICALLY REACTING PLUMES

The principal aim of this thesis is to develop means of accounting for the effect of turbulence on the rates of chemical reactions. As we noted in Chapter I, when the characteristic time for chemical reaction is comparable to or shorter than the characteristic time for turbulent mixing, the overall rate of reaction can be expected to be influenced by the turbulent mixing process. In atmospheric chemical reactions, such a situation is most likely to occur immediately after gases have been released into the atmosphere. The usual case is that the gases, once emitted, react rapidly with other gases present in the ambient air. It is possible that the overall rate of conversion in such a case is strongly dependent on the rate of mixing of the effluent and the ambient air.

An important example of this phenomenon is the reaction of nitric oxide and ozone which occurs when NO is emitted into ozone-containing air,

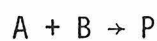


The rate constant for this reaction at 25°C is $29 \text{ ppm}^{-1} \text{ min}^{-1}$ (Johnston and Crosby, 1954). At an O_3 concentration of 0.1 ppm, the time constant for NO decay is 5 seconds. However, observations such as those of Davis et al. (1974) indicate that the time scale for 50% conversion of NO to NO_2 has a value of 12 to 60 minutes in a power plant plume*. Thus, it is suspected that in this instance the rate of turbulent mixing may limit the speed and extent of conversion of NO to NO_2 .

* The conversion of NO to NO_2 in a power plant plume may not, of course, be attributable solely to the reaction of NO and O_3 . All possible reactions must be considered.

Two major examples of pollutant sources in the vicinity of which chemical reactions may be important are highways and power plants. Because the characterization of turbulence on a highway containing vehicles is extremely difficult, we shall confine our attention to chemical reactions occurring in a plume from an elevated point source. We consider in particular a continuous, elevated point source of A in an ambient atmosphere containing B. The objective of the study is to develop means of predicting the conversion of A as a function of downwind distance and turbulence characteristics.

In principle, the steady-state ensemble mean concentrations of two species A and B in a turbulent flow, when A and B react according to



can be determined by solution of the continuity equations.

$$\langle u_k \rangle \frac{\partial \langle A \rangle}{\partial x_k} = - \frac{\partial}{\partial x_k} \langle u_k' A' \rangle - k \langle AB \rangle \quad (3.1a)$$

$$\langle u_k \rangle \frac{\partial \langle B \rangle}{\partial x_k} = - \frac{\partial}{\partial x_k} \langle u_k' B' \rangle - k \langle AB \rangle \quad (3.1b)$$

where

$$A' = A - \langle A \rangle$$

$$B' = B - \langle B \rangle,$$

$$u_k' = u_k - \langle u_k \rangle .$$

The derivation of these equations was given in Chapter I. If appropriate expressions are available for the turbulent flux terms $\langle u_k' A' \rangle$ etc.

and the kinetic term $\langle AB \rangle$, (3.1a) and (3.1b) can be solved to predict $\langle A \rangle$ and $\langle B \rangle$ as functions of location in the plume. As discussed in detail in Chapters I and II, the product moment $\langle AB \rangle$ cannot always be approximated by $\langle A \rangle \langle B \rangle$, and the diffusion-zone model was developed to represent $\langle AB \rangle$. Since the diffusion-zone model requires knowledge of the concentration statistics of inert species, it is necessary to obtain these statistics for the given problem. Consequently, a considerable portion of this chapter is devoted to the generation of inert concentration statistics in atmospheric turbulence.

If the necessary inert concentration statistics are known, $\langle AB \rangle$ can be evaluated by the diffusion-zone model, and (3.1a) and (3.1b) can be solved. However, we generally do not require all the information provided by $\langle A \rangle$ and $\langle B \rangle$ as functions of x_1, x_2 , and x_3 . We are often interested only in the overall level of conversion (integrated over the plume) as a function of distance from the point of release. Furthermore, in this study our concern is with concentration correlation terms like $\langle AB \rangle$ and not with turbulent flux terms. Thus, we seek a form of (3.1a,b) that is free of the flux terms $\langle u_k' A' \rangle$ and $\langle u_k' B' \rangle$. It might be noted that such a formulation for computing overall chemical reaction rates in plumes has not been available heretofore.

III.1 Diffusion Zone Model for a Chemically Reacting Plume

Let us consider a continuous, elevated point source in the atmospheric boundary layer with the x-axis aligned in the direction of the mean wind. If the slender plume approximation can be made, i.e. that turbulent diffusion in the x-direction can be neglected relative to advection by the mean flow, (3.1a) becomes

$$\langle u \rangle \frac{\partial \langle A \rangle}{\partial x} = \frac{\partial}{\partial y} \{ - \langle v'A' \rangle \} + \frac{\partial}{\partial z} \{ - \langle w'A' \rangle \} - k \langle AB \rangle \quad (3.2)$$

where (x_1, x_2, x_3) is replaced by (x, y, z) and (u_1, u_2, u_3) by (u, v, w) . Since the equation for $\langle B \rangle$ is similar, we shall consider this equation only.

We define an arbitrary area A_C in the y-z plane. If z_i is the vertical extent of the atmospheric boundary layer, and L_y is a distance in the y-direction from the plume centerline large enough so that the spread of the plume in the y-direction does not exceed L_y for all values of x of interest, A_C can be defined as the area bounded by $z = 0$ to z_i and $y = -L_y$ to $y = L_y$. We now define the spatial averages of $\langle A \rangle$ and $\langle A^2 \rangle$ over A_C as

$$\tilde{A}(x) = \frac{1}{A_C} \int_{-L_y}^{L_y} \int_0^{z_i} \langle A(x, y, z) \rangle dz dy \quad (3.3)$$

$$\tilde{A}^2(x) = \frac{1}{A_C} \int_{-L_y}^{L_y} \int_0^{z_i} \langle A^2(x, y, z) \rangle dz dy \quad (3.4)$$

The main quantity of interest to us is $\tilde{A}(x)A_c$ which is the mean value of the total amount of A at location x. As we shall see later, this quantity does not depend on the choice of A_c , neither does its rate of change due to chemical reaction.

If there is little wind shear and no surface flux of A or B, (3.2) can be integrated with respect to y and z over A_c to yield

$$\langle u \rangle \frac{d\tilde{A}}{dx} = -k \tilde{AB} \quad (3.5)$$

where

$$\tilde{AB} = \frac{1}{A_c} \int_{-L_y}^{L_y} \int_0^{Z_i} \langle AB \rangle dz dy \quad (3.6)$$

Equation (3.5), together with its counterpart,

$$\langle u \rangle \frac{d\tilde{B}}{dx} = -k \tilde{AB} \quad (3.7)$$

describes the overall rate of conversion as a function of downwind distance in the plume. To solve these equations we must develop a means of computing \tilde{AB} . The diffusion-zone model developed in Chapter II can be modified to represent \tilde{AB} , and we now carry out the modification.

In a stationary process the ensemble average of a variable can be obtained in principle by averaging the instantaneous values of the variable over a sufficiently long period of time. Let this period of time be T, then we can express \tilde{A} as

$$\tilde{A}(x) = \frac{1}{A_c} \int_{L_y}^L y \int_0^{z_i} \frac{1}{T} \int_{t_0}^{t_0+T} A(x,y,z,t) dt dz dy \quad (3.8)$$

provided that $T \rightarrow \infty$. Thus, $\tilde{A}(x)$ can be envisioned as a space-time average carried out at any position x . The time averaging is necessary to produce the ensemble average $\langle A(x,y,z) \rangle$, and then the space averaging is required to yield $\tilde{A}(x)$ from $\langle A \rangle$. At any x we can envision an infinite ensemble of plume cross sections as depicted in Figure 3.1. Let S be the hypothetical volume of space defined by $-L_y \leq y \leq L_y$, $0 \leq z \leq z_i$ and $t_0 \leq t \leq t_0+T$. Then (3.8) can be rewritten as

$$\tilde{A}(x) = \frac{1}{S} \int_S A(S,x) dS \quad (3.9)$$

where $dS = dydzdt$ and $S = A_c T$. At any downwind location x , this hypothetical volume of space S is defined and used for averaging the concentrations. Within this volume, the diffusion zone can be defined as the space where A and B coexist as inerts. Let s_{AB} be the total volume of the diffusion zones, and let s_A and s_B be the total volume of space occupied by A alone and B alone, respectively. Then s_{AB} , s_A and s_B , varying with x , are simply the counterparts to v_{AB} , v_A and v_B of Chapter II. (See Figure 2.1.)

Once we have defined the diffusion zone, the rest of the formulation will be essentially the same as that in Chapter II. Let the average concentration be decomposed into two parts:

$$\tilde{A} = \frac{1}{S} (s_A \hat{A}_I + s_{AB} \hat{a}) \quad (3.10a)$$

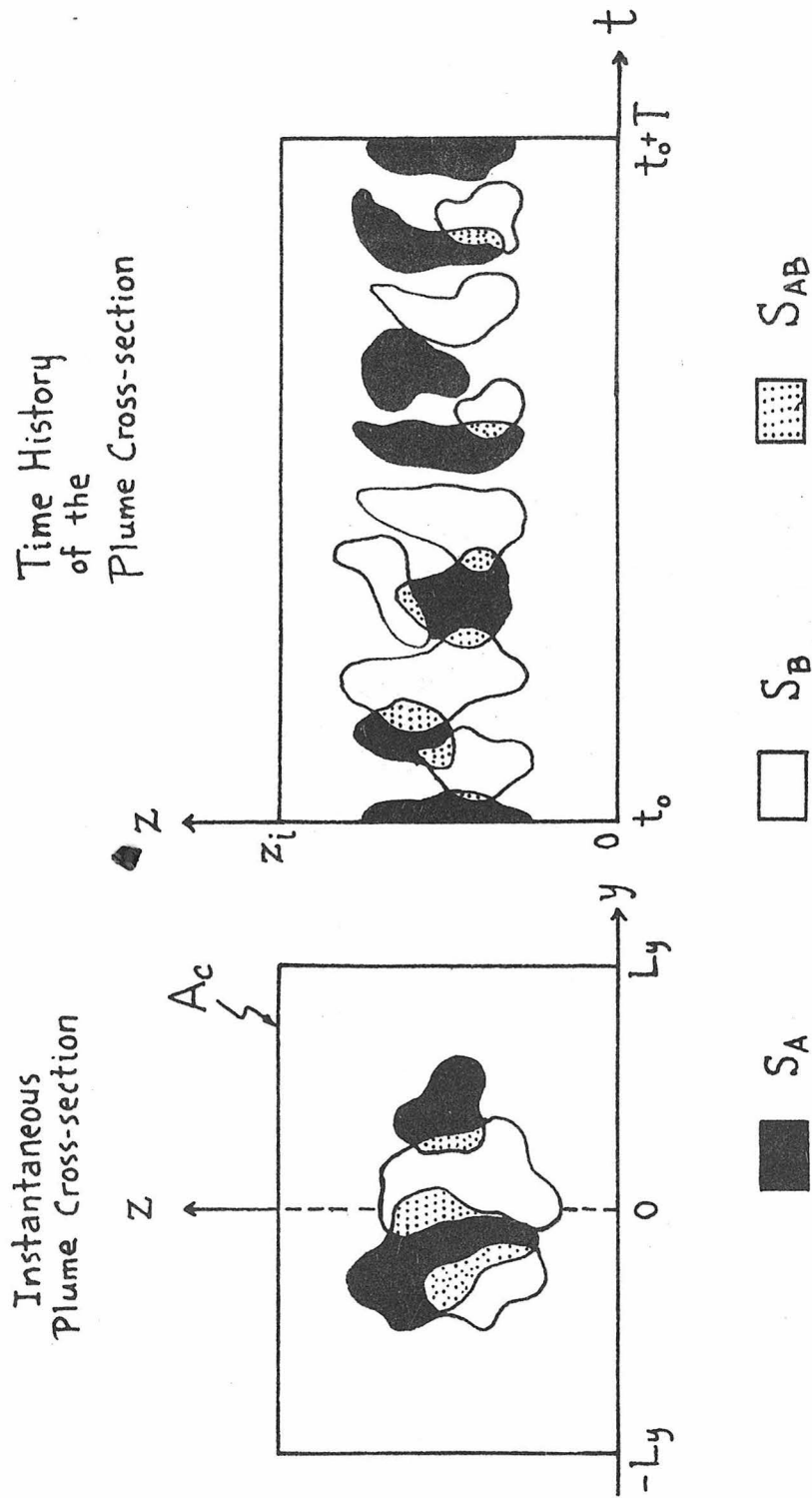


Figure 3.1 Schematic of diffusion zones in the hypothetical space S.

$$\tilde{B} = \frac{1}{S} (s_B \hat{B}_I + s_{AB} \hat{b}) \quad (3.10b)$$

where

$$\hat{A}_I = \frac{1}{s_A} \int_{s_A} A \, ds, \quad \hat{a} = \frac{1}{s_{AB}} \int_{s_{AB}} A \, ds$$

$$\hat{B}_I = \frac{1}{s_B} \int_{s_B} B \, ds, \quad \hat{b} = \frac{1}{s_{AB}} \int_{s_{AB}} B \, ds$$

The concentration product $\tilde{A}\tilde{B}$ is rewritten as

$$\tilde{A}\tilde{B} = \frac{1}{S} \mu \hat{a} \hat{b} s_{AB} \quad (3.11)$$

With (3.9) and (3.11), (3.3) becomes

$$\langle u \rangle \frac{d\tilde{A}}{dx} = -k \frac{\mu s_{AB}}{S} \left(\tilde{A} - \frac{s_A \hat{A}_I}{S} \right) \left(\tilde{B} - \frac{s_B \hat{B}_I}{S} \right) \quad (3.12)$$

Let A_I and B_I be the concentrations of A and B if they were inert,

$$\tilde{A}_I = \frac{1}{S} (s_A \hat{A}_I + s_{AB} \hat{a}_I) \quad (3.13a)$$

$$\tilde{B}_I = \frac{1}{S} (s_B \hat{B}_I + s_{AB} \hat{b}_I) \quad (3.13b)$$

and let

$$\tilde{\Gamma}_{AB} = \frac{\mu_I \hat{a}_I \hat{b}_I s_{AB}}{\tilde{A}_I \tilde{B}_I S} \quad (3.14)$$

where \hat{a}_I , etc are defined similarly to (2.14)-(2.16).

Invoking the hypothesis that the average concentration outside the diffusion zone are the same for both inert and reactive cases, we obtain from (3.12)

$$\langle u \rangle \frac{d\tilde{A}}{dx} = - \frac{k\lambda\tilde{\Gamma}_{AB}}{\zeta_A\zeta_B} (\tilde{A} - (1-\zeta_A)\tilde{A}_I)(\tilde{B} - (1-\zeta_B)\tilde{B}_I) \quad (3.15a)$$

$$\langle u \rangle \frac{d\tilde{B}}{dx} = \frac{k\lambda\tilde{\Gamma}_{AB}}{\zeta_A\zeta_B} (\tilde{A} - (1-\zeta_A)\tilde{A}_I)(\tilde{B} - (1-\zeta_B)\tilde{B}_I) \quad (3.15b)$$

and

$$\lambda = \frac{\mu}{\mu_I} \quad \zeta_A = \frac{\hat{a}_I s_{AB}}{\tilde{A}_I s} \quad \text{and} \quad \zeta_B = \frac{\hat{b}_I s_{AB}}{\tilde{B}_I s}$$

We note that (3.15) is similar in form to (2.13). Although (3.15) and (2.13) are defined in different vector spaces, the physical interpretations of the two equations are basically the same.

Equations (3.15a) and (3.15b) constitute the basic equations for \tilde{A} and \tilde{B} as a function of x in a plume in which A and B are undergoing a second-order reaction. The equations apply to the situation in which the reaction occurs among emitted species in the plume as well as between a species emitted and one in the ambient atmosphere. Subsequently we apply (3.15) to the computation of the conversion of NO in a plume through reaction with O_3 in the ambient air.

To employ (3.15) we must obtain inert concentration statistics. The next section is therefore devoted to the problem of simulating the dispersion of a plume of inert material in the atmospheric boundary layer.

III.2 Simulation of a Plume in the Atmospheric Boundary Layer

The diffusion zone model requires that one have inert concentration statistics for the turbulent field in question. In principle, such data can be obtained experimentally in the field or in the laboratory, or through numerical simulation. Because of the difficulties in obtaining field data on atmospheric diffusion, there is growing interest in direct numerical simulation of turbulent flows (Deardorff, 1970a, Orszag and Patterson, 1972). These simulations provide a "turbulent field" within which diffusion "experiments" may be carried out. Given the turbulent velocity field, diffusion studies can be performed in either a Lagrangian or Eulerian fashion. In the Lagrangian experiments, particles are released from a designated source, and their movements are traced. This technique has been used to study Lagrangian velocity statistics (Deardorff and Peskin, 1970; Riley and Patterson, 1974) as well as the mean concentration from a point source (Lamb et al., 1975). No studies currently exist in which the Eulerian continuity equation has been solved with the fluid velocities derived from a numerical turbulence simulation. In the present work, the Lagrangian particle approach was used in conjunction with the numerical turbulence simulation of Deardorff (1974ab) to generate the statistics required for use of the diffusion-zone model.

III.2.1 Description of the Velocity Field

The velocity field chosen for this study is that generated by Deardorff (1974ab) simulating the wind field on a particular day of the Wangara study conducted in Southeastern Australia. The simulated

turbulence statistics and the temperature profiles have been shown to be in good agreement with the observations of Clarke et al (1971).

Briefly, Deardorff's simulation involves the numerical solution of the three-dimensional equations of motion, mass continuity, heat and moisture balances. The calculations use a grid size that falls within the inertial subrange of the turbulence. The grid-volume-averaged equations are solved using a second-order closure scheme (Deardorff, 1973). Initial conditions for horizontally averaged values of velocities temperature and moisture were taken from DAY 33 of Clarke et al (1971) starting at 9 A.M. This particular day of the Wangara experimental data was chosen because of clear skies, very little horizontal advection of heat or moisture, and lack of any frontal activity within 1000 km. To start up the motions on a resolvable scale, a temperature fluctuation was introduced initially at levels below $z = 200$ m.

The simulation covers an area of $5 \times 5 \text{ km}^2$ with a vertical depth of 2 km. This volume is divided into a $40 \times 40 \times 40$ grid, each cell of which measures $\Delta x = \Delta y = 125$ m and $\Delta z = 50$ m. Due to the vast amount of data generated, only data for the period from 12:29 to 13:21 P.M. local time were stored and available to us. We found this set of wind data favorable in studying our integrated plume model because there is little wind shear in most of the mixed layer and because the turbulence is in a quasi-steady state. The major difficulty in using Deardorff's velocity field to compute second order concentration statistics is the need to parameterize

the effects of subgrid-scale (SGS) velocity fluctuations. All velocity fluctuations of a spatial scale smaller than the grid size, namely 125x125x50 meters, are not resolvable in Deardorff's simulation and this leads to problems in estimating the mean square concentrations in plumes released from a chimney stack.

To be more specific, the mean square concentration of an inert material in a turbulent field is expressed theoretically as (Lamb, 1972a)

$$\langle c^2(x,t) \rangle = \frac{1}{\Delta V^2} \int_{\Delta V} \int_{\Delta V} \int_0^t \int_{-\infty}^{\infty} \int_0^t \int_{-\infty}^{\infty} p(\underline{x}', \underline{x}'', t | \underline{x}'_0, t'_0, \underline{x}''_0, t''_0) \cdot S(\underline{x}'_0, t'_0) S(\underline{x}''_0, t''_0) d\underline{x}'_0 dt'_0 d\underline{x}''_0 dt''_0 d\underline{x}' d\underline{x}'' \quad (3.16)$$

where p is the probability density function that two particles released, one at \underline{x}'_0 at time t'_0 , the other at \underline{x}''_0 at time t''_0 , are found at \underline{x}' and \underline{x}'' , respectively, at time t ; $S(\underline{x}_0, t_0)$ is the source distribution and ΔV is the sample volume centered at \underline{x} . Thus, second-order concentration statistics are determined by the statistics of the displacements of particle pairs. If two particles are released several meters apart, only SGS turbulence can effectively separate them and contribute to two-particle statistics. Therefore, to obtain the mean square concentration of an inert species for the present study of chemical reactions in a plume, it is necessary that a simulation of SGS velocity fluctuations be developed for use with Deardorff's field.

III.2.2 Sub-Grid Scale Velocity Simulation

The most direct solution to the problem of subgrid scale velocity simulation is a reduction in the grid size in the direct calculation. However, the enormous amount of computing time required prohibits this avenue. One must resort to an alternative simulation which produces particle statistics that are as realistic as possible.

The two most important statistical quantities in determining p are the one-particle mean square displacement and the two-particle mean square separation. The former is mainly controlled by energy-containing eddies and hence is determined by the supergrid-scale motions. The latter, however, is controlled by turbulent eddies with sizes comparable to the distance between the two particles. If the particles are initially close to each other, their separation rate is controlled by SGS motions. As the particles are separated farther and farther, large scale motions become more and more important in influencing their rate of separation.

Our criterion for an acceptable SGS velocity simulation is that it gives a mean square separation that agrees with available theory. In the inertial subrange, the rate of separation of two particles can be predicted by similarity theory (Batchelor, 1950):

$$\frac{\langle \ell^2(t) \rangle - \ell_0^2}{\ell_0^2} = \frac{10}{3} c_1 \left(\frac{t \epsilon^{1/3}}{\ell_0^{2/3}} \right)^2 + c_2 \left(\frac{t \epsilon^{1/3}}{\ell_0^{2/3}} \right)^3 \quad (3.17)$$

where $\langle \ell^2(t) \rangle$ is the mean square separation of two particles at time t , or $\langle | \underline{x}' - \underline{x}'' |^2 \rangle$ using the notation of (3.16); ℓ_0 is the initial

separation $|x_0' - x_0''|$; ϵ is the rate of energy dissipation; and c_1, c_2 are constants of the order of unity. It has been confirmed by observation in the atmosphere (Gifford, 1957ab) that (3.17) applies as long as the root-mean-square separation is much smaller than the characteristic length scale of the energy-containing eddies. In Deardorff's calculation, the grid size was chosen so that it falls within the inertial subrange (Deardorff, 1973). Thus, an acceptable SGS velocity parameterization is one which provides a separation rate consistent with (3.17) up to a point where $\langle \ell^2 \rangle^{\frac{1}{2}}$ is comparable to the grid size.

There are two approaches to generate two-particle statistics

- 1) Lagrangian - one assigns random velocities to fluid particles and tracks an ensemble of particles
- 2) Eulerian - one creates an Eulerian velocity field within which particles are imbedded.

Let us consider each approach.

Consider first the Lagrangian approach. Let $u_{p1}(t)$ and $u_{p2}(t)$ be one-dimensional velocities of particles 1 and 2. Then the mean square separation at any time can be written as

$$\begin{aligned}
 \langle \ell^2(t) \rangle &= \langle (x_1 - x_2)^2 \rangle \\
 &= \langle \left(\int_0^t [u_{p1}(t') - u_{p2}(t')] dt' + x_{10} - x_{20} \right)^2 \rangle \\
 &= \ell_0^2 + \int_0^t \int_0^t \langle u_{p1}(t') u_{p1}(t'') \rangle dt' dt'' \\
 &+ \int_0^t \int_0^t \langle u_{p2}(t') u_{p2}(t'') \rangle dt' dt'' - 2 \int_0^t \int_0^t \langle u_{p1}(t') u_{p2}(t'') \rangle dt' dt''
 \end{aligned} \tag{3.18}$$

Obviously, the simulated mean square separation depends entirely on the Lagrangian velocity correlation functions chosen. Unfortunately, except for a few simple cases such as statistically independent particle velocities, a systematic approach to obtain the mean square separation from known correlation functions is lacking. Furthermore, there is no general numerical algorithm from which one can generate Lagrangian random velocities with desired correlations. We therefore turn to the Eulerian approach.

Eulerian methods have been used by a number of previous investigators. Patterson and Corrsin (1966) constructed a binary field with zero autocorrelation to study one-dimensional dispersion. Kraichnan (1970) dealt with two- and three-dimensional homogeneous isotropic velocity field using a random wave simulation (see below). Riley and Corrsin (1971) used a similar method but included a weak mean shear in an otherwise homogeneous flow. Thompson (1971) simulated turbulent diffusion by constructing an Eulerian field with given correlation. Lindberg and Thompson (1974) introduced a method of generating a random field with correct Reynolds' stresses.

The random wave simulation used by Kraichnan (1970) and Riley and Corrsin (1971) is considered here. This method does not simulate the wave number interactions that constitute a fundamental feature of turbulent flows but it does possess the means by which the energy spectrum can be arbitrarily chosen to give desired turbulent statistics. Since in this study we are not as concerned with the exact features of

the turbulence as we are with the Lagrangian statistics, and since the SGS spectrum is well-defined, (see below) the random wave method is chosen to simulate the SGS velocity field.

Let $\tilde{u}(x)$ be the three-dimensional SGS Eulerian velocity.

Following Kraichnan (1970), we have

$$\tilde{u}(x) = \sum_{n=1}^N A(\tilde{\kappa}_n) \cos(\tilde{\kappa}_n \cdot x) + B(\tilde{\kappa}_n) \sin(\tilde{\kappa}_n \cdot x) \quad (3.19)$$

where $\tilde{\kappa}_n = (\kappa_{n1}, \kappa_{n2}, \kappa_{n3})$, and the random amplitudes A and B are calculated by

$$A(\tilde{\kappa}_n) = \zeta_n \times \tilde{\kappa}_n \quad (3.20a)$$

$$B(\tilde{\kappa}_n) = \xi_n \times \tilde{\kappa}_n \quad (3.20b)$$

The vectors ζ_n , ξ_n are independent random variables. Eq. (3.20) is consistent with the continuity equation,

$$\tilde{u} \cdot \tilde{\kappa}_n = 0.$$

(Kraichnan, 1970). The choices of $\tilde{\kappa}_n$, ζ_n and ξ_n are discussed below.

In Deardorff's simulation, the energy spectrum has been truncated at the wave number corresponding approximately to the grid size. Since this wave number was chosen to be in the inertial subrange, the SGS spectrum follows the well-known $\frac{5}{3}$ power law, namely

$$E(\kappa) = \alpha \epsilon^{2/3} \kappa^{-5/3} \quad (3.21)$$

where α is a universal constant. (In Deardorff's calculations α has the value 1.81) The SGS energy is therefore given by

$$E_{SGS} \equiv \frac{1}{2} (\langle u_1^2 \rangle + \langle u_2^2 \rangle + \langle u_3^2 \rangle)$$

$$= \int_{\frac{\pi}{\Delta}}^{\infty} \alpha \epsilon^{2/3} \kappa^{-5/3} d\kappa \quad (3.22)$$

where $\Delta = (\Delta x \Delta y \Delta z)^{1/3}$. The SGS energy E_{SGS} is one of the variables included in Deardorff's data set. With this information and a relationship between ϵ and E_{SGS} given by Deardorff and Drake (1975), the amplitudes of the number vectors κ_n may be chosen.

Thus, let the elements of ζ_n and ξ_n be generated from a Gaussian independent random number generator with zero mean and standard deviation σ_n (to be determined). The components κ_{ni} of the vector κ_n are picked randomly from a spherical surface of radius $\kappa_n = |\kappa_n|$, with the restriction that $|\kappa_{ni}| \geq \frac{\pi}{\Delta}$. This gives

$$\kappa_n^2 = \kappa_{n1}^2 + \kappa_{n2}^2 + \kappa_{n3}^2 \quad (3.23)$$

From (3.19-20), the first component of $u(x)$ is

$$u_1(x) = \sum_{n=1}^N \left\{ (\zeta_{n2} \kappa_{n1} - \zeta_{n3} \kappa_{n2}) \cos(\kappa_n \cdot x) \right. \\ \left. + (\zeta_{n2} \kappa_{n3} - \zeta_{n3} \kappa_{n2}) \sin(\kappa_n \cdot x) \right\}$$

Since ζ_{ni} , ξ_{ni} and κ_{ni} are statistically independent, each with zero mean, we find that

$$\langle u_1^2 \rangle = \sum_{n=1}^N \sigma_n^2 (\langle \kappa_{n2}^2 \rangle + \langle \kappa_{n3}^2 \rangle)$$

Similar expressions can be found for $\langle u_2^2 \rangle$ and $\langle u_3^2 \rangle$. By the definition of E_{SGS} and by (3.23), we see that our simulation implies

$$E_{SGS} = \sum_{n=1}^N \sigma_n^2 \kappa_n^2 . \quad (3.24)$$

In order that (3.24) be equivalent to (3.22), we must have

$$\sum_{n=1}^N \sigma_n^2 \kappa_n^2 = \sum_{i=1}^N \int_{\kappa_n - \frac{1}{2} \Delta \kappa_n}^{\kappa_n + \frac{1}{2} \Delta \kappa_n} \alpha \epsilon^{2/3} \kappa^{-5/3} d\kappa \quad (3.25)$$

where $\kappa_{n+1} - \frac{1}{2} \Delta \kappa_{n+1} = \kappa_n + \frac{1}{2} \Delta \kappa_n$. Integrating (3.25), we have

$$\sigma_n^2 = \frac{3}{2} \alpha \frac{\epsilon^{2/3}}{\kappa_n^2} \left\{ \left(\kappa_n - \frac{1}{2} \Delta \kappa_n \right)^{-2/3} - \left(\kappa_n + \frac{1}{2} \Delta \kappa_n \right)^{-2/3} \right\} \quad (3.26)$$

Using relationship (3.26), random vectors $\underline{\kappa}_n$, $\underline{\zeta}_n$, $\underline{\xi}_n$ can be chosen and the SGS velocity (3.19) can be calculated.

In principle, the number of terms N in (3.19) should be as large as possible. However, in practice since we are interested only in particle pairs with initial separations λ_0 of the order of several meters, a cut-off wave number $\kappa_N \simeq \frac{\pi}{\lambda_0}$ is sufficient. Considering the computing time and storage requirements of this SGS simulation, we have chosen N to be 15.

A trial calculation was carried out using an energy dissipation rate of $0.00195 \text{ m}^2/\text{sec}^3$. Pairs of particles initially located 1 meter apart were advanced by the velocity field (3.19) through the simple integration:

$$\underline{x}(t + \Delta t) = \underline{x}(t) + \Delta t \underline{u}(t).$$

The accuracy of the trajectories produced by this expression is contingent upon $\Delta t u$ being small compared to the grid size (because the latter is a measure of the smallest spatial variations resolvable in Deardorff's velocity field). To meet this requirement, we chose Δt to be 1 second. The particles are followed until $\frac{t \epsilon^{1/3}}{\ell^{2/3}} \approx 600$ when $\langle \ell^2 \rangle$ becomes larger than Δ^2 . Figure 3.2 shows that the simulated mean square separation is in good agreement with Batchelor's theory. The one-dimensional SGS spectrum (x-component) given by (Hinze, 1959)

$$E(\kappa_1) = \frac{2}{\pi} \int_0^{\infty} \langle u(x)u(x+x') \rangle \cos(\kappa_1 x') dx'$$

and the spectrum of the super-grid scale velocity field computed by Deardorff are shown in Figure 3.3. It can be seen that the SGS simulation complements the energy spectrum of Deardorff's velocity field. We therefore conclude that (3.19) is an adequate description of the SGS velocity field.

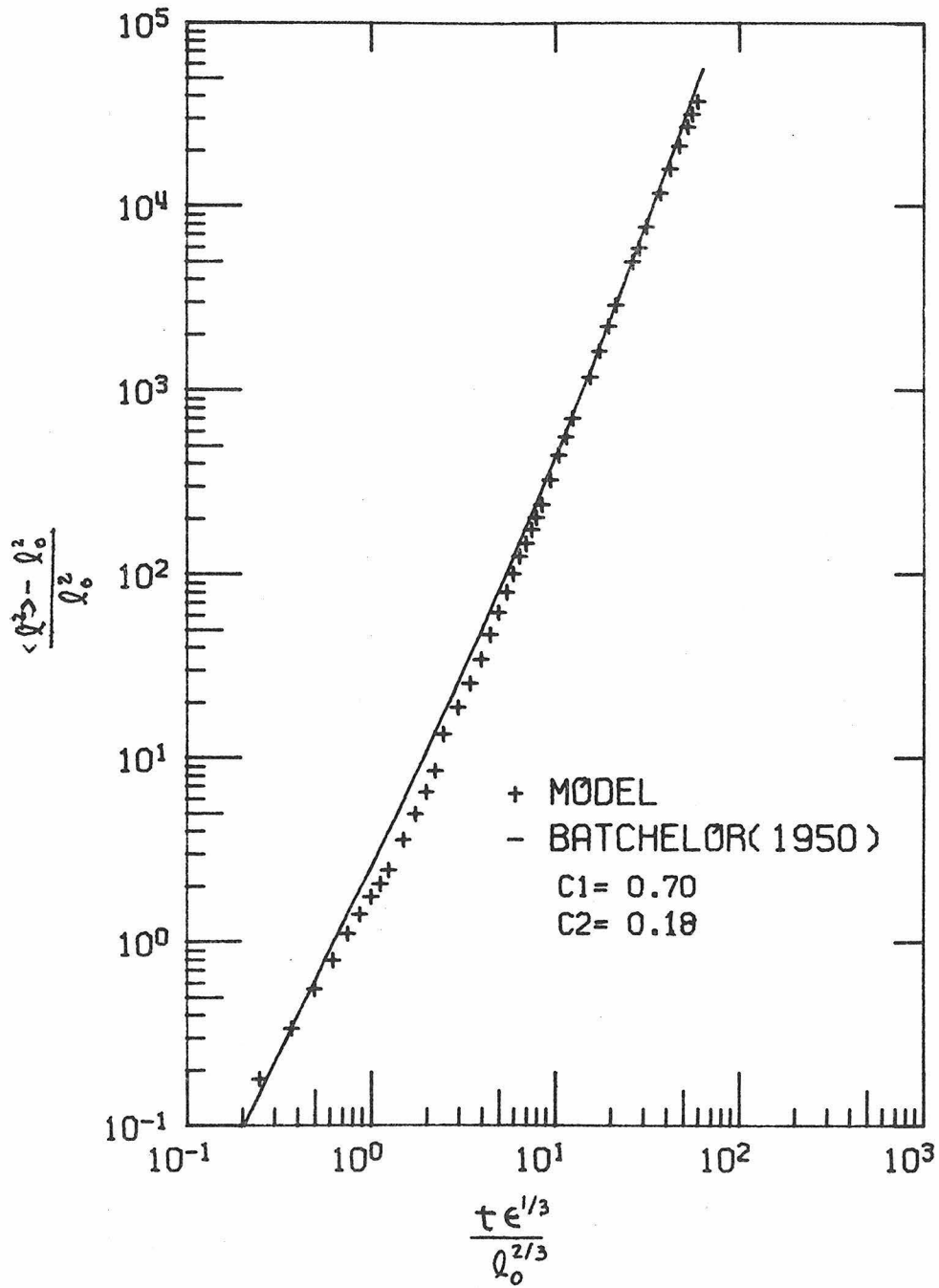


Figure 3.2 Comparison of two-particle mean square separation in a simulated velocity field with Batchelor's prediction.

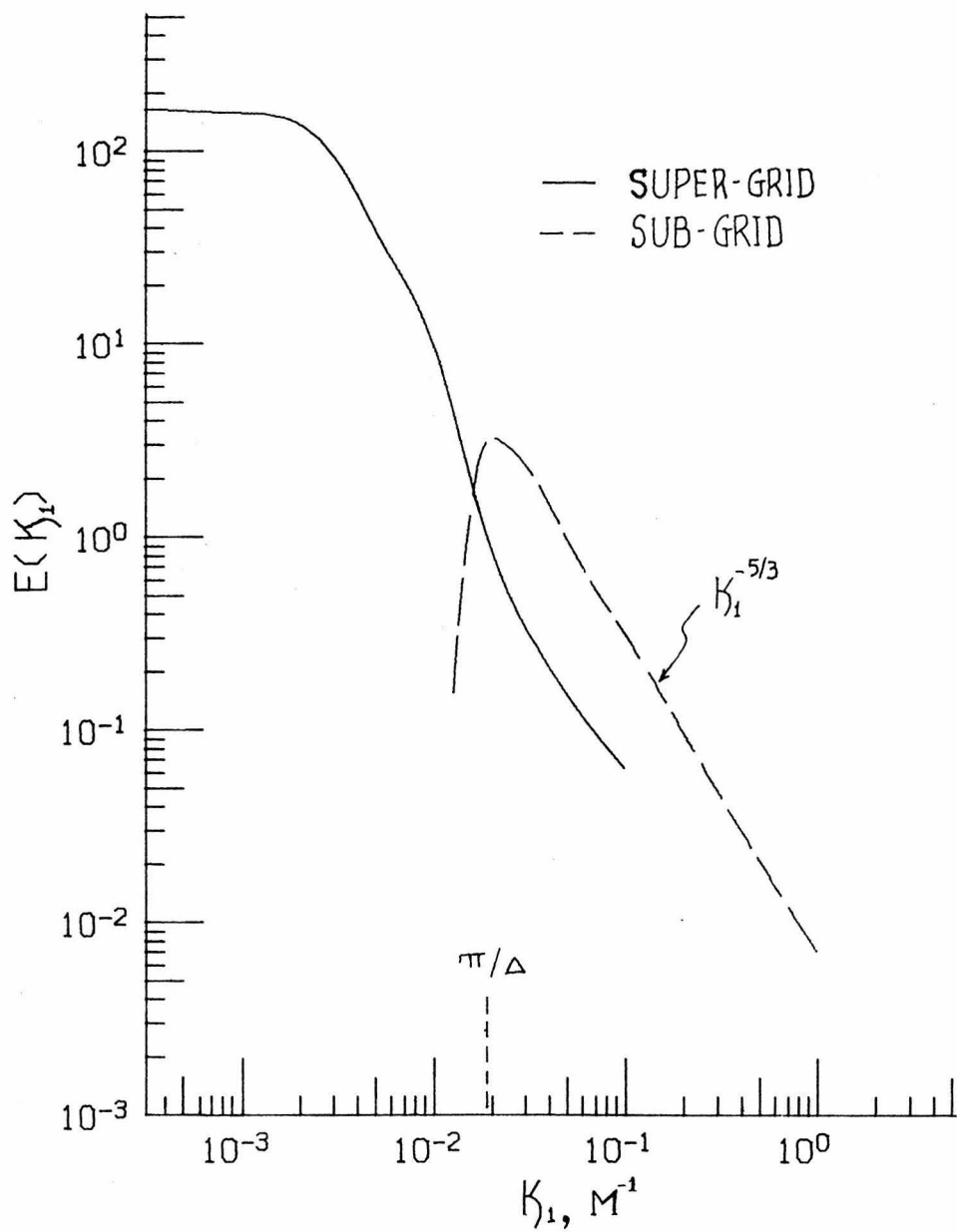


Figure 3.3 One-dimensional energy spectra for super-grid and sub-grid scale velocity fields. The grid scale is denoted by Δ .

III.2.3 Diffusion Experiment in Simulated Atmospheric Turbulence

Having parameterized the SGS velocity field, we can now construct an ensemble of particle trajectories, from which Lagrangian statistics can be computed, by releasing particles into the flow field and tracing their motions. A particle located at \underline{x} has the instantaneous velocity,

$$\underline{u}_p = \underline{u}_D(\underline{x}) + \underline{u}_{SG}(\underline{x})$$

where $\underline{u}_D(\underline{x})$ is the supergrid scale velocity calculated at \underline{x} through linear interpolation of the velocities at neighboring grid-point values and \underline{u}_{SG} is calculated from (3.19). Experiments start at 12:31 local time of the simulated Wangara data. Sixteen hundred pairs of particles, each initially separated by a distance of 4 meters in the y-direction, were released from the center of each grid square on the horizontal plane at $z = 125$ m. The particles were tracked until 12:55 resulting in a total simulation time of 1432 seconds.

In Deardorff's original calculation, the governing equations were solved using a time step of 4 seconds, but velocities were recorded only at alternate time steps. After some tests, it was found that a time step Δt of 8 seconds was sufficiently small, compared to the time scale of variation in the super-grid scale velocity field, that interpolation of the velocities between the recorded time steps was unnecessary. As mentioned earlier, the SGS field requires a much smaller time step. As a compromise to the computing speed we used $\frac{\Delta t}{8}$ for the first 1000 seconds and $\frac{\Delta t}{6}$ for the remainder of the simulation. The trajectory is then computed as follows:

$$\tilde{x}_p(t+\Delta t) = \tilde{x}_p(t) + u_D(\tilde{x}_p(t))\Delta t + \int_t^{t+\Delta t} u_{SG}(\tilde{x}_p(t')) dt$$

The trajectories of the 3200 particles were stored on tape for use in subsequent calculations of the concentration statistics. Some typical particle trajectories are shown in Figure 3.4.

Several checks were performed to test the qualitative features of the particle displacements. First, the mean rise and the mean vertical spread of the particles were calculated for comparison with other data. (Figures 3.5-6) The non-dimensional quantities are defined as follows: (Deardorff, 1972)

$$t^* = tw^*/z_i$$

$$\langle z_p \rangle^* = \langle z_p \rangle / z_i$$

$$\sigma_z^* = \langle (z_p - 125)^2 \rangle^{1/2} / z_i$$

where z_i is the mixed layer depth, z_p is the z-component of particle displacement, and $\langle \rangle$ denotes average over the 1600 pairs. The convective velocity scale w^* is defined by

$$w^* = \left[\frac{g}{\theta_m} \overline{w'\theta'_s} z_i \right]^{1/3} \quad (3.27)$$

where g is the gravitational acceleration, θ_m is the mean potential temperature in the mixed layer, and $\overline{w'\theta'_s}$ is the kinematic heat flux close to the ground. During the simulation period (local time 12:31 to 12:55). We find by interpolating the published values of z_i and w^* (Deardorff, 1974b)

$$z_i = 0.039t + 1134.4 \quad (3.28a)$$

$$w^* = -9.03 \times 10^{-10} t^2 + 1.92 \times 10^{-5} t + 1.957 \quad (3.28b)$$

where t is in seconds, z_i in meters and w^* is in units of m/sec.

Deardorff and Willis (1974) observed both numerically and experimentally that $\langle z_p \rangle^*$ and σ_z^* vary with t^* in a manner resembling that of a damped oscillator. The oscillation is attributed to the dominance of large vertical velocity eddies whose depths are comparable to that of the mixed layer. The numerical result of Deardorff and Willis (1974) is plotted with our data. The agreement of the two appears to be good. (Figure 3.4-5) The difference in the magnitude of overshooting is probably due both to differences in the particle release heights and differences in the aspect ratios of the two experiments. (Aspect ratio is defined the ratio of the horizontal extent of the simulated or measured fluid to the vertical extent. See Willis and Deardorff, 1975.)

As a second check of our particle trajectory calculations, we computed the single particle mean square displacement σ_t^2 and the two particle mean square separation $\langle l^2 \rangle - l_0^2$. (Figure 3.7) Two necessary conditions are (1) $2\sigma_t^2 \geq \langle l^2 \rangle - l_0^2$, and (2) $\langle l^2 \rangle - l_0^2$ should be proportional to t^3 when $\langle l^2 \rangle^{1/2}$ reaches a value comparable to the grid size Δ . It is shown in Figure 3.7 that both conditions are satisfied.

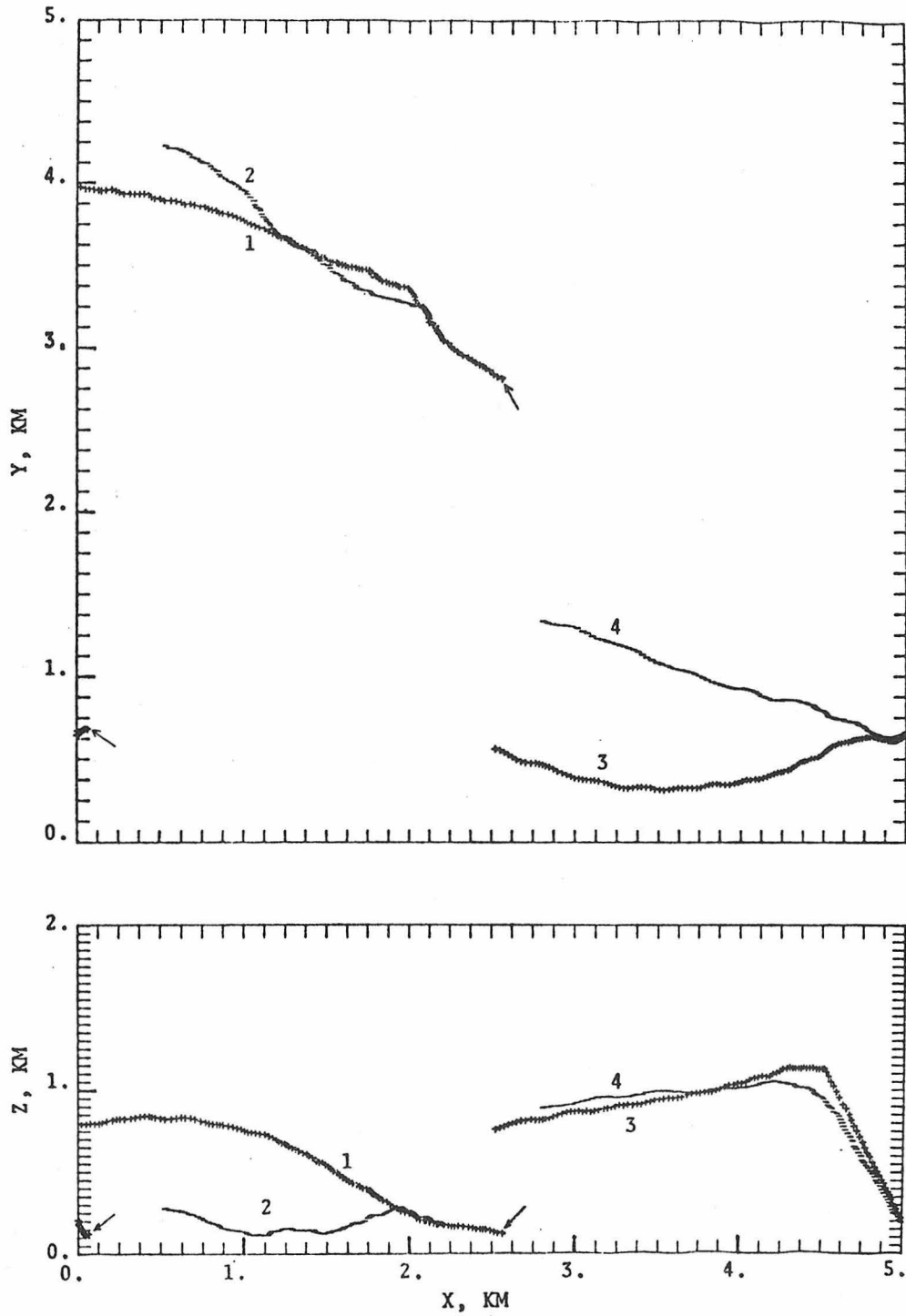


Figure 3.4 Typical particle trajectories as observed in the numerical diffusion experiment. Simulation period: 1000 sec. Arrow indicates starting point. Particles leaving the field are inserted back at the opposite boundary.

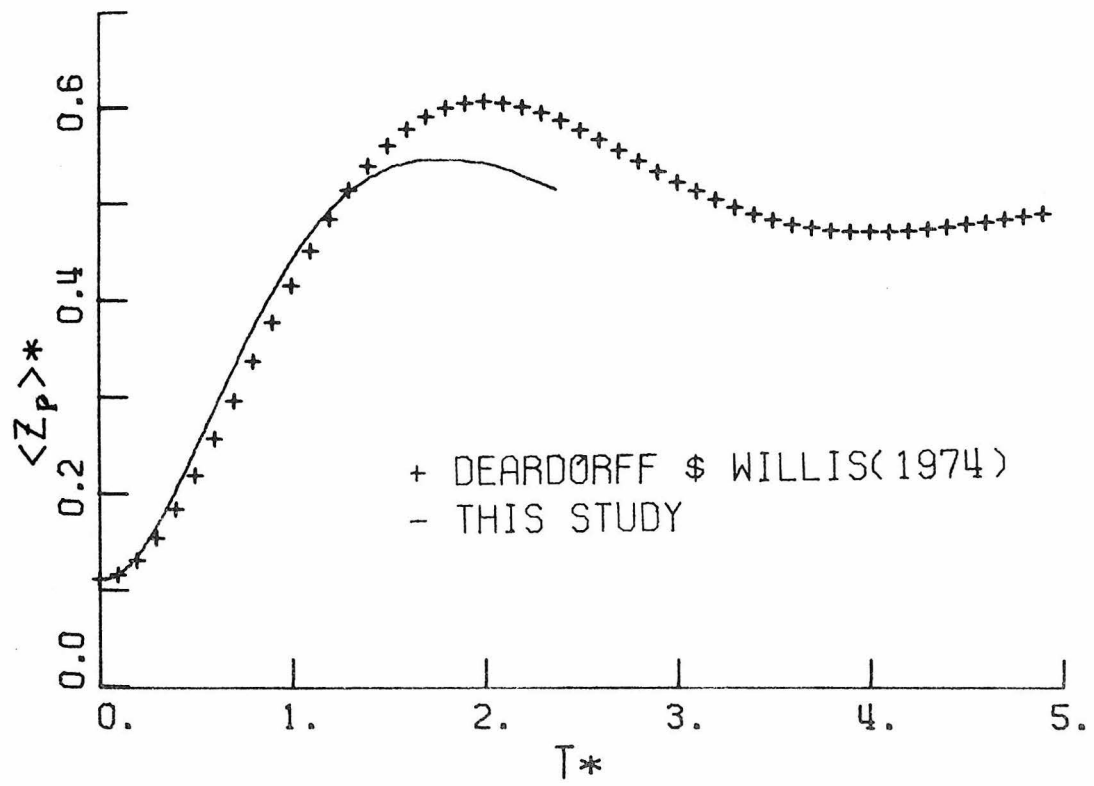


Figure 3.5 Comparison of simulated mean rise of particles with that of Deardorff and Willis(1974).

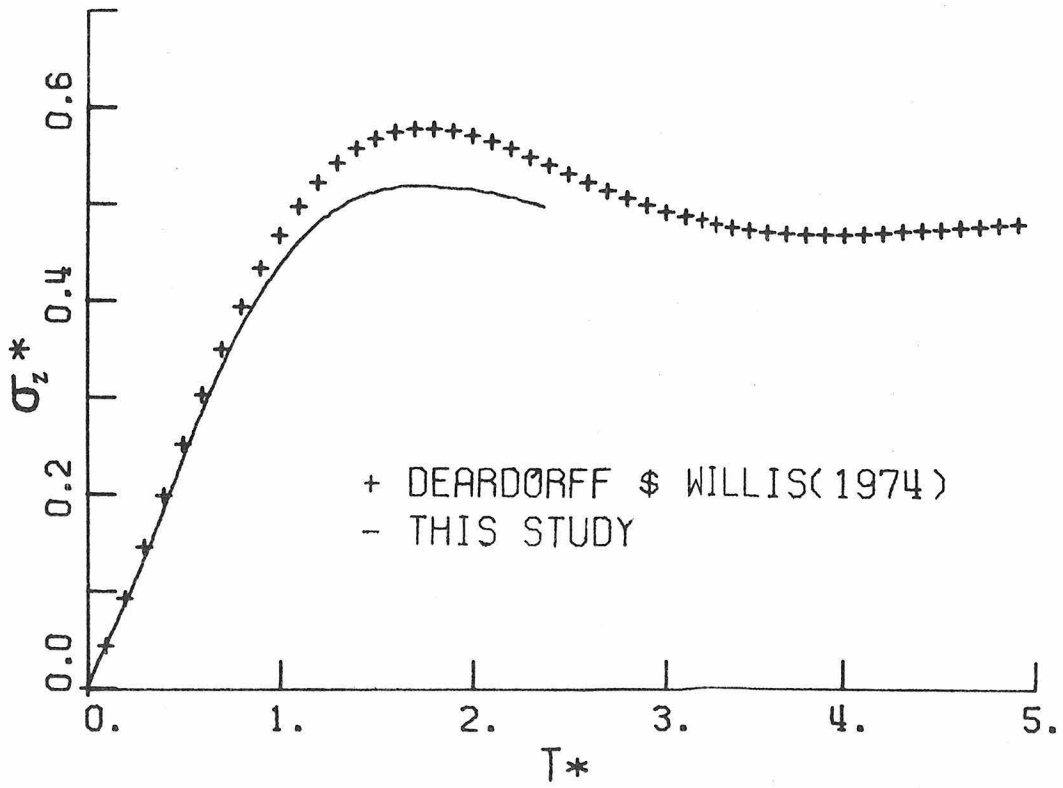


Figure 3.6 Comparison of mean vertical spread of particles with that of Deardorff and Willis(1974).

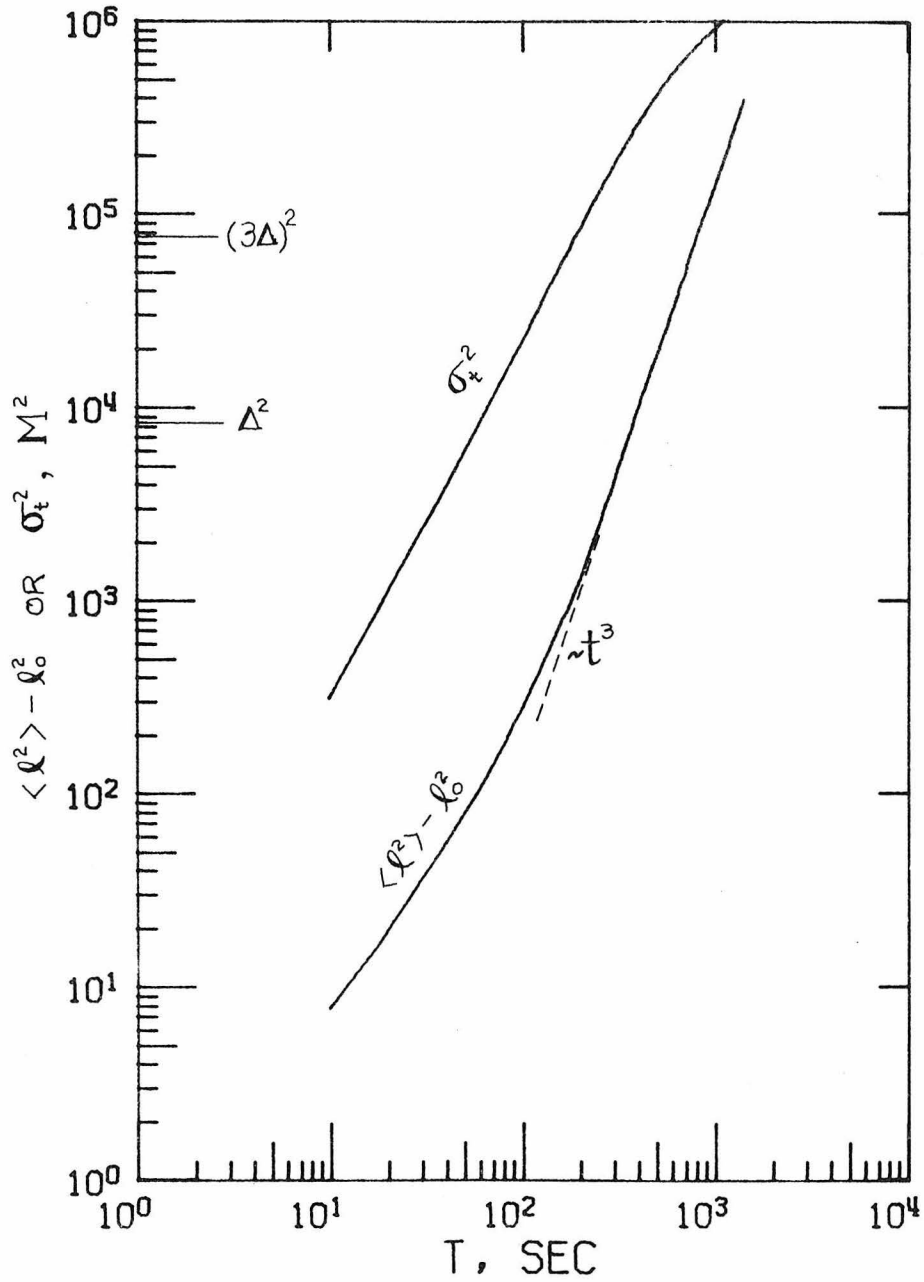


Figure 3.7 One-particle mean square displacement and two-particle mean square separation as observed in the numerical diffusion experiment.

Tests have also been performed by setting the SGS particle velocity to zero and observing the changes that result in the mean square separation. It was found that super-grid scale motions begin to dominate at about 300 seconds after release and contribute more than 98% of the separation for $t \geq 700$ seconds. This result and the calculations shown in Figure 3.6 indicate that the t^3 - region is created partly by the SGS simulation and partly by the supergrid scale velocities.

Some Eulerian and Lagrangian super-grid scale velocity correlation coefficients were also computed and are shown in Figures 3.8-11 for future reference. As expected the Lagrangian autocorrelation coefficient is in general larger than the Eulerian one-point correlation coefficient.

Having established confidence in the accuracy of the simulated particle trajectories, we turn next to the estimation of concentration statistics in a point source plume.

III.3 Concentration Statistics in a Simulated Plume

In applying working equations (3.15a,b) of the DZ model, we need information on the inert concentration statistics. More specifically, we need to know \tilde{A}_I , \tilde{B}_I , $\tilde{\Gamma}_{AB}$, ζ_A and ζ_B . Throughout the rest of this chapter, we shall concentrate on the problem of a plume containing species A emitted into a uniform environment of B. In this case, we shall see in Section III.4 that $\tilde{\Gamma}_{AB} = \zeta_A = 1$ and $\zeta_B = 1/\tilde{\Gamma}_A$, where

$$\tilde{\Gamma}_A = \frac{\tilde{A}_I^2}{\tilde{A}_I^2}$$

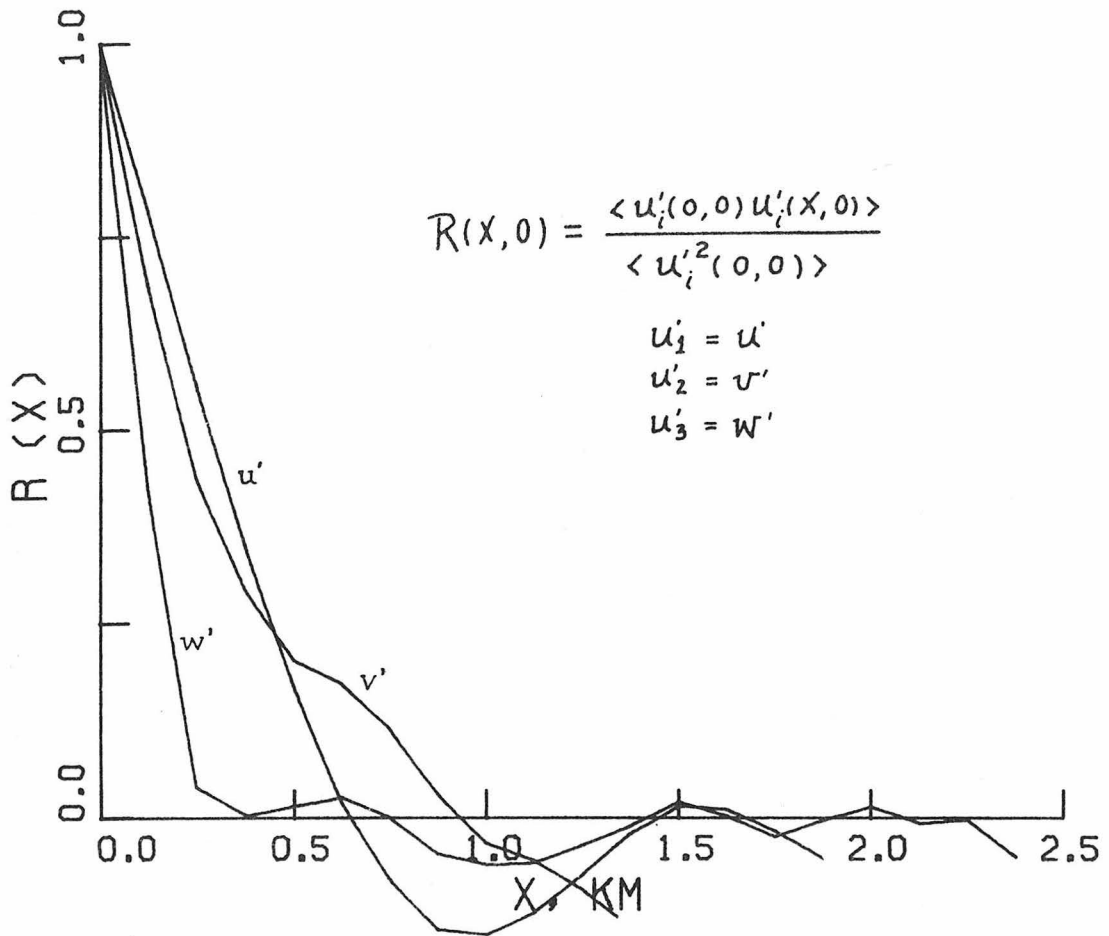


Figure 3.8 Spatial velocity correlations for Deardorff's flow field at 12:31 local time (t=0 in the simulation.)

Note: In Figures 3.8-11, (u',v',w') and (u_p,v_p,w_p) are Eulerian and Lagrangian super-grid scale velocities respectively.

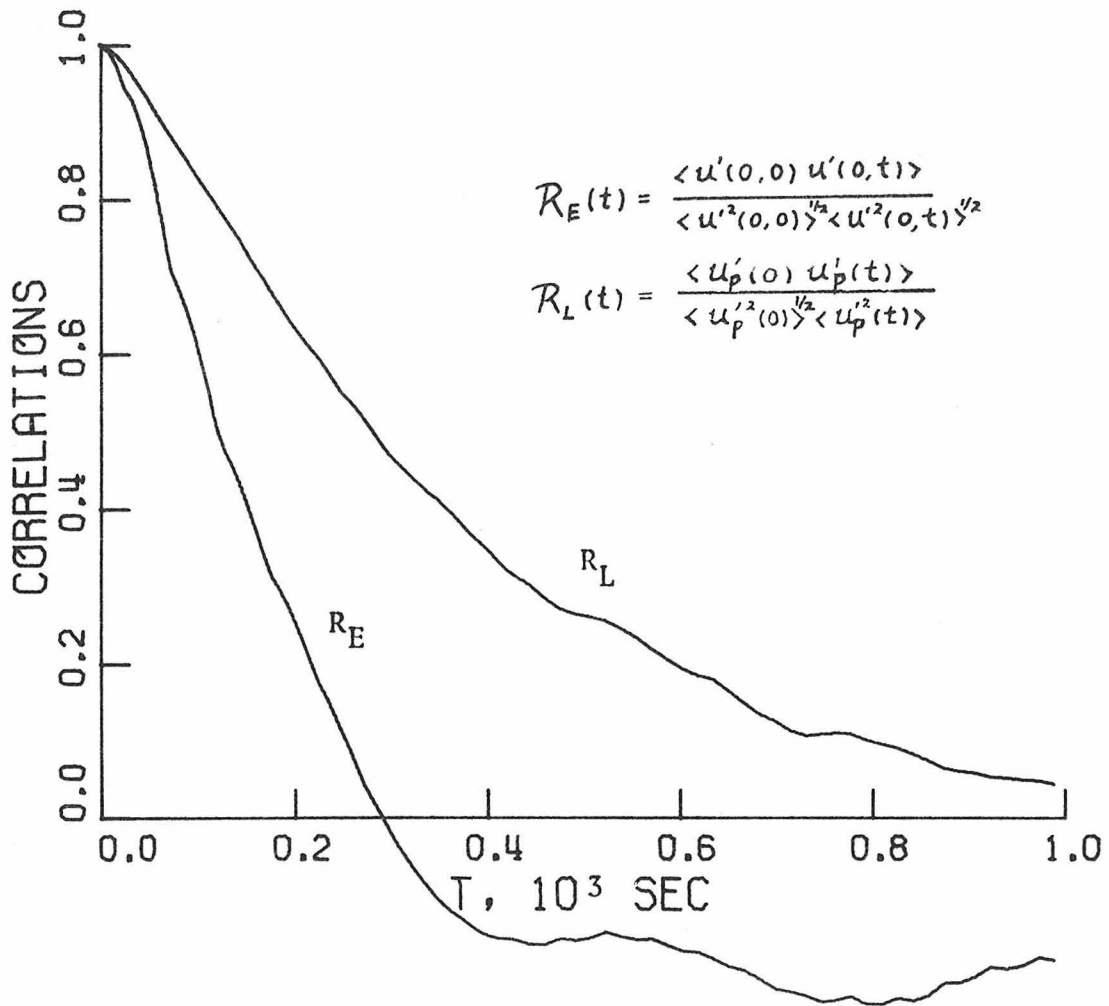


Figure 3.9 Eulerian and Lagrangian velocity correlations of Deardorff's flow field as observed in the diffusion experiment: u' -component.

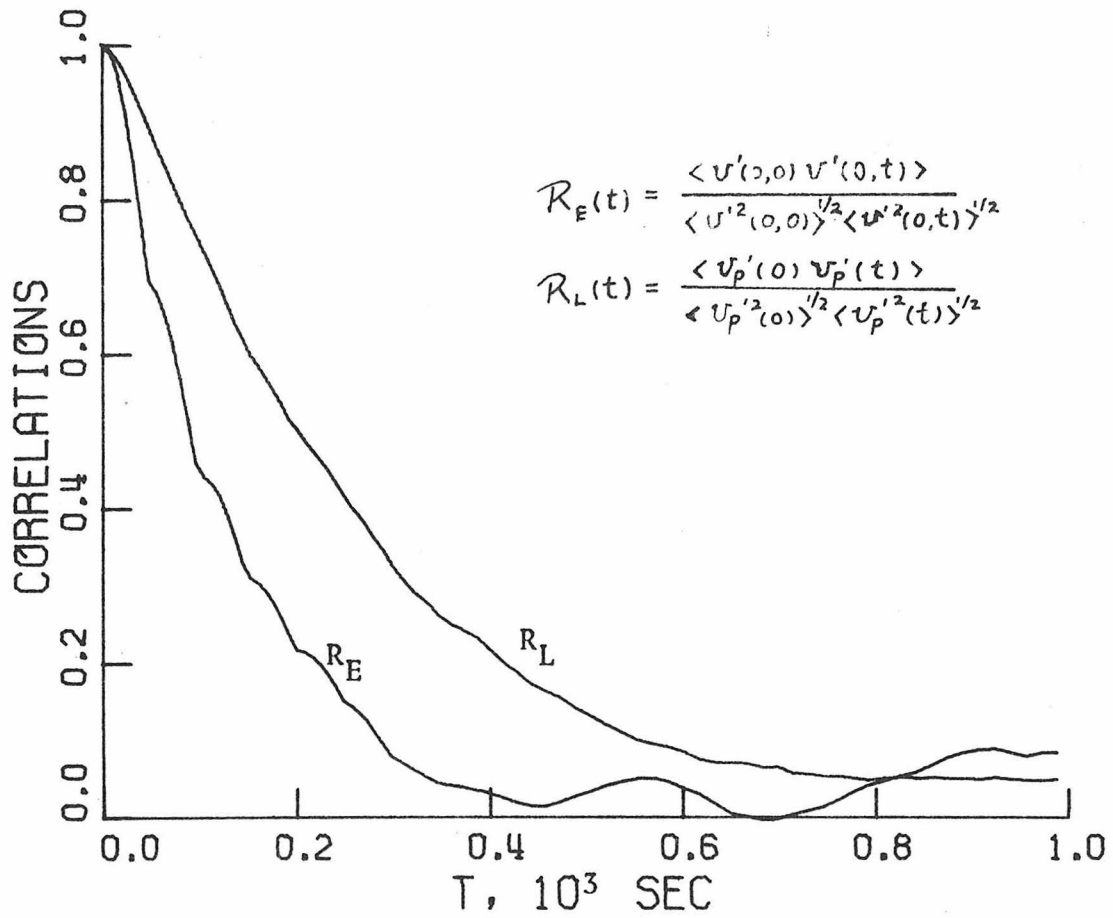


Figure 3.10 Eulerian and Lagrangian velocity correlations of Deardorff's flow field as observed in the diffusion experiment: v' -component.

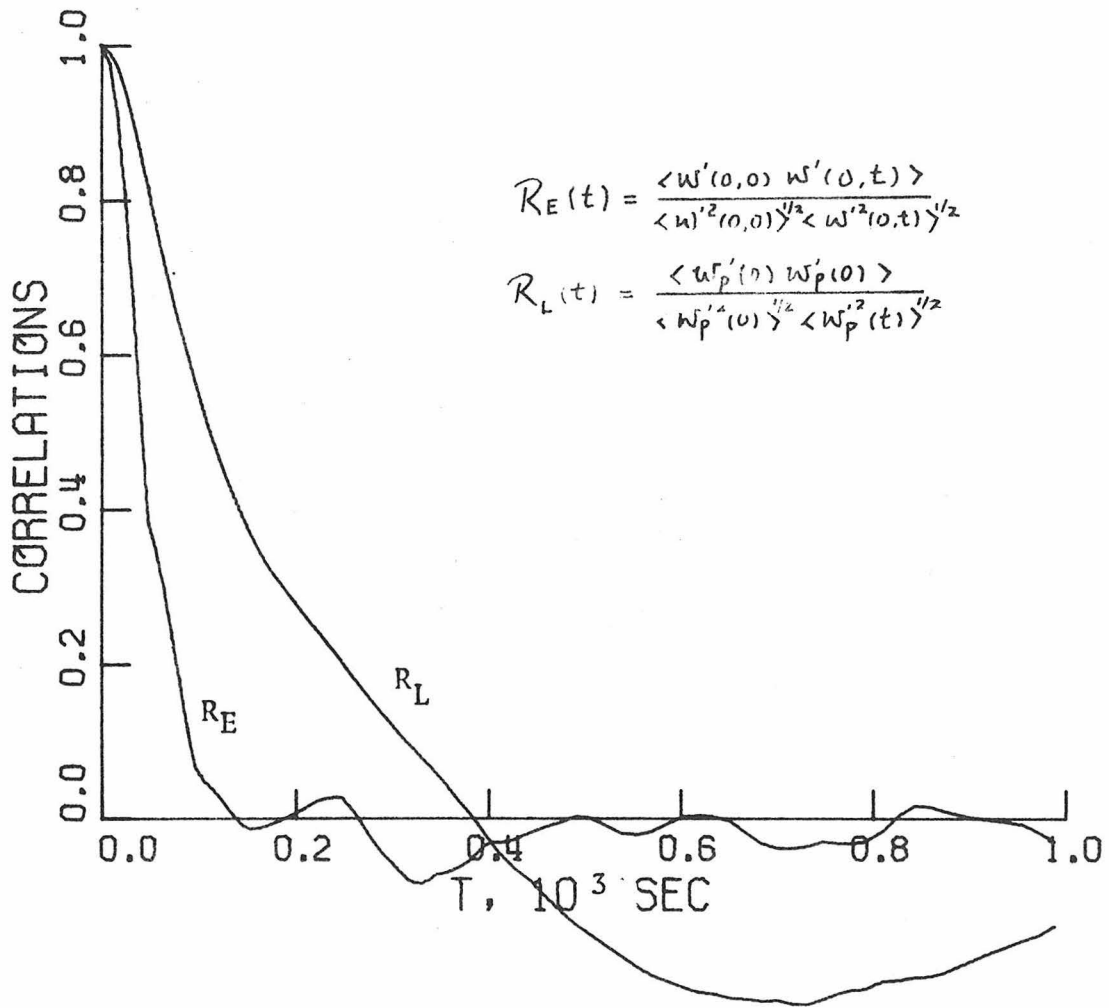


Figure 3.11 Eulerian and Lagrangian velocity correlations of Deardorff's flow field as observed in the diffusion experiment: w' -component.

Therefore the minimum information required to apply the DZ model will be \tilde{A}_I and \tilde{A}_I^2 . By definition,

$$\tilde{A}_I(x) = \frac{1}{A_C} \int_{-L_y}^L y \int_0^{Z_i} \langle A_I(x,y,z) \rangle dz dy \quad (3.29a)$$

$$\tilde{A}_I^2(x) = \frac{1}{A_C} \int_{-L_y}^L y \int_0^{Z_i} \langle A_I^2(x,y,z) \rangle dz dy \quad (3.29b)$$

When x-dispersion is ignored and L_y is sufficiently large, \tilde{A}_I is independent of x but \tilde{A}_I^2 is not.

To evaluate $\tilde{A}_I^2(x)$ one must have the distribution of $\langle A_I^2(x,y,z) \rangle$. The general expression for $\langle A_I^2 \rangle$ is given by (3.16) in terms of a two-particle probability density function (pdf). If the sampling volume ΔV is sufficiently small* that spatial variations of the pdf are negligible within Δv , (3.16) reduces to the form

$$\langle A_I^2(x,t) \rangle = \int_0^t \int_0^t \int_0^t \int_0^t p(\tilde{x}, \tilde{x}, t | \tilde{x}_0, t_0; \tilde{x}'_0, t'_0) S(\tilde{x}_0, t_0) S(\tilde{x}'_0, t'_0) dx_0 dt_0 dx'_0 dt'_0 \quad (3.30)$$

*The sample volume Δv can not be arbitrarily small or sample fluctuations become seriously large (Lamb, 1972a.) It will be shown later that ΔV must be much smaller than $\langle \ell^2 \rangle^{3/2}$.

where $\underline{x}_0 = (x_0, y_0, z_0)$, $d\underline{x}_0 = dx_0 dy_0 dz_0$. Assuming the source is a circular area perpendicular to the x-axis, with area a_s and uniform strength Q_A , we then have

$$\langle A_I^2(\underline{x}, t) \rangle = Q_A^2 \int_0^t \int_0^t \int_{a_s} \int_{a_s} p(\underline{x}, \underline{x}, t | \underline{y}_0, t_0; \underline{y}'_0, t'_0) d\underline{y}_0 d\underline{y}'_0 dt_0 dt'_0 \quad (3.31)$$

where $\underline{y}_0 = (y_0, z_0)$ and $d\underline{y}_0 = dy_0 dz_0$. In stationary turbulence, p is a function of the time differences $t - t_0$ and $t - t'_0$ only, i.e.

$$p(\underline{x}, \underline{x}, t | \underline{y}_0, t_0; \underline{y}'_0, t'_0) = p(\underline{x}, \tau; \underline{x}, \tau' | \underline{y}_0, \underline{y}'_0) \quad (3.32)$$

where $\tau = t - t_0$ and $\tau' = t - t'_0$. If the plume is slender, x-dispersion can be neglected and (3.32) can be approximated by

$$p = \delta(x - \langle u \rangle \tau) \delta(x - \langle u \rangle \tau') p(\underline{y}, \tau; \underline{y}, \tau' | \underline{y}_0, \underline{y}'_0) \quad (3.33)$$

Substituting (3.33) into (3.31), changing variables, and integrating, we obtain

$$\langle A_I^2(\underline{x}) \rangle = Q_A^2 \int_{a_s} \int_{a_s} p(\underline{y}, \underline{y}, t | \underline{y}_0, \underline{y}'_0) d\underline{y}_0 d\underline{y}'_0 \quad (3.34)$$

where $t = x/\langle u \rangle$. Substituting (3.34) into (3.29b), we then find that for a slender plume emitted from an area source with uniform strength into a stationary turbulence

$$\tilde{A}_I^2(t) = \frac{Q_A^2}{\langle u \rangle^2 A_C} \int_{a_s} \int_{a_s} P(t, \underline{y}_0, \underline{y}'_0) d\underline{y}'_0 d\underline{y}_0 \quad (3.35)$$

where

$$P(t, \underline{y}_0, \underline{y}'_0) = \int_{A_C} p(\underline{y}, \underline{y}, t | \underline{y}_0, \underline{y}'_0) d\underline{y} \quad (3.36)$$

Judging from the mean square separation (3.17), we note that the dependence of P on \underline{y}_0 and \underline{y}'_0 vanishes as $t\epsilon^{1/3}/\ell_0^{2/3}$ becomes large. Therefore, in the limit of large t, we can write (3.35) in the form

$$\tilde{A}_I^2(t) = \frac{Q_A^2 a_S^2}{\langle u \rangle^2 A_C} P(t, \underline{\xi}, \underline{\xi}') \quad (3.37)$$

where $\underline{\xi}$ and $\underline{\xi}'$ are any two points in the source area a_S . When t is small, in particular $t\epsilon^{1/3}\ell_0^{-2/3} \leq 5$, P is not independent of \underline{y}_0 and \underline{y}'_0 and the general expression (3.35) must be evaluated numerically. This poses quite a computational burden, however, because the evaluation of P for a single set of initial release points $(\underline{y}_0, \underline{y}'_0)$ requires a large amount of computer time. Therefore, we seek a way of reducing (3.35) to a form like (3.37). This can be achieved if we can find the set of vectors $(\bar{\underline{y}}_0, \bar{\underline{y}}'_0)$ for which

$$P(t, \bar{\underline{y}}_0, \bar{\underline{y}}'_0) = \frac{1}{a_S^2} \int_{a_S} \int_{a_S} P(t, \underline{y}_0, \underline{y}'_0) d\underline{y}'_0 d\underline{y}_0 \quad (3.38)$$

because with this expression we can write (3.35) in the desired form

$$\tilde{A}_I^2(t) = \frac{Q_A^2 a_S^2}{\langle u \rangle^2 A_C} P(t, \bar{\underline{y}}_0, \bar{\underline{y}}'_0) . \quad (3.39)$$

Let us attempt to estimate $(\bar{\underline{y}}_0, \bar{\underline{y}}'_0)$.

III.3.1 Simplification of Equation (3.35).

For an arbitrary function $\xi(x, x_0, x'_0)$, we note that

$$\frac{1}{V^2} \int_V \int_V \xi(x, x_0, x'_0) dx_0 dx'_0 = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \xi(x, \ell, x'_0) g(\ell, x'_0) d\ell dx'_0$$

where $\ell = x - x'_0$, and $g(\ell, x'_0)$ is the joint pdf that any two points chosen at random in the space V will be separated by ℓ and that one point will be located at x'_0 . Applying this relationship to the RHS of (3.38) and letting $\underline{\ell}_0 = \underline{y}_0 - \underline{y}'_0$, we have

$$P(t, \underline{\bar{y}}_0, \underline{\bar{y}}'_0) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} P(t, \underline{\ell}_0, \underline{y}'_0) g(\underline{\ell}_0, \underline{y}'_0) d\underline{\ell}_0 d\underline{y}'_0 \quad (3.40)$$

To proceed we need a description of P but unfortunately little is known about the form of the two-particle density $p(\underline{y}, \underline{y}', t | \underline{y}_0, \underline{y}'_0)$ of which P is a function (see (3.36)). It is known only that in homogeneous stationary turbulence the marginal density $p(\underline{y}, t | \underline{y}_0)$ is Gaussian (Batchelor, 1949). Therefore, following the study of Lamb (1972a), we shall assume that $p(\underline{y}, \underline{y}', t | \underline{y}_0, \underline{y}'_0)$ is jointly normal. (It turns out that the estimate of $(\underline{\bar{y}}_0, \underline{\bar{y}}'_0)$ is only a weak function of the form assumed for p .) We then have for the case of isotropic turbulence

$$p(\underline{y}, \underline{y}', t | \underline{y}_0, \underline{y}'_0) = \frac{1}{(2\pi\sigma^2)^2(1-\rho^2)} \exp \left[- \frac{|\underline{y} - \langle \underline{y} \rangle|^2 + |\underline{y}' - \langle \underline{y}' \rangle|^2 - 2\rho(\underline{y} - \langle \underline{y} \rangle) \cdot (\underline{y}' - \langle \underline{y}' \rangle)}{2\sigma^2(1-\rho^2)} \right] \quad (3.41)$$

where $\sigma^2 = \langle (\underline{y} - \langle \underline{y} \rangle)^2 \rangle = \langle (\underline{z} - \langle \underline{z} \rangle)^2 \rangle$

$$\begin{aligned}\rho &= \langle (y - \langle y \rangle)(y' - \langle y' \rangle) \rangle / \sigma^2 \\ &= \langle (z - \langle z \rangle)(z' - \langle z' \rangle) \rangle / \sigma^2\end{aligned}$$

It is easy to show that

$$\rho = 1 - \frac{\langle \ell^2 \rangle - \ell_0^2}{4\sigma^2} \quad (3.42)$$

where

$$\begin{aligned}\langle \ell^2 \rangle &= \langle (y - y')^2 \rangle + \langle (z - z')^2 \rangle \\ \ell_0^2 &= (y_0 - y'_0)^2 + (z_0 - z'_0)^2\end{aligned}$$

With this assumption, $P(t)$ is obtained by setting $y' = \underline{y}$ and integrating y over A_C . By virtue of the definition of A_C , integration over this domain is equivalent to an integration over an infinite region. Thus we have

$$P(t, \underline{z}_0, \underline{y}'_0) = \frac{1}{(2\sqrt{\pi}\sigma)^2(1-\rho)} \exp \left[- \frac{(\langle y \rangle - \langle y' \rangle)^2 + (\langle z \rangle - \langle z' \rangle)^2}{4\sigma^2(1-\rho)} \right] \quad (3.43)$$

In the diffusion experiment, there is no mean motion in the y -direction. Hence, $\langle y \rangle = y_0$ and $\langle y' \rangle = y'_0$. However, the mean verticle position of each particle does change with time (see Figure 3.5). This mean motion is a weak function of height of release; but since the area of the source is small, the difference in mean rise due to different release heights of particles is negligible. We thus have $\langle z \rangle - \langle z' \rangle = z_0 - z'_0$, and (3.43) becomes

$$P(t, \lambda_o) = \frac{1}{(2\sqrt{\pi}\sigma)^2(1-\rho)} \exp \left[- \frac{\lambda_o^2}{4\sigma^2(1-\rho)} \right] \quad (3.44)$$

where $\lambda_o = | \tilde{\lambda}_o |$. Substituting (3.44) into (3.40) we get

$$P(t, \bar{\lambda}_o) = \int_{-\infty}^{\infty} \frac{1}{4\pi\sigma^2(1-\rho)} \exp \left[- \frac{\lambda_o^2}{4\sigma^2(1-\rho)} \right] g(\lambda_o) d\lambda_o \quad (3.45)$$

where

$$g(\lambda_o) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(\lambda_o, \theta, \tilde{y}'_o) d\tilde{y}'_o d\theta$$

$$\bar{\lambda}_o = | \bar{y}_o - \bar{y}'_o |$$

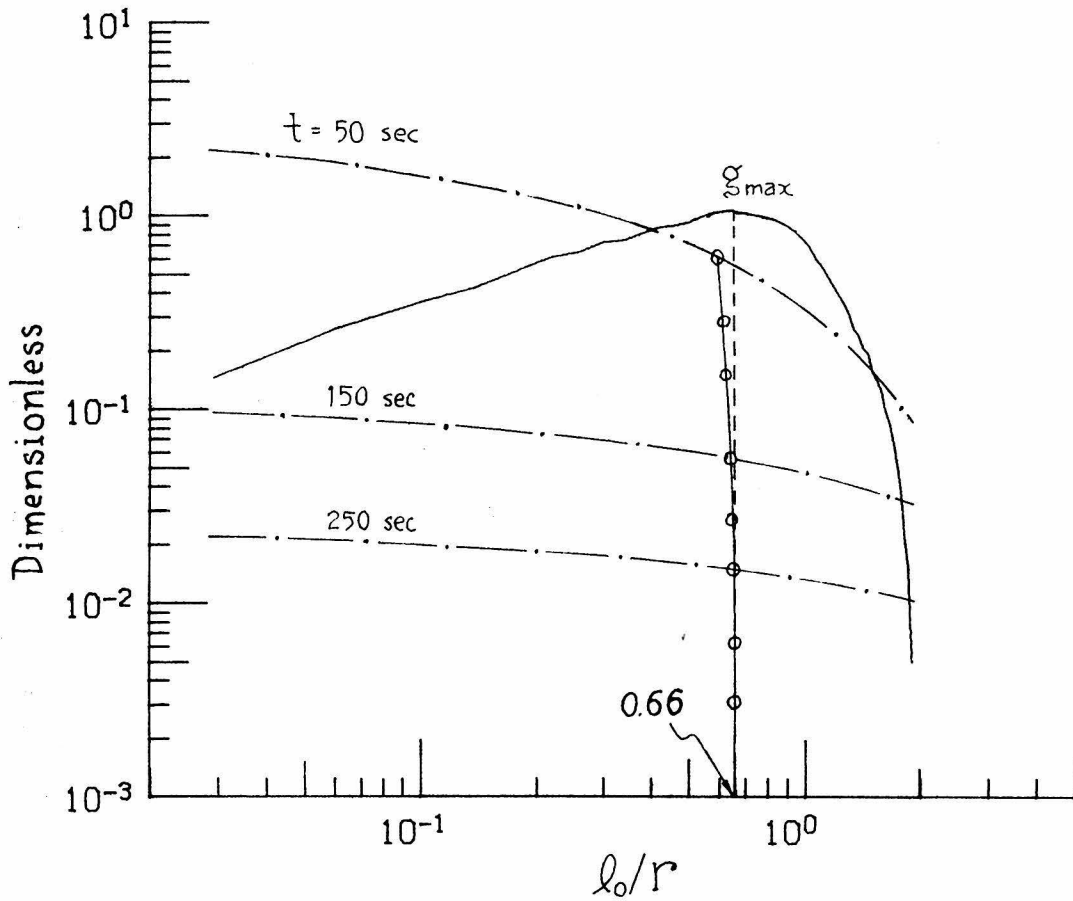
The remaining task is to evaluate the right hand side of (3.45) and to find an initial separation $\bar{\lambda}_o$ such that (3.45) holds.

The function $g(\lambda_o)$ can be obtained numerically by selecting points at random from a circle of radius r and counting the frequency of occurrence of each value of λ_o . The function $rg(\lambda_o/r)$ has a universal form for any value of r and is shown in Figure 3.12. Note that it has a maximum at $\lambda_o/r = 0.66$.

The parameters σ and ρ required in (3.45) are taken from the diffusion experiment data. Fitting a polynomial to each of the curves in Figure 3.7, and applying (3.42), we obtain

$$\sigma^2 = \frac{1}{3} \sigma_t^2 = 0.646t^2 + 5.513t \quad (3.46)$$

$$4\sigma^2(1-\rho) = 0.0047\lambda_o^{2/3}t^2 + 0.825 \times 10^{-4}t^3 \quad (3.47)$$



Legend:
 ——— $rg(l_0/r)$
 -.-.- $a_s P(t, l_0) @ r = 5.56$ m
 ○-○ $\bar{l}_0(t)$
 - - - $l_0/r = 0.66$

Figure 3.12 Calculated distribution functions $g(l_0/r)$ and $P(t, l_0)$. The location of \bar{l}_0 of the mean value of P is strongly associated with g_{max} .

These relationships and the function g allow us to determine $\bar{\ell}_0$ as a function of travel time for various values of r . Figure 3.12 shows a family of curves of $a_s P(t, \ell_0)$ for the case of $r = 5.56$ m. The locus of $a_s P(t, \bar{\ell}_0)$ is indicated by the circled curve in Figure 3.12. It is seen that $\bar{\ell}_0$ moves from $0.59r$ at $t = 50$ sec. to a limiting value $0.66r$ as t gets larger. Clearly the location of $\bar{\ell}_0$ is strongly associated with the behavior of the weighting function $g(\ell_0)$. (Recall that $g(\ell_0)$ has a maximum at $\ell_0 = 0.66r$.) These results suggest that except for small values of t , the function P can be approximated from an ensemble of trajectories of particle pairs of initial separation of $\bar{\ell}_0 = 0.66r$ where r is the radius of the source area. Once $P(t, \bar{\ell}_0)$ is obtained, \tilde{A}_I^2 can be computed from (3.39) with $P(t, \bar{y}_0, \bar{y}'_0)$ replaced by $P(t, \bar{\ell}_0)$.

In summary, we propose the following approximation for $\tilde{A}_I^2(t)$:

$$\tilde{A}_I^2(t) = \frac{Q_A^2 a_s^2}{\langle u \rangle^2 A_c} \int_{A_c} p_2(\underline{y}, t) d\underline{y} \quad (3.48)$$

where $p_2(\underline{y}, t) = p(\underline{y}, \underline{y}, t | \bar{\ell}_0)$ is the pdf that two particles released a distance $\bar{\ell}_0$ apart will be found together in the sample volume centered at \underline{y} at time t . The task now is to measure $p_2(\underline{y}, t)$.

III.3.2 Sampling Procedure and Statistical Error of the

Estimate of \tilde{A}_I^2 .

As noted earlier, the sample volume Δv used to estimate \tilde{A}_I^2 must be sufficiently small that spatial variations of p_2 within Δv are negligible. To estimate the size of Δv , we deduce from the joint-normal

form (3.41) that the length-scale of variation of p_2 is $\sigma(1-\rho)^{\frac{1}{2}}$. By (3.42), we have

$$\sigma(1-\rho)^{\frac{1}{2}} = \frac{1}{2} (1+\rho)^{\frac{1}{2}} (\langle \ell^2 \rangle - \ell_0^2)^{\frac{1}{2}}$$

Since ρ varies from 0 to 1, this length-scale is approximately $(\langle \ell^2 \rangle - \ell_0^2)^{\frac{1}{2}}$, a suitable sample size $\Delta V = \Delta y \Delta z$ is

$$\begin{aligned} \Delta y &= 0.1 \ell_y(t), \\ \Delta z &= 0.1 \ell_z(t), \end{aligned} \tag{3.49}$$

where $\ell_y(t)$, $\ell_z(t)$ are the y- and z-components of $\langle \ell^2 \rangle^{\frac{1}{2}}$ at time t.

(The adequacy of this sample volume size was confirmed through a series of test runs.)

Using this sample volume, we computed the mean square concentration in the following manner:

First, the coordinate system of the recorded particle trajectories was rotated to align the x-axis with the mean wind direction. The mean wind speed is 2.74 m/sec at an angle of 156.88° with respect to the east-west axis.

Next, the coordinates of each pair of particles were translated to a new system (η, ζ) :

$$\eta_1 = y_1(t) - y_1(t_0), \quad \zeta_1 = z_1$$

$$\eta_2 = y_2(t) - y_1(t_0), \quad \zeta_2 = z_2$$

Each pair of particles was then considered to constitute one realization of the ensemble of particle pair trajectories. After the above operations were performed, the sample volume ΔV was determined from (3.49) and the averaging area A_c , defined by (3.3), was divided into cells of size ΔV . Within this framework p_2 may be approximated by

$$p_2(y, t) \cong E\{p_2\} = \frac{1}{N_R} \sum_{i=1}^{N_R} c_i(y, t) \quad (3.50)$$

where N_R is the number of realizations and

$$c_i(y, t) = \begin{cases} \frac{1}{\Delta V^2}, & \text{if both particles of the } i\text{-th} \\ & \text{realization are located in } \Delta V \\ & \text{centered at } y \text{ at time } t; \\ 0 & \text{otherwise.} \end{cases}$$

On substituting (3.50) into (3.39) we obtain

$$\tilde{A}_I^2(t) \cong E\{\tilde{A}_I\} = \frac{Q_A^2 a_S^2}{\langle u \rangle^2 A_C} \int_{A_C} \frac{1}{N_R} \sum_{i=1}^{N_R} c_i(y, t) dy \quad (3.51)$$

That (3.51) provides an unbiased estimate of $\tilde{A}_I^2(t)$ can be seen by taking the ensemble average of this expression and using the definition $p_2 = \langle c_i \rangle$. We find

$$\begin{aligned} \langle E\{\tilde{A}_I^2\} \rangle &= \frac{Q_A^2 a_S^2}{\langle u \rangle^2 A_C} \int_{A_C} \frac{1}{N_R} \sum_{i=1}^{N_R} \langle c_i \rangle dy \\ &= \frac{Q_A^2 a_S^2}{\langle u \rangle^2 A_C} \int_{A_C} p_2 dy = \tilde{A}_I^2 \end{aligned} \quad (3.52)$$

This result insures us that we can make the error in the estimate $E\{\tilde{A}_I^2\}$ of \tilde{A}_I^2 arbitrarily small by making N_R sufficiently large. However, for all practical values of N_R , there is an error associated with $E\{\tilde{A}_I^2\}$.

Let the mean square error in (3.51) be defined by

$$\chi = \langle (E\{\tilde{A}_I^2\} - \tilde{A}_I^2)^2 \rangle$$

Applying (3.53) and (3.44), we obtain

$$\chi = \left(\frac{Q^2 a_s^2}{\langle u \rangle^2 A_c} \right)^2 \int_{A_c} \int_{A_c} \frac{1}{N_R^2} \sum_{i=1}^{N_R} \sum_{j=1}^{N_R} \langle c_i(\underline{y}) c_j(\underline{y}') \rangle d\underline{y} d\underline{y}' - \tilde{A}_I^2{}^2 \quad (3.53)$$

In our experiment, each particle pair released from every grid is considered to be one realization, but these realizations are not statistically independent because of the proximity of the particle pairs to one another at their time of release. Suppose that for each i , c_i is perfectly correlated with m other realizations but is statistically independent of all the rest, i.e.

$$\langle c_i(\underline{y}) c_j(\underline{y}') \rangle = \begin{cases} \langle c_i(\underline{y}) c_i(\underline{y}') \rangle, & \text{if the } i\text{-th and the } \\ & \text{j-th realizations} \\ & \text{are correlated;} \\ \langle c_i(\underline{y}) \rangle \langle c_j(\underline{y}') \rangle, & \text{otherwise.} \end{cases} \quad (3.54)$$

Substituting (3.54) into (3.53), we find that

$$\chi = \frac{m+1}{N_R} \left[\left(\frac{Q^2 a_s^2}{\langle u \rangle^2 A_c} \right)^2 \int_{A_c} \int_{A_c} \langle c_i(\underline{y}) c_i(\underline{y}') \rangle d\underline{y} d\underline{y}' - \tilde{A}_I^2{}^2 \right] \quad (3.55)$$

Recalling the definition of $c_i(y)$, this function can have non-zero value at one location, namely

$$c_i(\underline{y})c_i(\underline{y}') = \begin{cases} c_i^2(\underline{y}), & \text{if } \underline{y}' = \underline{y} ; \\ 0, & \text{if } \underline{y}' \neq \underline{y}. \end{cases}$$

Applying this relationship to (3.55) we have

$$\chi = \frac{m+1}{N_R} \left[\left(\frac{Q^2 a^2}{\langle u \rangle^2 A_C} \right)^2 \Delta V \int_{A_C} \langle c_i^2(\underline{y}) \rangle d\underline{y} - \bar{A}_I^2 \right]$$

The mean fractional error ϵ is then

$$\epsilon \equiv \frac{\chi^{1/2}}{\bar{A}_I^2} = \left(\frac{m+1}{N_R} \right)^{1/2} (\gamma-1)^{1/2} \quad (3.56)$$

where

$$\gamma = \frac{\Delta V \int_{A_C} \langle c_i^2(\underline{y}) \rangle d\underline{y}}{\left[\int_{A_C} \langle c_i(\underline{y}) \rangle d\underline{y} \right]^2} \quad (3.57)$$

To estimate ϵ , we must first approximate γ . To this end we assume that in each realization the particles are equally likely to be found anywhere within a box of size $n\Delta V$. Two extreme cases are considered:

1) If the two particles are totally correlated, they are always observed together and consequently

$$\langle c_{\tilde{i}}(y) \rangle = \begin{cases} \frac{1}{n} \left(\frac{1}{\Delta V} \right)^2, & \text{if } y \in n\Delta V; \\ 0, & \text{otherwise,} \end{cases}$$

$$\langle c_{\tilde{i}}^2(y) \rangle = \begin{cases} \frac{1}{n} \left(\frac{1}{\Delta V} \right)^4, & \text{if } y \in n\Delta V, \\ 0, & \text{otherwise} \end{cases}$$

It follows that in this case $\gamma = 1$ and $\epsilon = 0$.

2) If positions of the two particles are statistically independent, then

$$\langle c_{\tilde{i}}(y) \rangle = \begin{cases} \frac{1}{n^2} \left(\frac{1}{\Delta V} \right)^2 & \text{if } y \in n\Delta V, \\ 0 & \text{otherwise} \end{cases}$$

$$\langle c_{\tilde{i}}^2(y) \rangle = \begin{cases} \frac{1}{n^2} \left(\frac{1}{\Delta V} \right)^4 & \text{if } y \in n\Delta V \\ 0 & \text{otherwise .} \end{cases}$$

and it follows that $\gamma = n$ and

$$\epsilon = \left(\frac{m+1}{N_R} \right)^{\frac{1}{2}} (n-1)^{\frac{1}{2}}$$

We note that the error is larger in the statistically independent case. This is because when the particles are correlated, they are more likely to be found together and fewer realizations are needed to detect all possible positions of the particle pair. For the same number of

realizations, the error should be smaller in the correlated case. In reality, particle pairs are correlated initially and then become independent gradually. Hence the true error is expected to lie somewhere between the two error bounds given above. Let us assume that

$$\epsilon = (1-\rho) \frac{m+1}{N_R} (n-1)^{\frac{1}{2}} \quad (3.58)$$

Based on the spatial velocity correlations in Figures 3.8 - 11, we estimate that initially m has a value of 10 and decays to zero in about 500 seconds. Using $N_R = 1600$, $n\Delta V = 4\sigma_y\sigma_y$ and taking ρ from (3.42), Eq. (3.58) yields the following values of ϵ :

| | | | | | | | |
|----------------|-----|-----|-----|-----|-----|------|------|
| t, sec | 10 | 100 | 200 | 400 | 800 | 1000 | 1400 |
| ϵ , % | 1.1 | 2.3 | 2.6 | 3.2 | 3.4 | 5.0 | 6.4 |

These results indicate that the statistical error is negligible during the period of simulation.

III.3.3 Numerical Results

Let us assume that on N of the N_R realizations the function c_i is non-zero somewhere in A_c . (Recall that by definition c_i can at most have non-zero value at one location.) Eq. (3.51) then reduces to

$$\tilde{A}_I^2(t) = \frac{Q^2 a^2}{\langle u \rangle^2} \frac{A_s \Delta V}{A_c} \frac{N}{N_R} \frac{1}{\Delta V^2} \quad (3.59)$$

Once N has been determined from the numerical experiments, $\tilde{A}_I^2(t)$ can be calculated from (3.59) and the result combined with

$$\tilde{A}_I = \frac{Q_A a_s}{\langle u \rangle A_C} \quad (3.60)$$

to obtain the desired quantity

$$\tilde{\Gamma}_A \equiv \frac{\tilde{A}_I^2}{\tilde{A}_I} = \frac{N}{N_R} \frac{A_C}{\Delta V} \quad (3.61)$$

The computed values of $\tilde{\Gamma}_A$ normalized by its initial value, i.e.

$$\frac{\tilde{\Gamma}_A(t)}{\tilde{\Gamma}_{A_0}} = \frac{a_s}{\Delta V} \frac{N}{N_R} \quad (3.62)$$

is plotted in Figure 3.13 (Note that $\tilde{\Gamma}_{A_0} = A_C/a_s$, which follows from (3.60) and the fact that at $t = 0, \tilde{A}_I^2 = Q_A^2 a_s / \langle u \rangle^2 A_C$.) The figure shows that the measured quantity does approach the correct analytical limits both at $t = 0$ ($\tilde{\Gamma}_A / \tilde{\Gamma}_{A_0} = 1$) and at $t \rightarrow \infty$ ($\tilde{\Gamma}_A \rightarrow 1$). As expected, $\tilde{\Gamma}_A / \tilde{\Gamma}_{A_0}$ decays continually with travel time because the plume is gradually mixing with its environment. We note that $\tilde{\Gamma}_A / \tilde{\Gamma}_{A_0}$ has decayed to one-tenth of its initial value after about five minutes.

Deardorff (1972) found that mean properties of the unstable planetary boundary layer can be made universal if all lengths are scaled by z_i , the height of the inversion below which convection is confined, and times are scaled by z_i/w^* , where w^* is the convective velocity scale defined in (3.27). The travel time t^* nondimensionalized in this manner is given in Fig 3.13. It is found through a least squares fit that

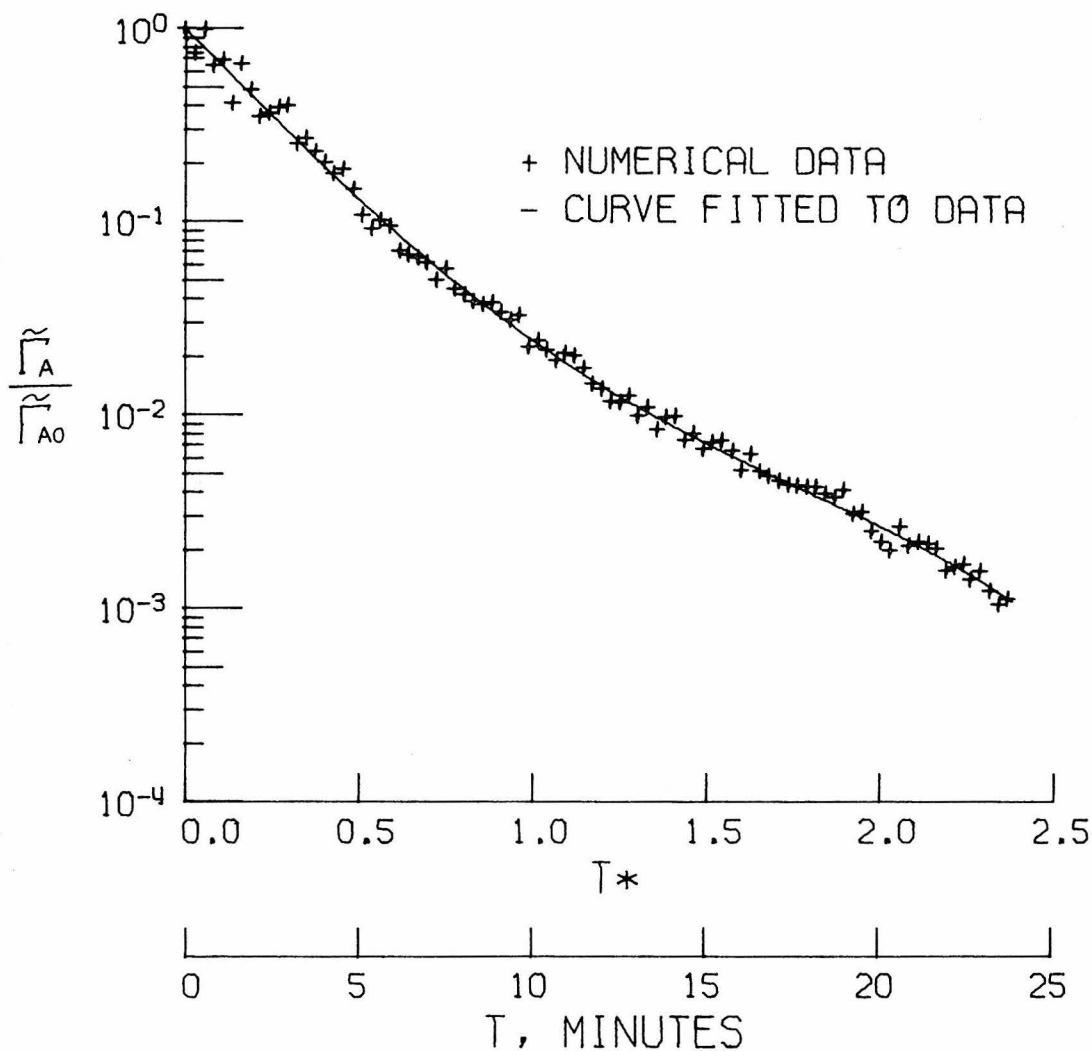


Figure 3.13 Second-order concentration statistics of an inert species in a simulated plume. Release height: 125 m. Source area: circular, radius 5.56 m. Inversion height Z_i : 1100 m. Convective velocity scale: 1.96 m/sec. Stability: $-Z_i/L \sim 911$, where L is the Monin-Obukhov Length.

$$\log_{10} \left[\frac{\tilde{\Gamma}_A}{\tilde{\Gamma}_{A_0}} \right] = a_1 t_* + a_2 t_*^2 + a_3 t_*^3 + a_4 t_*^4 \quad (3.63)$$

where $a_1 = -1.736$, $a_2 = -0.111$, $a_3 = 0.359$, $a_4 = -0.095$. In the next section we utilize these results in the diffusion zone model to simulate a reactive plume in the planetary boundary layer.

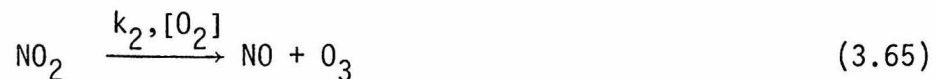
III.4 Study of the NO-O₃ Reaction in Simulated Atmospheric Turbulence

As we noted previously, the two major air pollutant sources for which turbulent chemical reactions may be important are highways and power plants. A pollutant emitted in significant magnitude from each source is NO. Since NO, once emitted, then participates in a complex series of reactions leading to photochemical smog, understanding the fate of NO from these two types of sources is important in understanding photochemical smog chemistry. Hilst et al. (1973) studied a case of NO emitted from a highway source. We concentrate here on NO emissions from an elevated point source in a simulated unstable planetary boundary layer containing a uniform concentration of ozone.

The specifications of the problem that we shall treat were formulated during an early phase of this project when observational data were not available to us. Just recently, however, we discovered measurements by Davis et al. (1974) of NO, O₃, and SO₂ concentrations in plumes produced by the power plant in Morgantown, Maryland. It turns out that the conditions under which these data were measured are sufficiently similar to those of the problem that we simulated that these data provide

a partial check of the validity of the diffusion zone model and our implementation of it. After describing the numerical simulation, we present a comparison of the predicted concentrations with these measured by Davis, et al.

It has been observed, e.g. by Davis et al. (1974), that in the initial position of power plant plumes, the reactions that dominate ozone concentrations are



The value of k_1 at 25°C is 29 ppm⁻¹ min⁻¹ (Johnston and Crosby, 1954). A typical value for k_2 is 0.3 min⁻¹, depending on the intensity of sunlight.

Let $A = [\text{NO}]$, $B = [\text{O}_3]$, and $C = [\text{NO}_2]$, the local instantaneous rate of reaction is

$$\frac{dA}{dt} = \frac{dB}{dt} = -k_1 AB + k_2 C$$

In this case, the diffusion-zone model for the extent of reaction in the plume takes the form

$$\frac{d\tilde{A}}{dt} = \frac{d\tilde{B}}{dt} = - \frac{k_1 \lambda \tilde{\Gamma}_{AB}}{\zeta_A \zeta_B} [\tilde{A} - (1 - \zeta_A) \tilde{A}_I] [\tilde{B} - (1 - \zeta_B) \tilde{B}_I] + k_2 \tilde{C} \quad (3.66)$$

weakly sensitive to the value of μ_a . We therefore assume $\mu_a = 1$ as an approximation. Figure 2.2 indicates that this assumption is more critical at small than at large t . Hence, initially the predicted reaction rate may be underestimated slightly.

The value of the reaction parameter λ was determined from the formula developed in Chapter II (Equation 2.41). According to (2.42), proper value for the parameter α that enters in (2.41) is based on the minimum of the initial concentrations and the maximum of the diffusivities. Since NO is always released in a higher concentration than the background O_3 concentration, we have

$$\alpha = \frac{k_1 B_0 \delta^2}{D}, \quad \lambda = \frac{1}{1+0.16\alpha} \quad (3.69)$$

The background concentration of O_3 assumed in this simulation (and typical of the observations of Davis et al.) is 80 ± 3 ppb. This value gives $\alpha \approx 0.003$ and $\lambda \approx 1$. Reactions with α values of the order of 0.01 to 1 are considered to be of intermediate speed and their rates are influenced by the rate of mixing.

Applying (3.67) and the condition of $\lambda = 1$ to (3.66), we find

$$\frac{d\tilde{A}}{dt} = -k_1 \tilde{\Gamma}_A \tilde{A} \left[\tilde{B} - \left(1 - \frac{1}{\tilde{\Gamma}_A}\right) \tilde{B}_I \right] + k_2 \tilde{C} \quad (3.70)$$

Since (3.64-65) are the only reactions among NO, NO_2 and O_3 , stoichiometry requires that

$$\tilde{A}_I - \tilde{A} = \tilde{B}_I - \tilde{B} = \tilde{C} \quad (3.71)$$

Substituting (3.71) into (3.70), normalizing by the constant \tilde{A}_I , we have

$$\frac{dF_A}{dt} = -k_1(\tilde{A}_I \tilde{\Gamma}_A) F_A \left(F_A - 1 + \frac{B_0}{\tilde{A}_I \tilde{\Gamma}_A} \right) + k_2(1-F_A) \quad (3.72)$$

where $F_A = \tilde{A}/\tilde{A}_I$. Applying the relationship (3.60) and the fact that $\tilde{\Gamma}_{A0} = A_C/a_s$, (3.72) becomes

$$\begin{aligned} \frac{dF_A}{dt} = & -k_1 \left(\frac{Q_A}{\langle u \rangle} \frac{\tilde{\Gamma}_A}{\tilde{\Gamma}_{A0}} \right) F_A \left[F_A - 1 + B_0 \left(\frac{\langle u \rangle \tilde{\Gamma}_{A0}}{Q_A \tilde{\Gamma}_A} \right) \right] \\ & + k_2(1-F_A) \end{aligned} \quad (3.73)$$

Equation (3.73) describes the decay of total amount of NO emitted due to chemical reaction with O_3 . The rate of decay is a function of the kinetic rate constant, background ozone level and turbulent mixing. Consistent with the observations of Davis et al. (1974), this model predicts a higher conversion with a higher background ozone level (larger B_0).

In order to eventually compare the results of this simulation with the plume data of Davis et al., we need to know the emission flux Q_A of NO in the Morgantown plume. Unfortunately, the value of Q_A is not available because the reported data were normalized by the local

NO_x concentrations*. Therefore some estimate has to be made. In Appendix C we estimate the NO -emission of the Morgantown power plant to be between 6500 and 11000 lb/hr at the stacks. Because of the tremendous amount of thermal energy exhausted by the power plant, the plume near the stacks behaves like a thermal jet rather than a passive plume carried by a turbulent fluid. Thus, only the portion of the plume dominated by ambient turbulence can be compared with our simulation. For this reason we treat the data as representing a plume emanating from a fictitious source perpendicular to the plume axis and located at the point downwind where the plume first becomes horizontal. (see Figure 3.14) Suppose that at this point, the plume radius is four times the radius of the stack. Assuming the twin stacks have radii of 5 m each, the average emission flux Q_A at the fictitious source is estimated to be 2.5-4.4 lb/hr.m². Using a mean wind speed of 24 km/hr as observed by Davis et al., we find the fictitious source concentration $Q_A/\langle u \rangle$ to be 39-67 ppm (assuming that each cubic meter of air is 0.089 lb-mole). Estimation of the location of this fictitious source so defined requires more information such as plume temperature etc. than we have. However, Davis and Klauber (1975) reported that at 0.8 km downwind, NO concentration was ~ 2 ppm. This suggests that the fictitious source must be rather near to the stack, and so we shall

*One of the authors indicated to us that the normalization NO_x was not recorded.

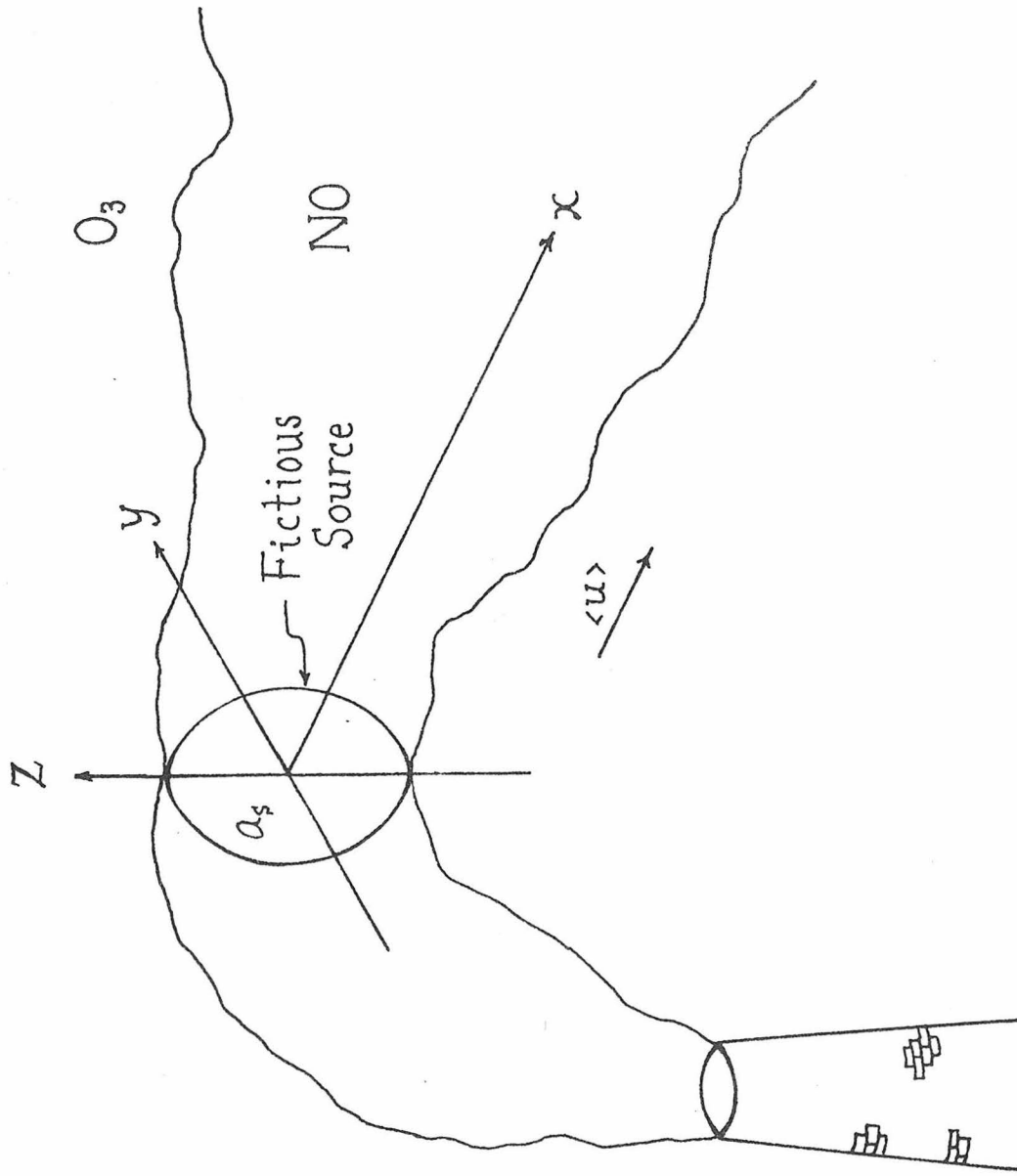
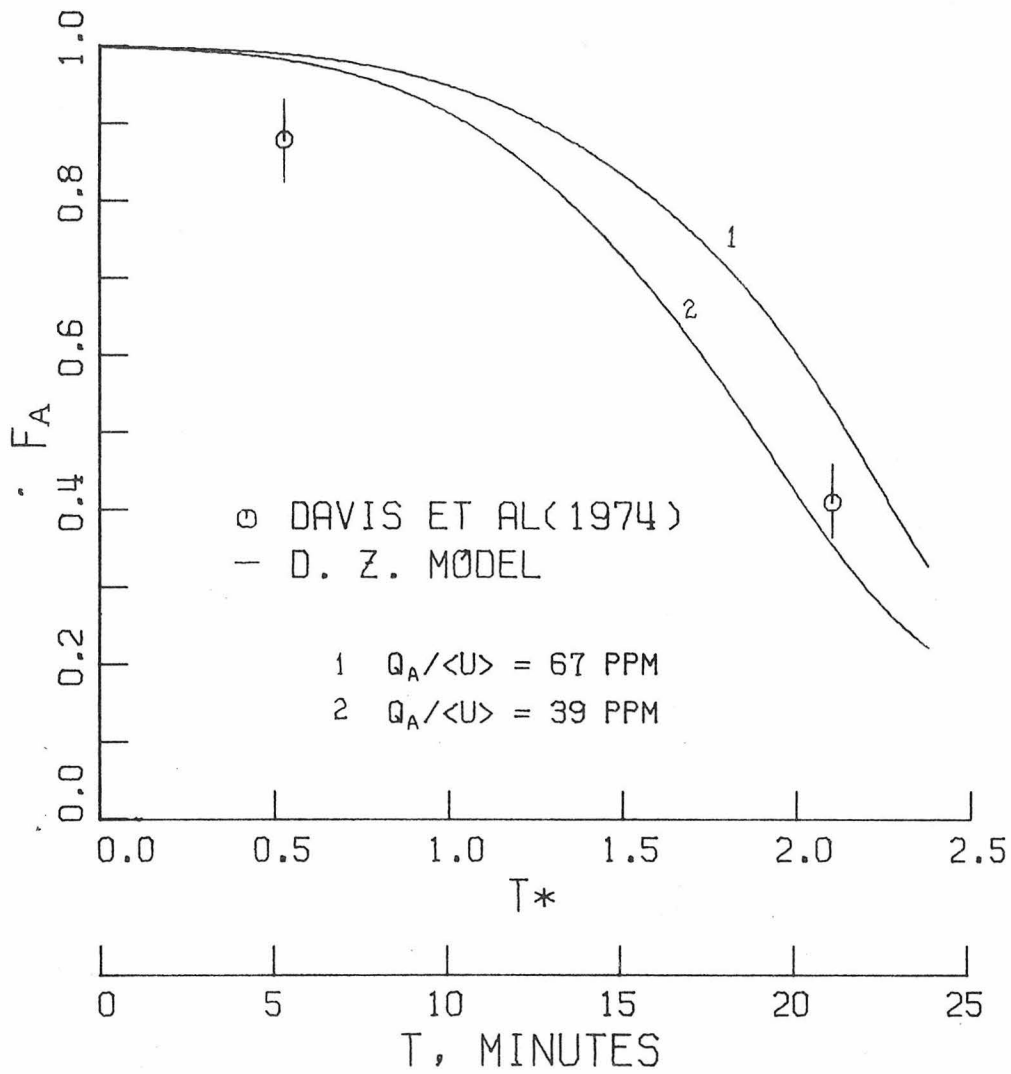


Figure 3.14 Schematic of a power plant plume. The fictitious source is located at the point downwind where the plume first becomes horizontal.

assume that no appreciable chemical conversion has taken place before NO reaches the fictitious source location. Admittedly, these estimates are rather crude but without them we have no basis for judging the accuracy of our simulation.

Using these estimates and (3.63), we solved (3.73) for F_A with the initial condition $F_A(0) = 1$. The solution F_A is plotted in Figure 3.15 as a function of $t^* = tw_*/z_i$, where w_* and z_i are given in (3.28). The data of Davis et al. (1974) are shown on the same graph for comparison. To connect these data with the t_* coordinate, we first transformed the distance-scale to real time-scale by using a uniform wind speed of 24 km/hr. Then we normalized t using $z_i = 1000$ m, supplied by Davis (private communication), and $w_* = 1.8$ m/sec estimated from the observed cloud cover, wind speed, and approximate surface roughness of the Morgantown area. The error bar on each of the data points indicates the range of observed values.

Due to the uncertainty in the estimate of NO-emission rate, we computed F_A for both the maximum and the minimum of the estimated $Q_A/\langle u \rangle$ values. Although there are too few data points to make a definite judgement, the computed and observed conversion rates are in good agreement. The fact that the observed conversion is greater at small travel times than that predicted by the model suggests that the buoyancy generated turbulence in the initial portion of the plume is more effective in mixing the NO and ambient ozone than we had assumed. Recall that since buoyancy effects are not included in our simulation, we treated



$B_0 = 0.08$ PPM
 $\langle U \rangle = 2.74$ M/SEC

Figure 3.15 Reactive decay of the total mass of NO emitted from a power plant stack into an ozone laden atmosphere. Solid curves are the DZ model prediction, while circles are observations.

the observational data as though the observed plume were neutrally buoyant and issuing from a fictitious source some distance from the stack. Part of the underestimate of the computed conversion may also be due to our assumptions that μ_a in (3.68) has unit value. There may also be some effect of the differences in the release heights of the simulated and observed plumes.

An important aspect of the computed and observed data that are in close agreement is the magnitude of the time scale of conversion of NO. As is evident in Figure 3.15, in both the simulation and the observations 50% of the NO released has been converted to the NO₂ during about the first 17 minutes of travel. Davis et al. (1974) reported that the conversion times they observed ranged from about 12 to 60 minutes depending on the meteorology and background ozone concentration.

In view of the rather large computational effort that is necessary to derive the parameters (such as $\tilde{\Gamma}_A$) required to implement the diffusion zone model, it is appropriate to inquire whether a plume simulation of comparable accuracy could be achieved using a simple, conventional approach. To investigate this matter, we constructed a Gaussian type model of the NO plume treated above and exercised it under the same conditions as those implicit in the data shown in Figure 3.15. The model equation is (cf.(3.2))

$$\langle u \rangle \frac{\partial \langle A \rangle}{\partial x} = \langle u \rangle \frac{\partial \langle B \rangle}{\partial x} = \frac{\partial}{\partial y} (K_y \frac{\partial \langle A \rangle}{\partial y}) + \frac{\partial}{\partial z} (K_z \frac{\partial \langle A \rangle}{\partial z}) - k_1 \langle AB \rangle + k_2 \langle C \rangle \quad (3.74)$$

where K_y and K_z are turbulent eddy diffusivities. For simplicity we assume $K_y = K_z = K$ and is a function of time only. If it is also assumed that the plume is axisymmetric, then (3.74) becomes

$$\frac{\partial \langle A \rangle}{\partial t} = \frac{\partial \langle B \rangle}{\partial t} = K(t) \left(\frac{\partial^2 \langle A \rangle}{\partial r^2} + \frac{1}{r} \frac{\partial \langle A \rangle}{\partial r} \right) - k_1 \langle AB \rangle + k_2 \langle C \rangle \quad (3.75)$$

where $t = x/\langle u \rangle$ and $r = \sqrt{y^2 + z^2}$. We assume that $K(t)$ is related to the mean square displacement of a single particle by

$$K(t) = \frac{d\sigma^2}{dt} \quad (3.76)$$

and use $\sigma^2(t)$ as given in (3.46). These assumptions are not critical, because since the conversion F_A is defined by

$$F_A(t) = \frac{\int_0^\infty 2\pi r \langle A(r,t) \rangle dr}{a_s \frac{Q_A}{\langle u \rangle}}, \quad (3.77)$$

the effect of cross-sectional diffusion is reduced. Test runs indicate a weak dependence of F_A on the eddy diffusivity $K(t)$.

The nonlinear kinetic term in (3.75) can be decomposed into two parts:

$$\langle AB \rangle = \langle A \rangle \langle B \rangle + \langle A'B' \rangle \quad (3.78)$$

The fluctuation term $\langle A'B \rangle$ represents the degree of segregation of the reactants, and for fast reactions it limits the rate of reaction. It was for the purpose of describing concentration fluctuation effects that the diffusion-zone model was developed. In keeping with the present practice of neglecting fluctuation terms, we shall omit $\langle A'B \rangle$ from the Gaussian plume model equation. With this assumption and the stoichiometry condition $\langle B \rangle + \langle C \rangle = B_0$, we obtain

$$\frac{\partial \langle A \rangle}{\partial t} = \frac{\partial \langle B \rangle}{\partial t} = K(t) \left(\frac{\partial^2 \langle A \rangle}{\partial r^2} + \frac{1}{r} \frac{\partial \langle A \rangle}{\partial r} \right) - k_1 \langle A \rangle \langle B \rangle + k_2 (B_0 - \langle B \rangle). \quad (3.79)$$

The initial conditions are

$$\langle A(r,0) \rangle = \begin{cases} Q_A / \langle u \rangle, & \text{if } 0 \leq r \leq r_0 \\ 0, & \text{otherwise} \end{cases} \quad (3.80a)$$

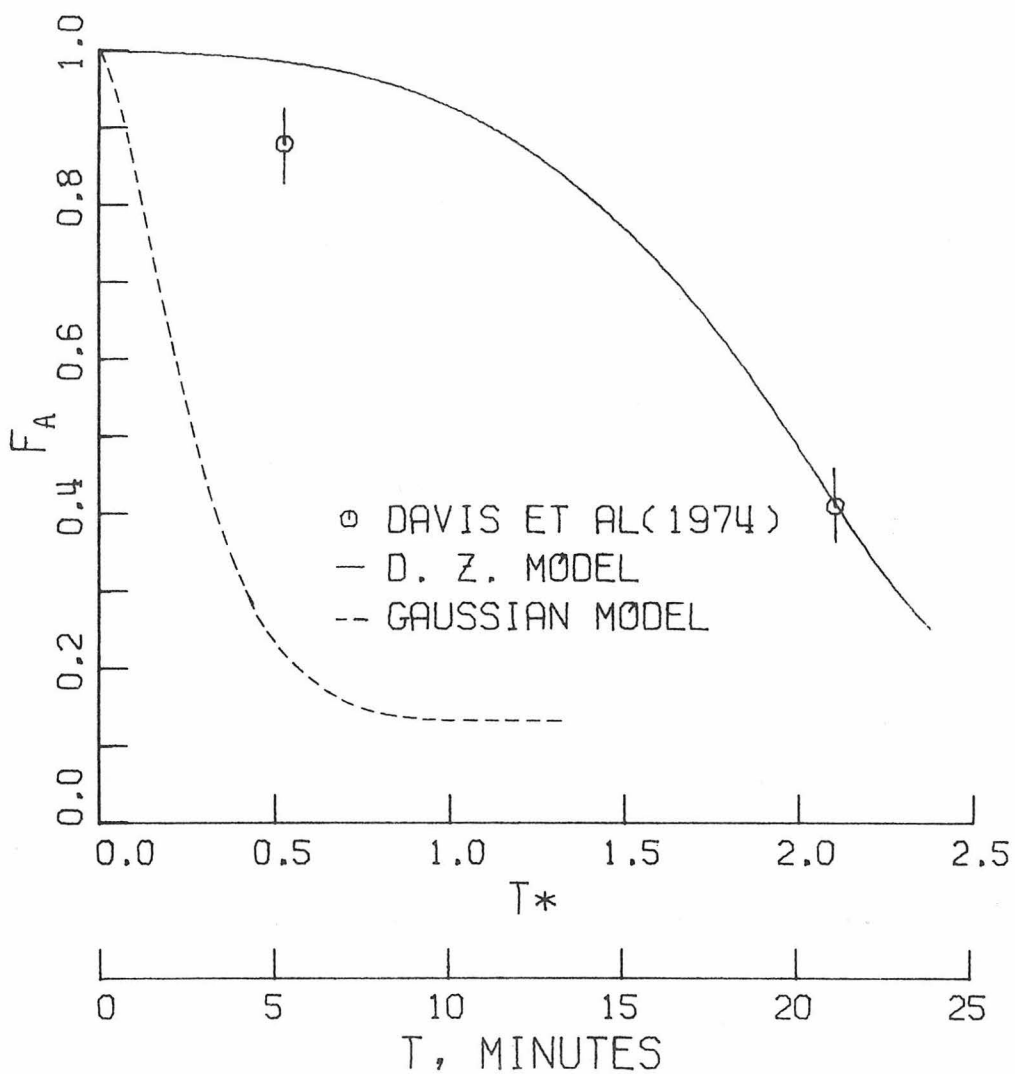
$$\langle B(r,0) \rangle = B_0 \quad \text{for all } r \quad (3.80b)$$

and the boundary conditions are

$$\frac{\partial \langle A \rangle}{\partial r} = \frac{\partial \langle B \rangle}{\partial r} = 0 \quad \text{at } r = 0 \quad (3.81a)$$

$$\langle A \rangle = 0, \quad \langle B \rangle = B_0 \quad \text{at } r \rightarrow \infty \quad (3.81b)$$

where r_0 is the source radius (=5.56 m). We choose a value of $Q_A / \langle u \rangle = 47$ ppm and solve (3.79) with (3.80-81) numerically. The solution $\langle A(r,t) \rangle$ is then used to compute F_A by (3.77). The results are plotted along with the DZ model prediction in Figure 3.16.



$Q_A / \langle U \rangle = 47 \text{ PPM}$
 $B_0 = 0.08 \text{ PPM}$
 $\langle U \rangle = 2.74 \text{ M/SEC}$

Figure 3.16 Comparison of predictions of the DZ model and a Gaussian model. The difference in the predictions is attributed to the neglect of $\langle A'B' \rangle$ in the Gaussian model.

It is immediately clear that the Gaussian model (3.79) is unacceptable: It predicts a 50%-conversion time of NO of three minutes and a much more rapid initial rate of reaction than is observed. If the value of $Q_A/\langle u \rangle$ is increased, the prediction of F_A comes into closer agreement with the observations. However, the NO-emission at the Morgantown plant would have to be 250,000 lb/hr for the predictions of the Gaussian model (3.79) to agree with the observations. Since the dependence of F_A on K is weak, the difference between the prediction of the DZ model and that of the Gaussian model can only be attributed to the neglect of the fluctuation term $\langle A'B' \rangle$. When NO is first emitted, the O_3 in the plume itself is quickly depleted by the reaction, and the excess NO must wait for turbulent mixing to supply fresh O_3 from outside the plume for the reaction to proceed. In this initial period, reactants are basically segregated and $\langle A'B' \rangle$ has a negative value comparable in magnitude to that of the product $\langle A \rangle \langle B \rangle$. Thus, neglecting $\langle A'B' \rangle$ causes an overestimate of the reaction rate as is evident in the rapid decay of F_A in the Gaussian model. By contrast, the DZ model, taking into consideration the effect of $\langle A'B' \rangle$, predicts a minimal decay of F_A during the initial period.

III.5 Discussion

In this chapter we have developed a special version of the DZ model for simulating a point source plume in the planetary boundary layer. The parameters required to implement the model were derived

from data generated by Deardorff's numerical turbulence model. Because the turbulent structure of the unstable boundary layer possesses certain universal characteristics (Deardorff, 1970b), the parameter values we derived may be used in an extremely wide range of simulation studies provided that variables are expressed in terms of the length and time scales z_i and z_i/w_* , respectively, introduced earlier. The large amount of computing time required to derive these parameters is therefore justified. Within the limitations of the available data, we have shown how the DZ model provides a realistic description of chemical reaction in the planetary boundary layer. Unlike many of the currently available closure schemes, the DZ model does not introduce additional differential equations. Also, the difficulty in determining eddy diffusivities was circumvented in our boundary layer plume simulation by computing the total material in cross-sections of the plume as a function of downwind distance. This approach may prove to be of considerable value in air pollution studies. At the same time it renders the computation on applying the DZ model to be extremely simple.

The effect of inhomogeneous mixing on the rates of atmospheric chemical reactions can be studied using the DZ model. We compared the prediction of the DZ model with that of a Gaussian model in which the concentration fluctuations were neglected. The comparison indicates that the rate of the NO-O_3 reaction is greatly reduced by inhomogeneities in the concentration fields produced by atmospheric turbulence. For this reason the common practice in air pollution modeling of neglecting

concentration fluctuation effects can produce large errors when applied to reactions of intermediate and very rapid rates.

We have therefore shown that the concept of diffusion zones is extremely versatile and useful in studying chemical reactions in turbulent flows. Further development of the DZ model should focus on extensions of the model to treat complex reaction systems and applications to sources of various types. There is also the need to develop expressions for the mixing parameters ζ_A , ζ_B that are applicable to situations other than those we have considered here.

IV. LAGRANGIAN MODELS OF TURBULENT CHEMICAL REACTIONS

As noted in Chapter I, there are two general approaches to describe the phenomenon of turbulent chemical reaction: an Eulerian approach, in which the problem is formulated in a coordinate system fixed in space; and a Lagrangian approach in which the behavior of the individual fluid particles is treated. In the previous chapters we proposed a hypothesis to handle the closure problem that one encounters in the Eulerian approach. In this chapter, we view the phenomenon of turbulent chemical reactions from a different perspective by analyzing it with Lagrangian* methods.

Since the Lagrangian description addresses the motions of the individual fluid elements that constitute the fluid flow, it is perhaps more natural physically than the Eulerian description. However, a factor that hinders the wide use of Lagrangian descriptions of turbulence problems is that in order to measure Lagrangian properties, one must somehow track individual fluid particles. This is often difficult to achieve experimentally; and consequently, while the mathematics of Lagrangian methods are inherently simpler than those of Eulerian closure methods, the required multi-particle statistical properties needed for Lagrangian analysis are virtually impossible to determine experimentally. Due to these difficulties very few studies have been based on this approach. (See review in Chapter I.)

* For a detailed discussion on the Lagrangian description, see Monin and Yaglom (1971), Chap. 5.

Recently Lamb (1972a) has succeeded in deriving expressions for the concentration statistics of inert foreign particles in a turbulent fluid from a Lagrangian property that we shall refer to as the occupancy probability density function (opdf). (See Eq. (3.16).) The opdf contains information on the trajectories of the particle motion and in some cases it can be measured experimentally. For example, a demonstration of the measurement of the two-particle opdf was given in Chapter III. However, in cases where a number of different species of particles interact chemically, it is generally infeasible to determine such an opdf for each type of particle. Therefore Lamb (1972b) further derived an approximate relationship that enables one to determine the opdf of a particular type of species which is undergoing a second order reaction with itself from the opdf for the inert particles. Lamb's approach provides in effect both a circumvention of the Eulerian closure problem and a different insight into the nature of nonlinear, turbulent reactions. In particular, this approach can provide concentration statistics of a turbulent plume near its source where many Eulerian closure approximations break down. It is the purpose of this chapter to 1) demonstrate this technique, 2) show that it can be extended to multicomponent chemistry, and 3) derive several simplified models from the general description. These extensions of the approach are necessary in order to facilitate its application to air pollution chemistry.

IV.1 A Lagrangian Description of a Second-Order Chemical Reaction
in a Turbulent Fluid

The problem under consideration in this section is the following. Given the initial distribution of, say, N inert particles and the opdf of the trajectories of these particles, what would be the marginal densities of the opdf if the particles were chemically reactive?

Let the N particles be initially located ($t = 0$) at

$$r_1^0, r_2^0, \dots, r_N^0$$

where

$$r_i^m = (x_i^m, y_i^m, z_i^m)$$

is the *position vector* of the i -th particle at time $t = m\Delta t$ where Δt is some small time interval to be defined subsequently. Let the *trajectory vector* that the i -th particle has followed in m steps in a discrete time-continuous space random process be

$$R_i^m = (r_i^m, r_i^{m-1}, \dots, r_i^2, r_i^1)$$

The opdf that gives the probability density of the trajectories of all N particles can then be defined as

$$P(R_1^m, R_2^m, \dots, R_N^m | r_1^0, r_2^0, \dots, r_N^0) \quad (4.1)$$

The probability density of the trajectories of all N particles, $p(R_1^m, R_2^m, \dots, R_N^m | r_1^0, r_2^0, \dots, r_N^0)$, is defined such that $p(R_1^m, R_2^m, \dots, R_N^m |$

$r_1^0, r_2^0, \dots, r_N^0$) $dR_1^m dR_2^m \dots dR_N^m$ is the probability that the i th particle followed a trajectory such that after the first time step the particle occupied the differential volume of space r_i^1 to $r_i^1 + dr_i^1$, after the second time step occupied the differential volume of space, r_i^2 to $r_i^2 + dr_i^2$, etc., for m steps. Thus, dR_i^m denotes $dr_i^m dr_i^{m-1} \dots dr_i^1$. The opdf defined in (4.1) is strictly a Lagrangian property of the turbulence because it is defined in terms of the positions of inert particles. It has been normalized such that

$$\int_{-\infty}^{\infty} \dots \int_{-\infty}^{\infty} p(R_1^m, \dots, R_N^m | r_1^0, \dots, r_N^0) dR_1^m \dots dR_N^m = 1 \quad (4.2)$$

where

$$dR_i^m = dr_i^m dr_i^{m-1} \dots dr_i^1,$$

and

$$dr_i^m = dx_i^m dy_i^m dz_i^m$$

is a differential volume element centered at r_i^m .

If the opdf in (4.1) is known, the probability that particles follow a particular trajectory can be generated from it simply by integrating over the given trajectory. For example, the probability that the i -th particle will follow a path s in m time steps and the other particles will follow arbitrary paths is

$$\int_{-\infty}^{\infty} \int_s p(R_i^m, \bar{R}_i^m | r_i^0, \bar{r}_i^0) dR_i^m d\bar{R}_i^m \quad (4.3)$$

where an overbar denotes the complement, i.e.,

$$\overline{R}_i^m = R_1^m, R_2^m, \dots, R_{i-1}^m, R_{i+1}^m, \dots, R_N^m \quad (4.4)$$

$$\overline{r}_i^m = r_1^m, r_2^m, \dots, r_{i-1}^m, r_{i+1}^m, \dots, r_N^m \quad (4.5)$$

It is commonly assumed that chemical substances can react only when they are sufficiently close to each other. Therefore, we maintain that in a turbulent fluid, particles have a chance to react only when their trajectories pass within some specified distance of each other at any instant. In particular, we propose the following Lagrangian model of chemical reactions. Consider the bi-molecular chemical reaction $A + B \rightarrow C$. Suppose that each particle has a uniform chemical composition, that is, it can be either of species A or B or the product C. We envision that associated with each particle there is a reaction volume defined as

$$\delta V = a \langle v'^2 \rangle^{\frac{1}{2}} \Delta t$$

where $\langle v'^2 \rangle^{\frac{1}{2}}$ is the r.m.s. amplitude of the particle velocity fluctuations, and a is a characteristic cross-sectional area of reaction for the particle. (If the turbulence is stationary and homogeneous, $\langle v'^2 \rangle^{\frac{1}{2}}$ is a function of neither time nor position.) Within this framework we postulate that if the trajectories of two particles are such that their reaction volumes intersect, then reaction occurs at the point of intersection with probability one. Therefore, during any interval Δt , the

probability that the i -th particle (say of species A) does not react is the probability that its reaction volume does not intersect that of any particle of B during that interval. Since we are approximating a continuous process by a discrete one, the smallness of Δt is necessary to insure that the probability of reaction is not overestimated. For this reason we stipulate that Δt must obey the condition,

$$\delta V = a \langle v'^2 \rangle^{\frac{1}{2}} \Delta t \ll \langle c_{\max} \rangle^{-1}$$

or

$$\Delta t \ll (\langle c_{\max} \rangle a \langle v'^2 \rangle^{\frac{1}{2}})^{-1} = T_R. \quad (4.6)$$

Here $\langle c_{\max} \rangle$ can be interpreted as the average volume occupied by each particle in the region of maximum concentration, and T_R is a reaction time scale.

After two particles collide, it is possible that they form either a single new particle or two new particles. We assume that in the first case, the product species can be represented as two individual uniform particles moving together. In this way the total number of particles under consideration is conserved.

In a situation where N_A particles of species A and $(N - N_A)$ particles of species B are initially distributed at $r_1^0, \dots, r_{N_A}^0$ and $r_{N_A+1}^0, \dots, r_N^0$, respectively, the probability of finding that the i -th particle has identity A and is in the differential volume dr_i^m centered at r_i^m at time $t = m\Delta t$ can be written as

$$p(r_i^m, A) dr_i^m, \quad (4.7)$$

According to the reaction model outlined above, (4.7) is equivalent to the product of the probabilities of two independent events, namely that:

1. The i -th particle was originally an A.
2. The i -th particle never collided with an B-particle, i.e.

it managed to stay outside the reaction volumes of all B-particles on its route to r_i^m .

Let the probability of the first event be ρ_i . Clearly

$$\rho_i = \begin{cases} 1 & \text{if } 1 \leq i \leq N_A, \\ 0 & \text{if } N_A + 1 \leq i \leq N \end{cases} \quad (4.8)$$

Let b^ℓ be the region of cartesian space (x,y,z) outside the reaction volumes δV of all B-particles during the ℓ -th time step, that is

$$\int_{b^\ell} (\cdot) dr = \int_{\infty} (\cdot) dr - \sum_j \int_{\delta V_j} (\cdot) dr \quad (4.9)$$

where j sums over all existing B-particles in the ℓ -th step. Then if we restrict attention to passive reactions, i.e. those that do not perturb the fluid motion, then the probability of the second event above is easily established from (4.3) to be

$$\left\{ \int_{\infty} \int_{b^1} \int_{b^2} \dots \int_{b^{m-1}} p(r_i^m, r_i^{m-1}, \dots, r_i^1, \overline{R_i^{m-1}}) \cdot dr_i^{m-1} \dots dr_i^1 d\overline{R_i^{m-1}} \right\} dr_i^m \quad (4.10)$$

where p is the opdf of a collection of inert particles introduced in (4.1). The condition $(|r_1, \dots, r_N)$ is not explicitly stated in (4.10) but it is understood that all densities under consideration are conditioned on the initial distribution.

Let B^ℓ be such that

$$\int_{B^\ell} p dR_i^\ell = \int_{b^1} \int_{b^2} \dots \int_{b^\ell} p dr_i^\ell \dots dr_i^2 dr_i^1 \quad (4.11)$$

then from (4.7), (4.8) and (4.10), (4.11) we have

$$p(r_i^m, A) dr_i^m = \rho_i \left\{ \int_{\infty} \int_{B^{m-1}} p(r_i^m, R_i^{m-1}, \overline{R_i^{m-1}}) dR_i^{m-1} d\overline{R_i^{m-1}} \right\} dr_i^m$$

Hence

$$p(r_i^m, A) = \rho_i \int_{\infty} \int_{B^{m-1}} p(r_i^m, R_i^{m-1}, \overline{R_i^{m-1}}) dR_i^{m-1} d\overline{R_i^{m-1}} \quad (4.12)$$

This expression gives us the marginal density of finding the i -th particle at r_i^m with identity A at the end of the m -th step.

In order to make (4.12) useful, we have to establish a relationship between $p(r_i^m, A)$ and the density of previous steps. By definitions (4.11) and (4.9) we have

$$\begin{aligned}
 p(r_i^m, A) &= \rho_i \int_{\infty} \int_{B^{m-2}} \int_{\infty} \int_{b^{m-1}} p(r_i^m, r_i^{m-1}, \overline{r_i^{m-1}}, R_i^{m-2}, \overline{R_i^{m-2}}) dr_i^{m-1} d\overline{r_i^{m-1}} dR_i^{m-2} d\overline{R_i^{m-2}} \\
 &= \rho_i \int_{\infty} \int_{B^{m-2}} \int_{\infty} p(r_i^m, r_i^{m-1}, \overline{r_i^{m-1}}, R_i^{m-2}, \overline{R_i^{m-2}}) dr_i^{m-1} d\overline{r_i^{m-1}} dR_i^{m-2} d\overline{R_i^{m-2}} \\
 &= \rho_i \int_{\infty} \int_{B^{m-2}} \int_{\infty} \sum_j \int_{\delta V_j} p(r_i^m, r_i^{m-1}, \overline{r_i^{m-1}}, R_i^{m-2}, \overline{R_i^{m-2}}) dr_i^{m-1} d\overline{r_i^{m-1}} dR_i^{m-2} d\overline{R_i^{m-2}} .
 \end{aligned}
 \tag{4.13}$$

where j sums over all existing B-particles in the $(m-1)$ -th time step. Further reduction of (4.13) requires an expression for the probability that the j -th particle be of species B at the $(m-1)$ -th step. For this to occur we must have:

1. The j -th particle was initially a B. (This event occurs with probability $(1-\rho_j)$.)

2. The j -th particle has a trajectory in A^{m-2} that lies outside the reaction volumes of all A-particles in the time domain $0 < t < (m-2)\Delta t$. Therefore we see that the second term in (4.13) is essentially the joint probability density that the i -th particle was an A until the $(m-1)$ th step, that it lies within the reaction volume of the j -th particle at $(m-1)\Delta t$, and that it is at r_i^m at time $t = m\Delta t$. Concurrently the j -th particle had to be a B-particle initially, it must not have suffered any collision with an A particle during its travels prior to time $t = (m-2)\Delta t$, and it must lie at r_j^{m-1} at time $t = (m-1)\Delta t$. Since any of the particles

with subscripts N_A+1, \dots, N can be a B-particle, this joint probability density can be defined as

$$\rho_i \sum_{j=N_A+1}^N (1-\rho_j) \int_{\infty} \int_{B^{m-2}} \int_{\infty} \int_{\delta V_j} p(r_i^m, r_i^{m-1}, r_j^{m-1}, r_{ij}^{\overline{m-1}}, R_i^{m-2}, R_j^{m-2}, R_{ij}^{\overline{m-2}}) dr_i^{m-1} dr_j^{m-1} dr_{ij}^{\overline{m-1}} dR_i^{m-2} dR_j^{m-2} dR_{ij}^{\overline{m-2}} \quad (4.14)$$

where $r_{ij}^{\overline{m-1}}$ = complement of (r_i^{m-1}, r_j^{m-1}) and $R_{ij}^{\overline{m-2}}$ = complement of (R_i^{m-2}, R_j^{m-2}) . Substituting (4.14) into (4.13), we have

$$\begin{aligned} p(r_i^m, A) = & \rho_i \int_{\infty} \int_{B^{m-2}} \int_{\infty} p(r_i^m, r_i^{m-1}, r_i^{\overline{m-1}}, R_i^{m-2}, R_i^{\overline{m-2}}) dr_i^{m-1} dr_i^{\overline{m-1}} dR_i^{m-2} dR_i^{\overline{m-2}} \\ & - \sum_{j=N_A+1}^N \rho_i (1-\rho_j) \int_{\infty} \int_{A^{m-2}} \int_{B^{m-2}} \int_{\infty} \int_{\delta V_j} \\ & p(r_i^m, r_i^{m-1}, r_j^{m-1}, r_{ij}^{\overline{m-1}}, R_i^{m-2}, R_j^{m-2}, R_{ij}^{\overline{m-2}}) dr_i^{m-1} dr_j^{m-1} dr_{ij}^{\overline{m-1}} dR_i^{m-2} dR_j^{m-2} dR_{ij}^{\overline{m-2}} \end{aligned} \quad (4.15)$$

In essence all we have done is decompose (4.12) into two parts: one associated with the event that the i -th particle suffered no collision with B in the first $(m-2)$ steps (but its path in the $(m-1)$ -th steps is arbitrary) and the second associated with the event that the i -th particle suffered no collision with B in the first $(m-2)$ steps but definitely suffered a collision with B in the $(m-1)$ -th step.

In the reaction $A + B \rightarrow C$, particles of species B behave in every respect like particles of A. Therefore the marginal density of finding the i -th particle at r_i^m and of species B is readily seen to be

$$\begin{aligned}
 p(r_i^m, B) &= (1-\rho_i) \int_{\infty} \int_{A^{m-2}} \int_{\infty} p(r_i^m, r_i^{m-1}, \overline{r_i^{m-1}}, R_i^{m-2}, \overline{R_i^{m-2}}) dr_i^{m-1} d\overline{r_i^{m-1}} dR_i^{m-2} d\overline{R_i^{m-2}} \\
 &\quad - \sum_{j=1}^{N_A} \rho_j (1-\rho_i) \int_{\infty} \int_{B^{m-2}} \int_{A^{m-2}} \int_{\infty} \int_{\infty} \int_{\infty} \\
 &\quad p(r_i^m, r_i^{m-1}, r_j^{m-1}, \overline{r_{ij}^{m-1}}, R_i^{m-2}, R_j^{m-2}, \overline{R_{ij}^{m-2}}) dr_i^{m-1} dr_j^{m-1} d\overline{r_{ij}^{m-1}} dR_i^{m-2} dR_j^{m-2} d\overline{R_{ij}^{m-2}}
 \end{aligned}
 \tag{4.16}$$

By the hypothesis of uniform composition of particles, the i -th particle found at r_i^m can be either of species A, B or C (provided that $A + B \rightarrow C$ is the only reaction possible). Therefore it is clear that the probability density that the i -th particle is of species C and is located at r_i^m at time $t = m\Delta t$ is

$$p(r_i^m, C) = p(r_i^m) - p(r_i^m, A) - p(r_i^m, B)
 \tag{4.17}$$

Eqs. (4.15-17) are our working equations. It should be kept in mind that they are conditioned on the initial distribution of particles (r_1^0, \dots, r_N^0) . Once the opdf is given, the concentration statistics of reactive materials undergoing a second-order reaction in turbulence can be evaluated.

The formulation given above can easily be extended to the case where particles are not released at the same moment. This is done by redefining the inert opdf as

$$p = p(r_1^m, r_2^m, \dots, r_{N_m}^m; r_1^{m-1}, \dots, r_{N_{m-1}}^{m-1}; \dots, r_1^1, r_2^1, \dots, r_{N_1}^1 | r_1^0, \dots, r_N^0; r_{n+1}^1, \dots, r_{N_1}^1, \dots, r_{N_{m-1}+1}^m \dots r_{N_m}^m) \quad (4.18)$$

where N_m is the total number of particles present in the m -th time step. Obviously, among the N_m particles present, $N_m - N_{m-1}$ are released in this current time step. The corresponding formulas based on this opdf are essentially the same as (4.15-17). Only the formula for $p(r_i^m, A)$ is presented below:

$$p(r_i^m, A) = \begin{cases} \rho_i \int_{\infty} \int_{B^{m-2}} \int_{\infty} \int_{\infty} p \, dr_i^{m-1} \overline{dr_i^{m-1}} dR_i^{m-2} \overline{dR_i^{m-2}} \\ - \sum_{\substack{j=1 \\ j \neq i}}^{N_{m-1}} \rho_i (1 - \rho_j) \int_{\infty} \int_{A^{m-2}} \int_{B^{m-2}} \int_{\infty} \int_{\delta V_j} p \, dr_i^{m-1} dr_j^{m-1} \overline{dr_{ij}^{m-1}} dR_i^{m-2} dR_j^{m-2} \overline{dR_{ij}^{m-2}} \\ \rho_i, \end{cases} \begin{cases} \text{for } i = 1, \dots, N_{m-1} \\ \text{for } i = N_{m-1} + 1, \dots, N_m \end{cases} \quad (4.19)$$

where p is defined in (4.18).

Because the opdf's defined in (4.1) and (4.18) are virtually impossible to determine, their forms must be approximated. This is done in the following section. Several Lagrangian models of turbulent chemical reactions are derived thereby.

IV.2 Lagrangian Models of Turbulent Chemical Reactions

Before we begin to derive the models, it is appropriate to introduce the following formulas from Lamb (1972a) which relates the opdf's to the concentration statistics. Let $c(r,t)$ denote the instantaneous concentration of some species measured at location r (r is a position vector) at time t , and let angle brackets denote ensemble average, then

$$\langle c(r,t) \rangle = \lim_{\Delta V \rightarrow 0} \frac{1}{\Delta V} \sum_{i=1}^N \int_{\Delta V} p(r_i^m) dr_i^m \quad (4.20a)$$

$$\langle c^2(r,t) \rangle = \lim_{\Delta V \rightarrow 0} \frac{1}{\Delta V^2} \sum_{\substack{i=1 \\ j \neq i}}^N \sum_{j=1}^N \int_{\Delta V} \int_{\Delta V} p(r_i^m, r_j^m) dr_i^m dr_j^m \quad (4.20b)$$

where ΔV is the sampling volume centered at r . Since the sampling volume is small, we can assume that $p(r_i^m)$ and $p(r_i^m, r_j^m)$ do not vary much in ΔV .

This leads to

$$\langle c(r,t) \rangle = \sum_{i=1}^N p(r_i^m) \Big|_{r_i^m=r} \quad (4.21a)$$

$$\langle c^2(r,t) \rangle = \sum_{\substack{i=1 \\ j \neq i}}^N \sum_{j=1}^N p(r_i^m, r_j^m) \Big|_{r_i^m=r_j^m=r} \quad (4.21b)$$

If the total number of particles is large, the summation over all the particles can be replaced by an integral and the formulas then read:

$$\langle c(r,t) \rangle = \int_0^t \int_{-\infty}^{\infty} p(r,t|\rho,\tau) S(\rho,\tau) d\rho d\tau \quad (4.22a)$$

$$\langle c^2(r,t) \rangle = \int_0^t \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} p(r,r,t|\rho,\tau;\rho',\tau') S(\rho,\tau) S(\rho',\tau') d\rho d\rho' d\tau d\tau' \quad (4.22b)$$

where $p(r,t|\rho,\tau)$ is the pdf that an inert particle released at location ρ and time τ will be found at location r and time t . Eq. (4.22b) was introduced earlier in (3.16) with similar notation. These formulas will be used frequently in the subsequent analyses.

Two-Step Model

In what follows we shall restrict ourselves to particles of reactive materials immediately after they are released. Setting $m = 2$ in (4.15), we have

$$p(r_i^2, A) = \rho_i p(r_i^2) - \sum_{\substack{j=1 \\ j \neq 2}}^N \rho_i (1 - \rho_j) \int_{r_j^{-\frac{1}{2}\delta V}}^{r_j^{+\frac{1}{2}\delta V}} p(r_i^2, r_i^1, r_j^1 | r_i^0, r_j^0) dr_i^1 dr_j^1 \quad (4.23)$$

where the dependence of p on r_ℓ^1 for $\ell \neq i, j$ has been removed by integration. Unlike the general formulas (4.15), the information required for this equation, viz $p(r_i^2, r_i^1, r_j^1 | r_i^0, r_j^0)$, is more readily accessible and hence the concentration statistics of reactive materials can be obtained. Summing the whole equation of (4.23) over for $i = 1, 2, \dots, N$, and applying (4.21), we obtain

$$\langle A(r,t) \rangle = \langle A_I(r,t) \rangle - \sum_{i=1}^N \sum_{\substack{j=1 \\ i \neq j}}^N \rho_i (1 - \rho_j) \int \int_{\infty} p \, dr_i' dr_j' \quad (4.24)$$

where

$$p = p(r_i^2, r_i^1, r_j^1 | r_i^0, r_j^0)$$

$$r = r_i^2$$

$$t = 2\Delta t$$

and A_I is the concentration that A would have if it were inert.

Since (4.24) is valid only for $0 < t < 2\Delta t$ and the magnitude of Δt is restricted by condition (4.6), namely $\Delta t \ll (k \langle c_{\max} \rangle^{-1})$, applications of this model to fast reactions are severely limited. However, in these cases (4.24) might prove to be of great value in resolving the initial phases of reaction that many closure schemes are unable to describe. For this reason we expect that the two-step model is useful only in conjunction with closure schemes that normally do not apply near the source.

Homogeneous Model

The second model derived from the Lagrangian description is the homogeneous model. This model considers the case of a dilute mixture of substances A and B reacting to form C ($A + B \rightarrow 2C$) in a well-stirred, continuous reactor. It is assumed that the cell is continually stirred so that if any foreign substance is introduced into the cell

at any time, it becomes thoroughly mixed in the reactor within a time T_m . Suppose that the time scale T_R in (4.6) of the reaction is sufficiently large that an interval Δt exists such that

$$T_R \gg \Delta t \gg T_m \quad (4.25)$$

This value of Δt enables us to use formula (4.15-17) because it satisfies (4.6).

Since Δt is much larger than the mixing time T_m , it is expected that the position of any particle at time $n\Delta t$ is statistically independent of its position at any other time $m\Delta t$, $m \neq n$, and that the positions of the particles at any given time are statistically independent of one another. These two results yield the following probability relationships:

$$\begin{aligned} p(R_i^m) &= p(r_i^m)p(R_i^{m-1}) = p(r_i^m)p(r_i^{m-1})p(R_i^{m-2}) = \dots \\ &= p(r_i^m)p(r_i^{m-1})p(r_i^{m-2}) \dots p(r_i^1) \end{aligned} \quad (4.26)$$

and

$$\begin{aligned} p(r_i^m, \overline{r_i^m}) &= p(r_i^m)p(\overline{r_i^m}) = p(r_i^m)p(r_j^m)p(\overline{r_{ij}^m}) = \dots \\ &= p(r_1^m)p(r_2^m) \dots p(r_N^m) \end{aligned} \quad (4.27)$$

respectively. Furthermore, when the above conditions are satisfied it is logical that the marginal densities $p(r_i^m)$ of each particle will be equal and will be constant in space and time. Thus, we assume

$$p(r_i^m) = q = \begin{cases} \frac{1}{V}, & r_i^m \text{ inside } V, \quad i = 1, 2, \dots, N \\ 0, & r_i^m \text{ outside } V, \quad m = 1, 2, \dots, \end{cases} \quad (4.28)$$

where V is the volume of the reactor. Hence

$$\int_{\infty} q \, dr_i^{\ell} = \int_V \frac{1}{V} \, dr_i^{\ell} = 1$$

We now turn to the evaluation of (4.15). By applying (4.26-28) it is easy to show that (4.15) reduces to

$$p(r_i^m, A) = p(r_i^{m-1}, A) - \sum_{j=N_A+1}^N \delta V_j \, p(r_i^{m-1}, A; r_j^{m-1}, B) \quad (4.29)$$

where

$$p(r_i^{m-1}, A; r_j^{m-1}, B) = \int_{\infty} \int_{A^{m-2}} \int_{B^{m-2}} p(r_i^{m-1}, r_j^{m-2}, R_j^{m-2}, \overline{R_{ij}^{m-2}}) \, dr_i^{m-2} dR_j^{m-2} d\overline{R_{ij}^{m-2}} \quad (4.30)$$

which in this case is simply $\int_{\infty} \int_{A^{m-2}} \int_{B^{m-2}} q^{(m-2)N+2} \, dR_i^{m-2} dR_j^{m-2} d\overline{R_{ij}^{m-2}}$

Summing (4.29) over i for $i = 1$ to N_A , and applying (4.21) we have

$$\langle A(r, t+\Delta t) \rangle = \langle A(r, t) \rangle - \delta V \langle A(r, t) B(r, t) \rangle \quad (4.31)$$

where $t = (m-1)\Delta t$ and the formula (4.21b) has been generalized to include the case of two species.

Since $\delta V = a\langle v'^2 \rangle^{\frac{1}{2}} \Delta t$ by definition, we can rearrange (4.31) to give

$$\frac{\langle A(r, t+\Delta t) \rangle - \langle A(r, t) \rangle}{\Delta t} = - a\langle v'^2 \rangle^{\frac{1}{2}} \langle A(r, t) B(r, t) \rangle \quad (4.32)$$

Taking the limit of (4.32) as Δt approaches zero, we have

$$\frac{\partial \langle A(r, t) \rangle}{\partial t} = - k \langle A(r, t) B(r, t) \rangle \quad (4.33)$$

where $k = a\langle v'^2 \rangle^{\frac{1}{2}}$

One interesting point is observed here is that in the case of turbulent mixing of inert substances, the concentrations will be uncorrelated if the particle motions are statistically independent. This can easily be seen by applying the condition (4.26-27) to (4.20-4.22). One finds

$$\langle c^2(r, t) \rangle = \langle c(r, t) \rangle^2 \quad .$$

Yet in the cases where the materials are reactive, we see from (4.33) that statistical independence of particle motions (a condition of the turbulence) does not necessarily cause uncorrelated concentrations. The concentration correlation here is caused by the chemical reaction itself. When two species react to form another species at certain point, the concentrations of these two reactants instantly become smaller than

concentrations at the vicinity of this point. Hence concentration fluctuations or correlations are created. Since reactions are occurring at the molecular level, only molecular diffusion can destroy concentration correlations of this kind.

Let T_D be the time scale associated with molecular diffusion. (Note that T_D could be greater than T_m , the turbulent mixing time scale in (4.25).) If $T_R \gg T_D$, molecular diffusion destroys the concentration correlation much faster than the reaction is able to create it, then we have

$$\frac{\partial \langle A(r,t) \rangle}{\partial t} = \frac{\partial \langle B(r,t) \rangle}{\partial t} = -k \langle A(r,t) \rangle \langle B(r,t) \rangle \quad (4.34)$$

Eq. (4.34) is the familiar expression often used in chemical kinetics in describing the rate of a second-order reaction $A + B \rightarrow C$ in a well-stirred tank reactor. Many applications of this formula can be found (e.g. Vassilatos and Toor, 1965). For reactions with higher speed, this formula often fails to apply. (See Glassman and Eberstein, 1963.)

Markov Process

In deriving the homogeneous model, the particle positions at any time t were assumed to be statistically independent of their positions at all previous times. In real turbulence, however, particles "remember" their past histories. To account for this characteristic of turbulent motion, we may assume that particles remember their

trajectories during the period Δt immediately prior to any given instant but not anything earlier. This is commonly known as the Markov assumption. If T is the Lagrangian time scale of the turbulence, the Markov assumption implies

$$\Delta t \geq T \quad (4.35)$$

With this assumption, we have

$$p(r_i^m, r_i^{m-1}, \overline{r_i^{m-1}}, R_i^{m-2}, \overline{R_i^{m-2}}) = p(r_i^m | r_i^{m-1}, \overline{r_i^{m-1}}) p(r_i^{m-1}, \overline{r_i^{m-1}}, R_i^{m-2}, \overline{R_i^{m-2}}) \quad (4.36)$$

Since particles do not interfere with one another's motion,

$$p(r_i^m | r_i^{m-1}, \overline{r_i^{m-1}}) = p(r_i^m | r_i^{m-1}) . \quad (4.37)$$

Applying (4.36) and (4.37) to (4.15), integrating, and using definitions (4.12) and (4.30), we obtain

$$\begin{aligned} p(r_i^m, A) &= \int_{\infty} p(r_i^m | r_i^{m-1}) p(r_i^{m-1}, A) dr_i^{m-1} \\ &- \sum_{j=N_A+1}^N \int_{\infty} \int_{\delta_j} p(r_i^m | r_i^{m-1}) p(r_i^{m-1}, A; r_j^{m-1}, B) dr_i^{m-1} dr_j^{m-1} \end{aligned} \quad (4.38)$$

We note that the transition probability density $p(r_i^m | r_i^{m-1})$ is independent of chemical reaction. If the turbulence is homogeneous and stationary, we have

$$p(r_i^m | r_i^{m-1}) = p(r_i^m - r_i^{m-1}) = p(\Delta x_1, \Delta x_2, \Delta x_3) \quad (4.39)$$

where Δx_k ($k = 1, 2, 3$) are components of the vector $\Delta r = r_i^m - r_i^{m-1}$. The function $p(\Delta x_1, \Delta x_2, \Delta x_3)$ is determined by the state of the turbulence.

Substituting (4.39) into (4.38), summing the equation over $i = 1$ to N_A and applying (4.21), we have (provided that the reaction volume ΔV is sufficiently small)

$$\begin{aligned} \langle A(r, t + \Delta t) \rangle &= \int_{\infty} p \langle A(r - \Delta r, t) \rangle d(r - \Delta r) \\ &- \delta v \int_{\infty} p \langle A(r - \Delta r, t) B(r - \Delta r, t) \rangle d(r - \Delta r) \end{aligned} \quad (4.40)$$

where $t = m\Delta t$ and $r = r_i^m$. Eq. (4.40) can be expanded into a Taylor series about the point (r, t) :

$$\begin{aligned} \langle A(r, t) \rangle + \Delta t \frac{\partial \langle A(r, t) \rangle}{\partial t} + O(\Delta t)^2 &= \int_{\infty} p \{ \langle A(r, t) \rangle \\ &- \Delta x_k \frac{\partial \langle A(r, t) \rangle}{\partial x_k} + \frac{1}{2} \Delta x_k \Delta x_\ell \frac{\partial^2 \langle A(r, t) \rangle}{\partial x_k \partial x_\ell} + O(\Delta x)^3 \} d(\Delta r) \\ - \delta v \int_{\infty} p \{ \langle A(r, t) B(r, t) \rangle - \Delta x_k \frac{\partial \langle A(r, t) B(r, t) \rangle}{\partial x_k} + O(\Delta x)^2 \} d(\Delta r) \end{aligned} \quad (4.41)$$

where $d(\Delta r) = d(\Delta x_1) d(\Delta x_2) d(\Delta x_3)$. The moments of the probability density $p = p(\Delta x_1, \Delta x_2, \Delta x_3)$ are defined by

$$1 = \int_{\infty} p d(\Delta r) \quad (4.42a)$$

$$\langle \Delta x_k \rangle = \int_{\infty} \Delta x_k p d(\Delta r) \quad (4.42b)$$

$$\langle \Delta x_k \Delta x_\ell \rangle = \int_{-\infty}^{\infty} \Delta x_k \Delta x_\ell p \, d(\Delta r) \quad (4.42c)$$

Substituting (4.42) into (4.41), we have

$$\begin{aligned} \frac{\partial \langle A(r,t) \rangle}{\partial t} + O(\Delta t) &= - \frac{\langle \Delta x_k \rangle}{\Delta t} \frac{\partial \langle A(r,t) \rangle}{\partial x_k} + \frac{1}{2} \frac{\langle \Delta x_k \Delta x_\ell \rangle}{\Delta t} \frac{\partial^2 \langle A(r,t) \rangle}{\partial x_k \partial x_\ell} \\ &+ O\left(\frac{\Delta x^3}{\Delta t}\right) - \frac{\delta V}{\Delta t} \langle A(r,t) B(r,t) \rangle + \frac{\delta V \langle \Delta x_k \rangle}{\Delta t} \frac{\partial \langle A(r,t) B(r,t) \rangle}{\partial x_k} + O\left(\frac{\delta V \Delta x^2}{\Delta t}\right) \end{aligned} \quad (4.43)$$

The moment $\langle \Delta x_k \rangle$ is the k-th component of the mean displacement of a particle in a time Δt . If the turbulence is homogeneous, only a mean fluid motion can produce a non-zero $\langle \Delta x_k \rangle$. Let the mean fluid velocity be a constant and have components $\langle u_k \rangle$. Then

$$\langle \Delta x_k \rangle = \langle u_k \rangle \Delta t \quad (4.44)$$

The second moments $\langle \Delta x_k \Delta x_\ell \rangle$ are influenced by the turbulent velocity fluctuation and can be written in the form

$$\langle \Delta x_k \Delta x_\ell \rangle = R_{k\ell} \Delta t + \langle u_k \rangle \langle u_\ell \rangle \Delta t^2 \quad (4.45)$$

where

$$R_{k\ell} = \lim_{t \rightarrow \infty} \int_0^t \{ \langle v'_k(t) v'_\ell(t-t') + v'_k(t-t') v'_\ell(t) \rangle \} dt' \quad (4.46)$$

and v'_k is the Lagrangian velocity of a particle measured with respect to the coordinate system moving with velocity $(\langle u_1 \rangle, \langle u_2 \rangle, \langle u_3 \rangle)$.

Substituting (4.44-45) into (4.43) and taking the limit of $\Delta t \rightarrow 0$, we get

$$\frac{\partial \langle A(r,t) \rangle}{\partial t} + \langle u_k \rangle \frac{\partial \langle A(r,t) \rangle}{\partial x_k} = \frac{1}{2} R_{k\ell} \frac{\partial^2 \langle A(r,t) \rangle}{\partial x_k \partial x_\ell} - k \langle A(r,t) B(r,t) \rangle \quad (4.47)$$

where $k = \delta v / \Delta t = a \langle v'^2 \rangle^{\frac{1}{2}}$ as in (4.33).

In actuality, the limit $\Delta t \rightarrow 0$ cannot be achieved because of the restriction (4.35). To approach this limit, we should first normalize the coordinates to $t^* = t/T$ and $x_k^* = x_k/L_k$, where T and L_k are characteristic time and length scales, respectively, of the concentration $\langle A \rangle$. We then let $\Delta t^* \rightarrow 0$ in the limit of large T and L_k . That is to say, (4.47) is valid when

$$T \gg \Delta t \quad (4.48a)$$

$$L_k \gg \Delta x_k \quad (4.48b)$$

Combining (4.35) with (4.48), and assuming the Lagrangian length scale of the turbulence to be $\langle v'^2 \rangle^{\frac{1}{2}} T$, we have

$$T \gg T \quad (4.49a)$$

$$L_k \gg \langle v'^2 \rangle^{\frac{1}{2}} T \quad (4.49b)$$

Also, conditions (4.35) and (4.6) can be combined to give

$$T k \langle c_{\max} \rangle \ll 1 \quad (4.50)$$

Equation (4.47) is similar to the Eulerian diffusion equation of reactive species which we introduced frequently in the previous

chapters. The LHS of (4.47) is the convective transport of species A, while the effects of turbulent and molecular diffusion are both included in the first term of the RHS. The second term in the RHS is the reaction decay. Because this term represents an additional unknown quantity, Eq. (4.47) is not closed. Through an extension of Prandtl's mixing-length hypothesis, Lamb (1973) showed that

$$\langle A(r,t)B(r,t) \rangle = \langle A(r,t) \rangle \langle B(r,t) \rangle . \quad (4.51)$$

provided that conditions (4.49-50) are satisfied and that the reaction time scale T_R is much larger than the diffusion time scale T_D . A special case of (4.47) with assumption (4.51) was solved in Chapter III. In general, the Markov model can only be used for slow reactions at distance far from the source.

IV.3 Discussion

We have demonstrated the Lagrangian techniques of describing turbulent chemical reaction. Three simplified models were presented but the applications of each appear to be limited. Although the Lagrangian description gives a better picture of the physical phenomena than the Eulerian methods, its use is greatly hindered by the lack of information of the opdf. Only in a limited number of cases can the opdf be measured or hypothesized (e.g., Chapter III.) Efforts were made by the author to numerically measure the complete opdf in a one-dimensional turbulence simulated by the Burgers' equation (Burgers, 1948). Because Burger's model pertains to compressible flow,

it was found that particles tend to cluster together rather than separate. Consequently the result of such studies are of little direct value to problems associated with incompressible fluids.

It therefore appears that the Lagrangian approach to the phenomena of turbulent chemical reaction has rather limited application at present. Experimental techniques to measure Lagrangian properties of turbulence will probably remain very difficult. However, with the present speed of growth in computational technology, it may soon become possible to obtain a complete opdf through numerical simulations. Further development of the Lagrangian method therefore will rely on the advancement of numerical simulation of turbulent flows.

V. CONCLUSION AND RECOMMENDATIONS FOR FURTHER WORK

In the thesis we have demonstrated the importance of the effect of inhomogeneous mixing on the rates of chemical reactions in turbulent flows and have developed what we call a diffusion zone (DZ) model for describing these effects. Turbulent mixing (macromixing), molecular diffusion (micromixing), and chemical kinetics have been incorporated into the diffusion zone model. It was shown that the model gives a realistic description of the phenomenon of turbulent chemical reactions in chemical reactors. The model was also applied to the study of the NO-O_3 reaction in a power-plant plume in the planetary boundary layer. The prediction of the model showed qualitative agreement with observations in a case study. We also developed a Lagrangian model of turbulent chemical reactions but its applicability is much more restricted than that of the diffusion zone model. Following is a list of conclusions drawn on the work of this thesis and recommendations for future work.

1. The rates of decay of the mean concentrations of species involved in second-order isothermal reactions in a turbulent fluid can be predicted using the so-called diffusion zone model developed here. The model requires as inputs the mean and mean square concentration of inert species released from the same sources in the same turbulent fluid.

2. The diffusion zone model can be applied to turbulent chemical reactions of any speed. However, in the limits of very slow and very rapid reaction, some simplification may be made. In the case of slow reactions, the mean concentrations alone are adequate to determine the reaction rates. In the case of very rapid reaction, Toor's model may be used for reactor design purposes. The Damkohler number α , defined in (2.42), provides a measure of the reaction speed. The ranges of Damkohler number to which each of the models just mentioned is applicable are summarized in Table 5.1.

3. The inert concentration statistics required by the DZ model can be obtained in one of the three ways: direct measurement, measurement of the conversion rate of a very rapid reactions (Toor, 1962) or by numerical simulation. A numerical technique for measuring the mean-square concentration of an inert species was introduced in this work.

4. The effect of inhomogeneous mixing on the rate of the NO-O_3 reaction in the atmosphere appears to be large. It was shown that the segregation of NO in a point source plume limited considerably the rate of reaction of this species with ambient ozone.

5. The 50% conversion time-scale of NO emitted from a stationary point source into an ozone laden atmosphere was pre-

Table 5.1 Recommended Use of Models for Turbulent Chemical Reactions of the Type $A + B \rightarrow P$

| (*) | Recommended Model for $-k \ll AB$ |
|---|---|
| $\alpha = \frac{k \delta^2 \min\{a_0, b_0\}}{\max\{D_A, D_B\}}$ | Reactor Analysis Atmospheric Reactions |
| Definition of the Reaction Regime | |
| < 0.01 | -k < A < B > |
| Very Slow (Kinetics-Controlled) | -k < A < B > |
| 0.01 - 50. | DZ model |
| Intermediate or Rapid | DZ model |
| 50. < | Toor (1962) |
| Very Rapid (Diffusion-Controlled) | DZ model |

* See (2.42) for details.

dicted by the DZ model. The prediction is in good agreement with observations.

6. The Lagrangian approach is hindered by two main problems: the lack of information regarding the trajectories of fluid particles in turbulent fluid, and the closure problem that arises when materials that undergo nonlinear chemical reactions are treated. The first problem can be alleviated to a large degree by simulating particle trajectories in turbulence numerically. A technique for simulating the trajectories and joint displacement probability density function of two particles in a three-dimensional turbulence was developed in this work. The closure problem associated with the Lagrangian method is subject to treatment by any one of the Eulerian schemes described in Chapter I.

In summary, we have studied turbulent chemical reactions in general and have developed a useful tool, the diffusion zone model, for treating this phenomenon mathematically. The capability of the DZ model to predict decay of reactive species in turbulent flows has been tested using both laboratory and atmospheric observations. It is recommended that the possibility of applying the DZ model to more complex kinetics be investigated; with an expanded

capability of this type it would be possible, for instance, to examine the effect of turbulent mixing on the selectivity* or the product distribution in a reactor (Brodkey, 1974), or in the area of air pollution modeling, to study controversial issues such as ozone build-up in the far end of a power-plant plume (Davis et al, 1974). Some ideas of this extension are discussed in Appendix D.

A few other studies may give improvements of the predictions of the DZ model. Experimental measurements of the diffusion zones is one example. By choosing proper sample sizes, continuous measurements of the instantaneous concentrations of two inert species mixing in a turbulent flow would provide a quantitative picture of the diffusion zones. The mixing parameters ζ_A , ζ_B could then be computed from observed data.

Another problem of interest is the treatment of mixing fluid elements of various 'ages'. This problem arises in studying atmospheric reactions among pollutant species emitted at different times. This problem is also briefly discussed in Appendix D.

* The selectivity problem refers to the fact that in a complex reaction system, a species is often involved in more than one reaction, and that depending on the reaction condition, such as temperature or turbulent mixing, the species may be consumed by certain reactions but not others.

APPENDIX A. MULTIJET REACTOR DATA

The purpose of this appendix is to give a brief description of the multijet reactor data, and how we used the data to calculate correlations like r_{AB} in Chapter II.

As noted in Chapter I, Toor (1962) developed a theory which relates the concentration statistics of two species undergoing a very rapid irreversible reaction in a turbulent mixer to that of an inert species in an identical mixer. To test the theory experimentally, Vassilatos and Toor (1965) designed an ideal one-dimensional turbulent reactor the head of which is made of some 100 small nozzles. (Figure A.1) Reactants are fed through alternate jets to simulate a cross-sectionally uniform concentration profile. The reactor was modified in a subsequent work (Mao, 1969). Both Vassilatos and Mao made extensive measurements in the reactor, covering a wide range of reaction speeds. However, their results only offer a partial test to Toor's theory because no inert mixing data were taken. This inert mixing data as well as axial velocity profile were later supplied by Brodkey's group (McKelvey, 1968; Zakařycz, 1971). Toor's theory has since been shown to be good by these works and some others (Keeler et al, 1965; Torrest and Ranz, 1970). The reactor is then used to test other hypothesis (e.g., Mao and Toor, 1971). It appears that data relating to this reactor are the only set of data available which contains the minimal required information to test models for turbulent chemical reactions.

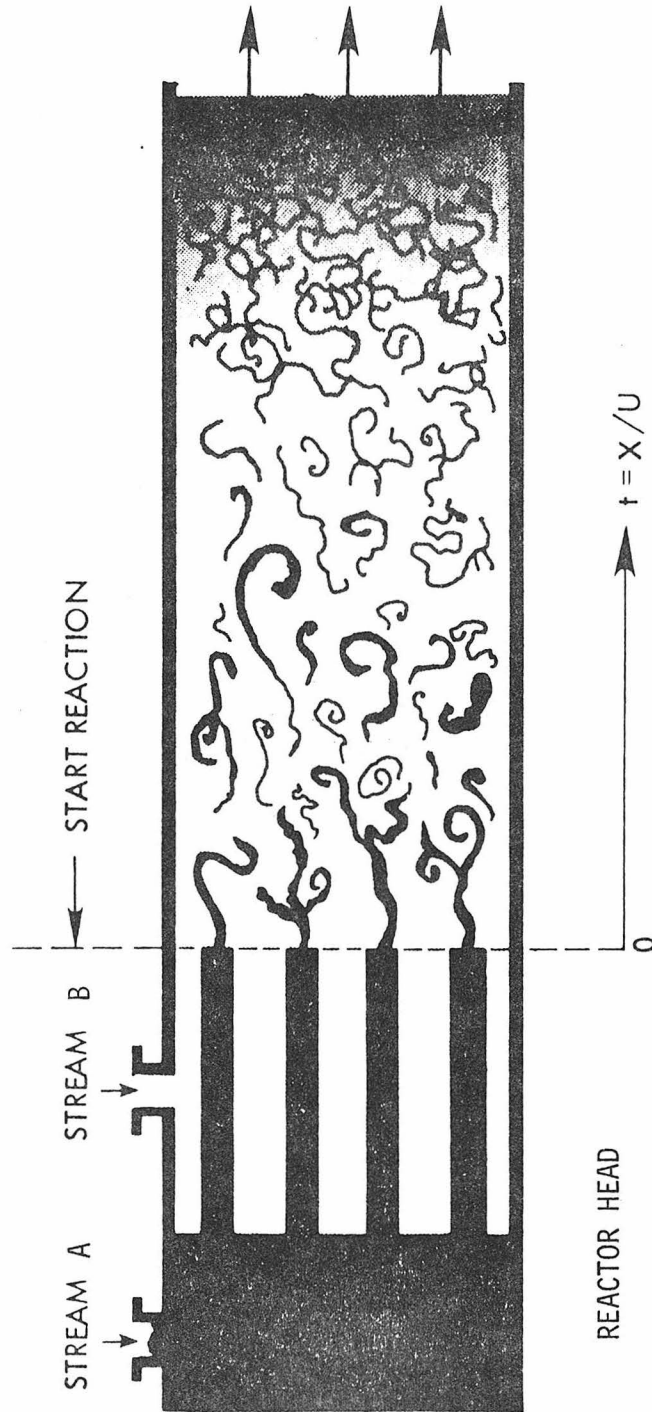


Figure A.1 Schematic diagram of the multijet reactor. The reactor head is composed of some 100 small tubes. Reactants are fed through alternate tubes to simulate a cross-sectionally uniform concentration profile. (Source: Hill, 1976)

Ideally, we can directly apply the inert mixing data of McKelvey (1968) to our diffusion-zone model and test the model prediction against reaction data. Unfortunately, McKelvey reported his mean square concentration fluctuation data in a normalized form without specifying the normalizing factor. An alternative is to use Toor's theory to convert the data for very rapid stoichiometric reaction to the inert mixing data. This conversion is reliable because Toor's theory has been well established in this reactor. However, it rules out the possibility of comparing the DZ model with Toor's model in the case of very rapid stoichiometric reactions.

As noted in (1.23), the theory of Toor (1962) says that for very rapid reactions with stoichiometric feed of two species

$$F_{\infty}(x) \equiv \frac{\langle A(x) \rangle}{\langle A(x_0) \rangle} = \frac{\langle A_I' B_I' (x) \rangle^{\frac{1}{2}}}{\langle A_I' B_I' (x_0) \rangle^{\frac{1}{2}}} \quad (\text{A.1})$$

where $x = \langle u(x) \rangle t$ is the axial position along the reactor length and x_0 is some reference point where the reactor has reached the state of cross-sectional homogeneity. Toor and his coworkers have designed the multijet reactor in such a way that this state is approximately reached at the inlet of the reactor, and hence $x = 0$. Toor (1969) further derived a relationship between the concentration fluctuations of two unpremixed inert species fed into this reactor:

$$\frac{\langle A_I' B_I' (x) \rangle}{\langle A_I' B_I' (0) \rangle} = \frac{\langle A_I'^2 (x) \rangle}{\langle A_I'^2 (0) \rangle} = \frac{\langle B_I'^2 (x) \rangle}{\langle B_I'^2 (0) \rangle} \quad (\text{A.2})$$

Since the concentration field is assumed cross-sectionally homogeneous, the initial concentration fluctuations can be obtained theoretically (Keeler et al, 1965) by averaging over the cross-section. If half of the nozzles feed A and half of them B, which is true in Mao's reactor and approximately true for Vassilatos', it is easy to show that

$$\begin{aligned} \langle A_I'^2(0) \rangle &= \langle A_I \rangle^2 \\ \langle B_I'^2(0) \rangle &= \langle B_I \rangle^2 \end{aligned} \quad (A.3)$$

Since the species are not premixed we have

$$\langle A_I' B_I'(0) \rangle = -\langle A_I \rangle \langle B_I \rangle \quad (A.4)$$

Combining (A.1) through (A.4), we obtain

$$\Gamma_{AB}(x) \equiv \frac{\langle A_I B_I(x) \rangle}{\langle A_I \rangle \langle B_I \rangle} = 1 + \frac{\langle A_I' B_I'(x) \rangle}{\langle A_I \rangle \langle B_I \rangle} = 1 - F_\infty^2(x) \quad (A.5)$$

$$\Gamma_A(x) \equiv \frac{\langle A_I^2(x) \rangle}{\langle A_I \rangle^2} = 1 + \frac{\langle A_I'^2(x) \rangle}{\langle A_I \rangle^2} = 1 + F_\infty^2(x) = \Gamma_B \quad (A.6)$$

where $\langle A_I \rangle$, $\langle B_I \rangle$ are independent of x , due to homogeneity.

Relationships (A.5) and (A.6) are used to convert the concentration decay of very rapid stoichiometric reactions $F_\infty(x)$ to the inert concentration correlations Γ_{AB} , Γ_A and Γ_B . Conversion data of reactions with rate constants greater than 10^7 l/g-mole-sec are not distinguishable from one another, and are considered to be diffusion-limited (Mao, 1969). We shall use the mean concentration data of these reactions for F_∞ . Figures A.2-A.4 show the data of F_∞ for three different flow conditions. A cubic spline function has been used to interpolate data between measurements. The interpolated curves of F_∞ are then used in the model calculation.

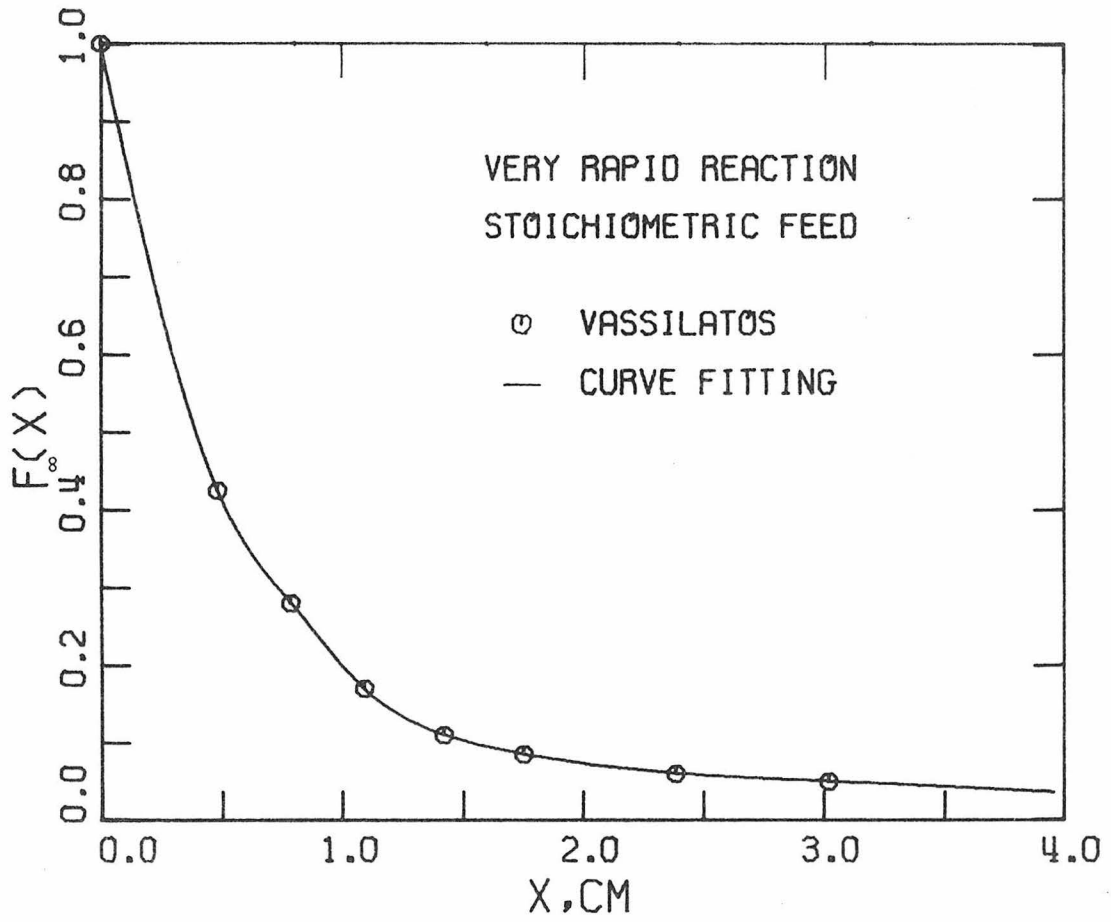


Figure A.2 Extent of reaction of a very rapid stoichiometric reaction: Data of Vassilatos.

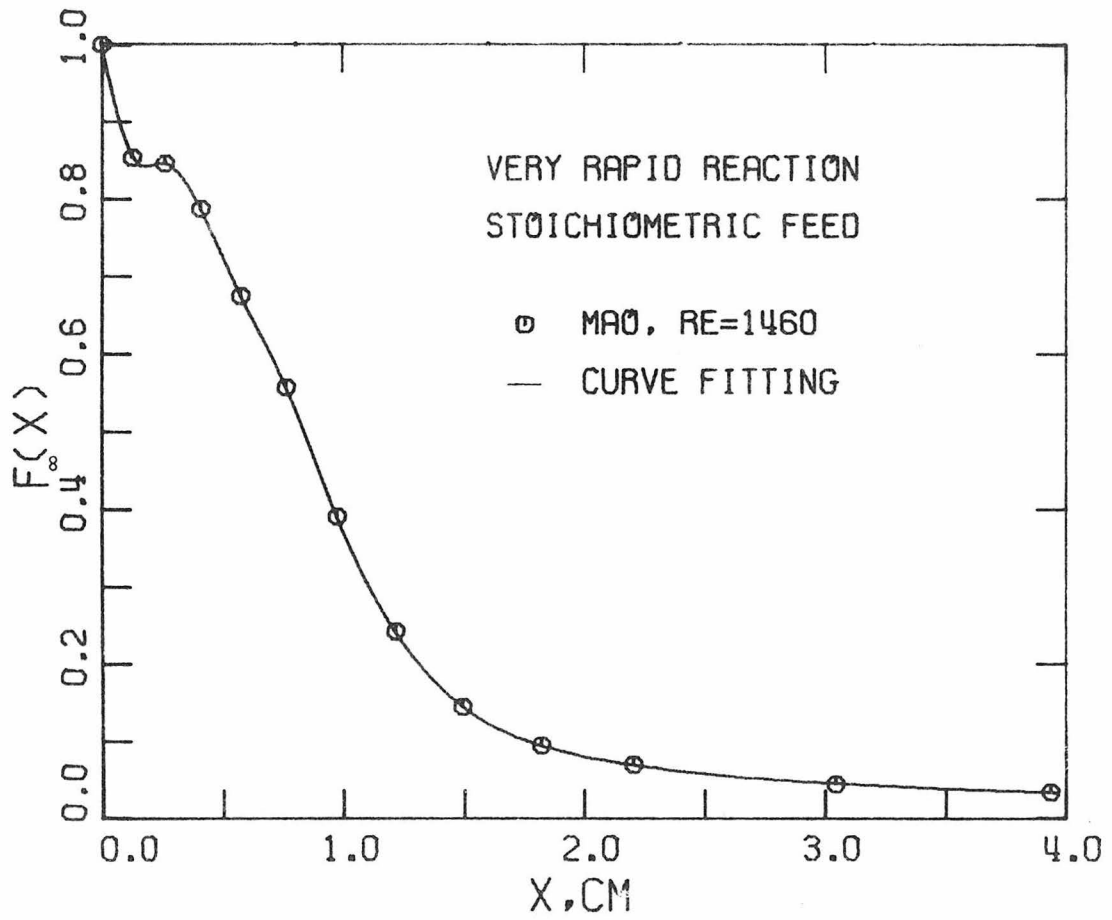


Figure A.3 Extent of reaction of a very rapid stoichiometric reaction: Data of Mao, laminar jets.

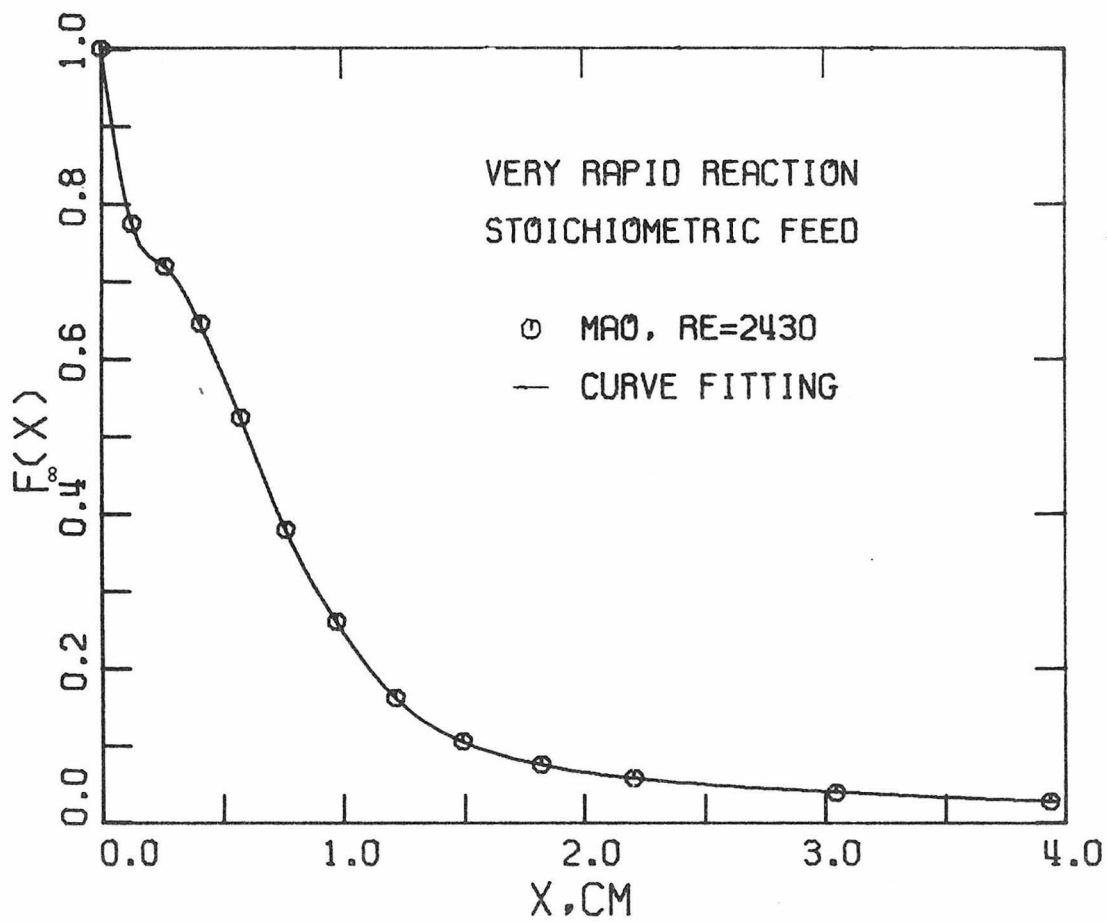


Figure A.4 Extent of reaction of a very rapid stoichiometric reaction: Data of Mao, turbulent jets.

In applying the data of $F_{\infty}(x)$ to the Dz model calculation where the independent variable is time, we have to know the axial velocity. Because of the jetting effects of the fluids coming out from the head, the velocity measured at the center of the nozzles is initially different from that measured between the nozzles (Yieh, 1970). Hence the use of a bulk-average velocity for $\langle u(x) \rangle$ will lead to considerable error. Yieh (1970) argued that in modeling Vassilatos' reactor, the velocity measured between the nozzles is more representative of the mean velocity because the area between the jets constitutes a major part of cross-sectional area of the reactor. In Mao's reactor, however, the reactor head is refined by using a larger number of nozzles, the velocity is hence more uniform. Following Yieh (1970), we shall use the velocity profile measured between the jets in our calculations. The axial velocity profiles for different flow conditions are shown in Figure A.5.

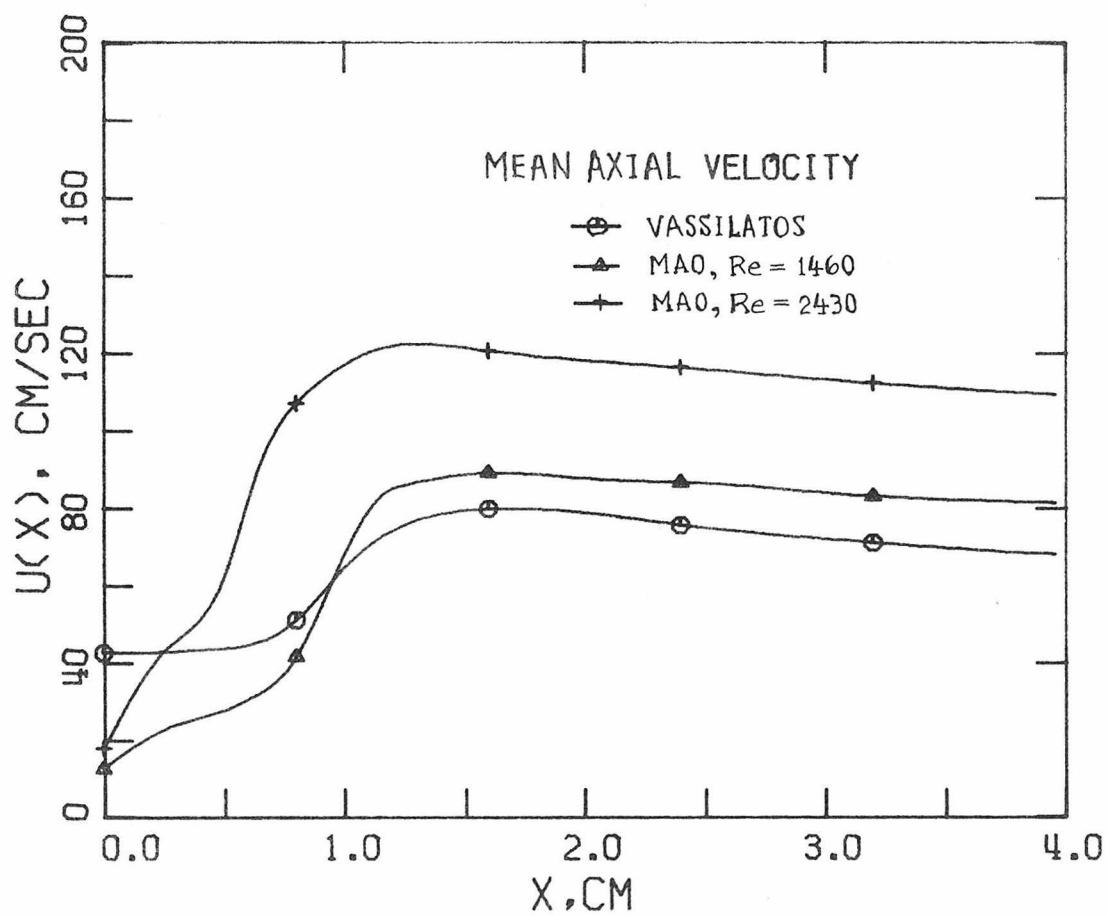


Figure A.5 Mean axial velocity profiles of the multijet reactor in three different flow conditions. Data were measured between the jets. (Yieh, 1970)

APPENDIX B. NUMERICAL CALCULATION OF REACTION PARAMETER

In this appendix the numerical method that was employed in calculating the reaction parameter λ will be discussed. We will also show that the minimum value of λ is approximately independent of the choice of initial conditions for the model equations (2.25-2.31).

Three basic steps involved in calculating λ are:

- 1) Solve (2.32-33) numerically for a^* and b^* ,
- 2) Determine the one-dimensional diffusion-zone L_{AB} using the profiles of a_I^* and b_I^* given in (2.39-40),
- 3) Calculate λ by (2.41).

Numerical solutions to a system like (2.32-33) have been studied previously by Brian et al (1961) and Mao and Toor (1970). Both of these calculations involved explicit schemes and a small time-step had to be used to insure numerical stability. In this work the quasilinearization scheme of Lee (1968) is used since it allows a larger time-step to cut down the amount of computation time. This scheme uses O'Brien-Hyman-Kaplan implicit scheme for integration and Newton-Ralphson interaction scheme for linearization and is briefly described below.

Let $a_{r,s}^n, b_{r,s}^n$ denote the concentrations a^*, b^* , respectively, at $t^*=s\Delta t, x^*=r\Delta x$ for the n -th iteration. We represent (2.32-33) by

$$\frac{a_{r,s+1}^{n+1} - a_{r,s}^n}{\Delta t} = \frac{a_{r+1,s+1}^{n+1} - 2a_{r,s+1}^{n+1} + a_{r-1,s+1}^{n+1}}{\Delta x^2} - \alpha\beta a_{r,s+1}^{n+1} b_{r,s+1}^n$$

$$\frac{b_{r,s+1}^{n+1} - b_{r,s}^n}{\Delta t} = \gamma \frac{b_{r+1,s+1}^{n+1} - 2b_{r,s+1}^{n+1} + b_{r-1,s+1}^{n+1}}{\Delta x^2} - \alpha b_{r,s+1}^{n+1} a_{r,s+1}^n$$

for $r= 1,2, \dots, N-1$

where N is the number of x -increments and $a_{r,s}, b_{r,s}$ denote known concentrations of the previous time step. The x -increment increases with t such that the boundary conditions

$$a_{0,s}^n = b_{0,s}^n = a_{N,s}^n = b_{N,s}^n = 0, \text{ for all } n \text{ and } s$$

are always maintained. To proceed we first let $a_{r,s+1}^1 = a_{r,s}, b_{r,s+1}^1 = b_{r,s}$ and solve for $a_{r,s+1}^2, b_{r,s+1}^2$ using Thomas method for inverting a tri-diagonal matrix. The iteration continues by using $a_{r,s+1}^n, b_{r,s+1}^n$ to calculate $a_{r,s+1}^{n+1}, b_{r,s+1}^{n+1}$ until $|a_{r,s+1}^{n+1} - a_{r,s+1}^n| \leq \epsilon$ and $|b_{r,s+1}^{n+1} - b_{r,s+1}^n| \leq \epsilon$, where ϵ is any desired accuracy.

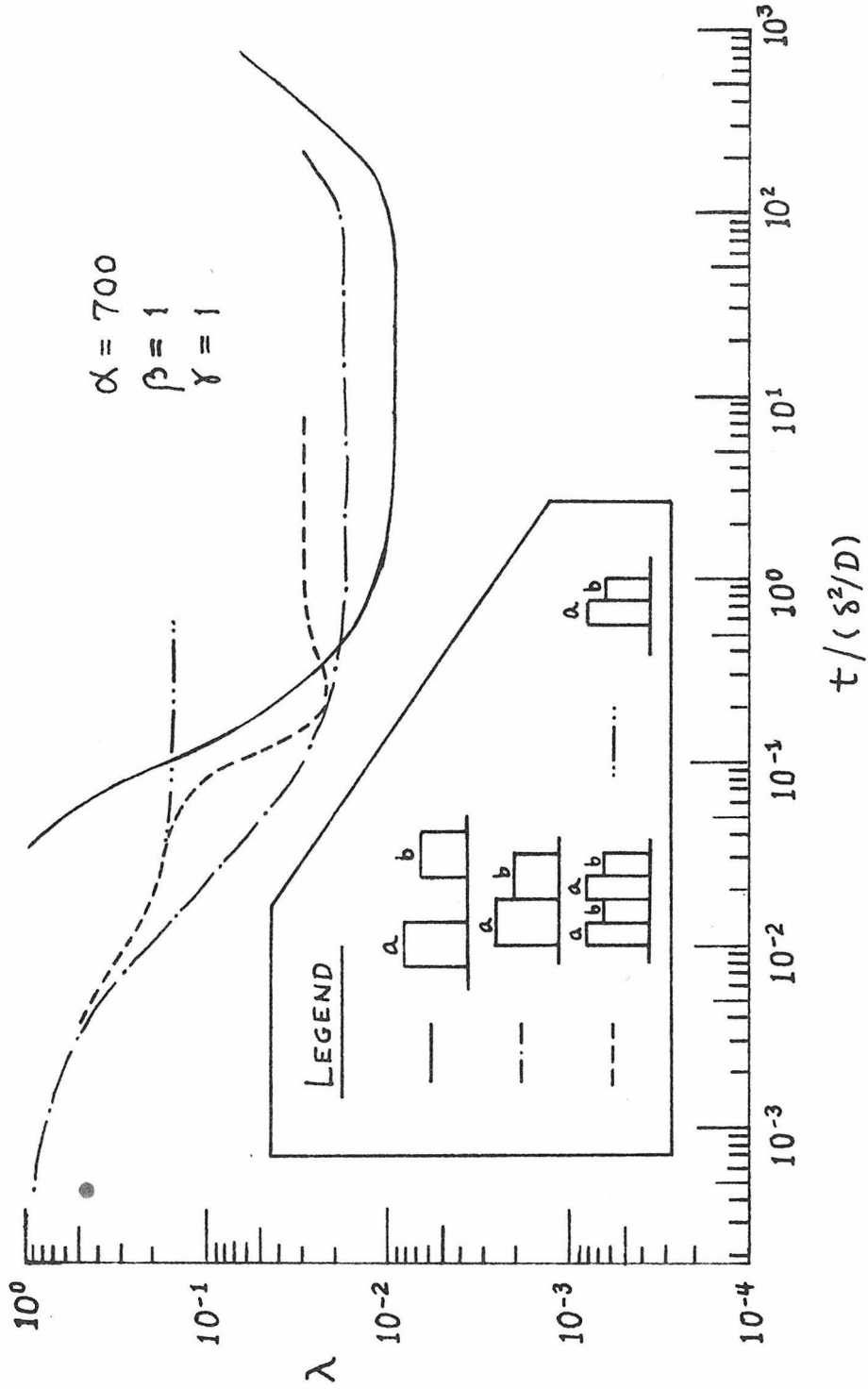
The boundaries of the diffusion-zone L_{AB} , in principle, in determined by where either a_I^* or b_I^* assumes a zero value, however, mathematical expressions (2.39-40) of a_I^* and b_I^* go to zero only in the limit of $x^* \rightarrow \pm \infty$. Hence in practice one has to determine approximate boundaries for L_{AB} . Friedlander and Keller (1963), in treating a similar problem, assumed the boundaries as where $|da_I^*/dx^*| \sim 0.05$. We found, however, to our satisfaction that L_{AB} can be simply defined as where

$$a_I^* \equiv \frac{a_I}{a_0} \geq c, \quad b_I^* \geq c/\beta$$

where ca_0 corresponds to some lower limit of a concentration measuring unit. Tests indicated that the minimum value of λ varies very little for $c \leq 0.01$. A value of $c=0.001$ was used in our calculation.

In solving (2.32-33), the initial conditions (2.36-37) have been used. However, it was discovered that the value of interest, namely λ_{\min} , only depends weakly on the choice of initial conditions. In Figure

B.1 λ vs. t^* with $\alpha=700$, $\beta= \gamma= 1$ is plotted for three different initial conditions. The solid curve is the result of γ using (2.36-37) as initial condition. If initially the two pulses are next to each other, γ decays earlier from unity as indicated by the dot-dashed curve. Lastly, the pulses are broken in half and alternately arranged. The dashed curve of λ first decays to some pseudo λ_{\min} which corresponds to the value λ_{\min} would assume if the two pairs of pulses were far apart, or only one of them is present. (Since the pulse width in the last case is half of the original, the pseudominimum of λ is the λ_{\min} for the case where $\alpha= 700/4$.) As the pulses gets wider, the interaction of neighboring pairs brings λ to its true minimum. The important aspect of these calculations is that the initial condition influences the behavior of λ for $t \leq \delta^2/D$, but only weakly affects the value of λ_{\min} . This supports our assumption in Chapter II that λ_{\min} is a function of α only.



• Figure B.1 Temporal behavior of the reaction parameter λ for various hypothetical initial concentration distributions. The minimum value of λ is only weakly dependent of the choice of initial concentrations.

APPENDIX C. ESTIMATION OF NO-EMISSION FOR A POWER PLANT

This appendix describes how we estimated the NO-emission for Morgantown power plant. Efforts made to obtain direct measurements of NO-emission for the plant were in vain. We therefore use the following methods to estimate the NO-emission:

1. According to a Morgantown plant engineer, the plant consumes 225 tons/hr of coal and 40.5 tons/hr of fuel oil currently (October 1975). Using a power rating of 1000 Mw and typical values of heat of combustion, 13000 Btu/lb of coal and 18000 Btu/lb of fuel oil, we calculate the plant efficiency to be 23.4%. During the time of flight observations (Summer 1974), the plant burned a fuel mixture of 25% coal and 75% oil. Thus 109 tons/hr of coal and 327 tons/hr of oil were required. The plant engineer estimated a 1.9% sulfur content for both fuel oil and coal. If the burning was complete the estimated SO₂ - emission is 16.5 tons/hr. D. D. Davis (private communication) reported that the concentration of NO was always about half of that of SO₂ during the flights. We therefore estimate that NO was emitted at 16500 lb/hr.

2. R. C. Flagan (private communication) indicated that power plant NO-emission is about 0.7-1.2 lb (as NO₂) per Mega Btu (Thermal). Using the estimated efficiency of 23.4%, Morgantown plant inputs 4 Mega Btu/sec. The NO-emission is hence 10000-17000 lb (as NO₂) /hr or 6500-11000 lb (as NO)/hr. The figures, although still high, appear to be more reasonable. (For reference, California power plants output NO_x ranging from 60 to 2000 lb/hr while the current goal is 15 lb/hr for 1976.)

APPENDIX D. CONSIDERATIONS OF MULTIPLE KINETICS

In many reaction systems, a chemical species of interest is often involved in reactions with more than one species. The application of the diffusion zone model to such systems is discussed here. We assume that the chemical kinetics of the reaction system (often determined by well-mixed reactor data) are given. The problem is to apply the diffusion zone model to studies of the effects of inhomogeneous mixing on the rates of turbulent chemical reactions in complex systems.

Consider a system of N chemical species: A_1, A_2, \dots, A_N . By combination, each species could be involved in a maximum of N bimolecular chemical reactions. That is, for the i-th species A_i the N reactions could be



for $i = 1, \dots, N$. Where in (D.1), k_{ij} is the rate constant and α_ℓ is the stoichiometric coefficient of the reaction. If A_i is inert to A_j k_{ij} is zero, and if A_ℓ is not a product species of the i-j reaction α_ℓ assumes a value of zero. Let A_i also denote the concentration of species, the instantaneous local rate of change of A_i can be written as

$$\frac{dA_i}{dt} = - \sum_{j=1}^N k_{ij} A_i A_j + R_i \quad (D.2)$$

The term R_i in (D.2) denotes all other kinetic terms influencing the decay of A_i , e.g., production of A_i by other species and decay of A_i due to

reactions of orders other than two. The average concentration of A_i is defined as

$$\bar{A}_i = \frac{1}{V} \int_V A_i dv \quad (D.3)$$

The rate equation of A_i is then

$$\frac{d\bar{A}_i}{dt} = - \sum_{j=1}^N k_{ij} \bar{A}_i \bar{A}_j + \bar{R}_i \quad (D.4)$$

Assuming the diffusion zone v_{ij} is defined as the region of space where A_i and A_j coexist, (to be further discussed) we can write

$$\bar{A}_i = \frac{1}{V} (v_i \hat{A}_i + v_{ij} \hat{a}_i)$$

where v_i is the region where A_i exists excluding v_{ij} , and \hat{A}_i and \hat{a}_i are average concentrations in v_i and v_{ij} respectively. Similarly, letting the subscript I denote inert concentrations, we have

$$\bar{A}_{iI} = \frac{1}{V} (v_i \hat{A}_{iI} + v_{ij} \hat{a}_{iI})$$

Following through similar arguments as in the case of a single reaction, we arrive at

$$\begin{aligned} \frac{d\bar{A}_i}{dt} = - \sum_{j=1}^N \frac{k_{ij} \lambda_{ij} \Gamma_{ij}}{\zeta_{ij} \zeta_{ji}} [\bar{A}_i - (1-\zeta_{ij})\bar{A}_{iI}] [\bar{A}_j - (1-\zeta_{ji})\bar{A}_{jI}] \\ + \bar{R}_i \end{aligned} \quad (D.5)$$

where

$$\Gamma_{ij} = \frac{\overline{A_{iI} A_{jI}}}{\overline{A_{iI}} \overline{A_{jI}}} \quad \lambda_{ij} = \frac{\mu_{ij}}{\mu_{ijI}}$$

$$\mu_{ij} = \left[\frac{1}{v_{ij}} \int_{v_{ij}} a_i a_j dv \right] / \hat{a}_i \hat{a}_j$$

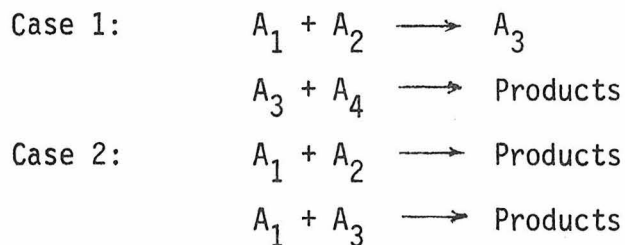
$$\mu_{ijI} = \left[\frac{1}{v_{ij}} \int_{v_{ij}} a_{iI} a_{jI} dv \right] / \hat{a}_{iI} \hat{a}_{jI}$$

$$\zeta_{ij} = \frac{v_{ij} \hat{a}_{iI}}{V \overline{A_{iI}}}$$

$$\zeta_{ji} = \frac{v_{ij} \hat{a}_{jI}}{V \overline{A_{jI}}}$$

Although (D.5) gives us a general relationship for the decay of concentration of a species involving in N second-order reactions and other reactions, the parameters contained in the equation are rather difficult to estimate. The difficulty arises mainly because the diffusion zone is not as well-defined as in the case of a single reaction.

To elucidate the problem, let us consider two special cases:



In the first case, the diffusion zone v_{34} is defined as the region of

space where A_3 and A_4 would coexist if they were inert; but since A_3 is a product species such a region of space is not defined in the inert mixing problem. In the second case, the application of the diffusion zone model is hindered by the lack of a generalized model of the reaction parameter λ . The problem here is that the diffusion zones v_{12} and v_{13} share the volume of space v_{123} and consequently the parameter λ_{12} and λ_{13} are not necessarily describable in terms of the material distributions in v_{12} and v_{13} as assumed by the model of λ that we developed in Section II.2.2.

It is easy to image other situations where the diffusion zone is not well-defined. We hence conclude that the application of diffusion zone model to complex reaction systems depends on the generalization of the diffusion zone concept. For example, we may note that in the first case the product species A_3 can exist only in v_{12} . Thus, as a first approximation, one might assume that

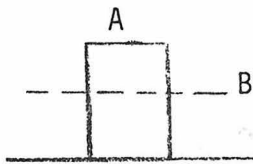
$$v_{34} = v_{124}$$

where v_{124} is the region of space where A_1, A_2, A_4 coexist if they were inert under the same mixing condition. In general, however, the problem remains to be a challenge to the future workers.

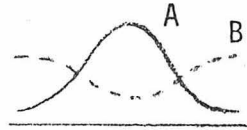
A somewhat similar problem arises in treating reaction systems where species are released at different times. One example is the emission of fresh reactive species into an atmosphere laden with both product and reactant species. In cases where ambient fluid is void of the emitted

species, say A , then the diffusion zone model may be applied in the manner developed in this work. However, if the emitted species A is already present in the background fluid, then a problem arises in the description of the reaction parameter λ . For example, the fluid into which the fresh species is emitted can possess diffusion zones in any of the following three stages:

(a) Initial period,
 $\lambda = 1$



(b) Reaction period,
 $\lambda = \lambda_{\min}$



(c) Completion period,
 $\lambda = 1$



Here A and B denote ambient species concentrations. As we showed in Ch. II, in cases where reactive species are released only at some initial instant t_0 , the parameter λ has a value λ_{\min} during most of the period of reaction. It was based on this observation that an expression for λ was developed. Needless to say, the model of λ is not necessarily valid if reactants are injected into the system at some time after t_0 . In view of the difficulty of developing a λ -model applicable to all possible conditions of reactant releases, it is suggested that the diffusion zone model be used in the cases where the ambient fluid is well-mixed at each time that fresh species are introduced. Inasmuch as the class of reactions that we have considered in this work are prevalent in both engineering and atmospheric pollution studies, the diffusion zone model developed here should find widespread uses.

REFERENCES

- Ajmera, P. V., Singh, M., Toor, H. L. 1975. Chem. Eng. Commun.
Submitted for publication.
- Batchelor, G. K. 1949. Aust. J. Sci. Res. 2:437-50
- 1950. Quart. J. Roy. Met. Soc. 76:133-46
- 1959. J. Fl. Mech. 5:113-33
- Brian, P. L. T., Hurley, J. F., Hasseltine, E. H. 1961. AICHE J. 7:226
- Borghi, R. 1974. Adv. Geophys. 18B:349-65
- Brodkey, R. S. 1974. Turbulent Mixing in Non-Reactive and Reactive Flows,
ed. S. N. B. Murthy, 225-26. Project SQUID Workshop, Purdue Univ.,
West Lafayette, Indiana, May 1974 (To be published). 461 pp.
- Burgers, J. M. 1948. Adv. Appl. Math. 1:171-99
- Bush, W. B., Fendell, F. E. 1974. Acta Astronaut. 1:645-66
- Carslaw, H. S., Jaeger, J. C. 1959. Conduction of Heat in Solids.
Oxford, 510 pp.
- Chung, P. M. 1969. AIAA J. 7:1982-91
- 1970. Phys. Fluids 13:1153-65
- 1971. AIAA J. 9:2282-84
- 1972. Phys. Fluids 15:1753-46
- Clarke, R. H., Dyer, A. J., Brook, R. R., Reid, D. G., Troop, A. J.
1971. Boundary Layer Data. CSIRO Div. of Meteorol. Phys.,
Tech. Paper No. 19, 340 pp.

- Corrsin, S. 1958. Phys. Fluids 1: 42-47
- 1961. J. Fluid Mech. 11:407-16
- 1962 . Proc. 1961 Symp. on Fluid Dyn. and Appl. Math., ed.
J. B. Diaz, S. I. Pai, 105-24. New York: Gordon and Breach.
- 1964. Phys. Fluids 7: 1156-59, 8:1920
- Davis, D. D., Klauber, G. 1975. Proc. Symp. on Chem. Kinetics Data
for the Lower and Upper Atm. Warrenton, Va., Sep. 1974
(in International J. Chem. Kinetics 1975)
- Davis, D. D., Smith, G., Klauber, G. 1974. Science Nov. 22, 1974.
186: 733-36
- Deardorff, J. W. 1970a. J. Fluid Mech. 41:453-80
- 1970b. J. Atm. Sci. 27:1211-13
- 1972. J. Atm. Sci. 29:91-115
- 1973. J. Fluid Eng. Sep. 1973, 429-38
- 1974a. Boundary Layer Meteorol. 7: 81-106
- 1974b. Boundary Layer Meteorol. 7:199-226
- , Drake, M. A. 1975. Boundary Layer Data from a Numerical Integration
in Three Dimensions. NCAR report.
- , Peskin, R. L. 1970. Phys, Fluids 13:584-95
- , Willis, G. E. 1974. Adv. Geophys 18B:187-200
- Donaldson, C. du P., Hilst, G. R. 1972a. Environ. Sci. Technol. 6:812-16
- 1972b. Proc. 1972 Hear Transfer Fluid Mech. Inst., ed. R. B. Landis,
G. J. Hordemann, 253-61
- Donaldson, C. du P., Rosenbaum, H. 1969. Compresible Turb. Boundary
Layers, 231-53. NASA SP-216.

- Dopazo, C., O'Brien, E. E. 1973. Phys. Fluids 16: 2075-81
- Evangelista, J. J., Katz, S., Shinnar, R. 1969 . AIChE J. 15: 843-53
- Flagan, R. C., Appleton, J. P. 1974. Combust. Flame 23: 249-67
- Friedlander, S. K., Keller, K. H. 1963. Chem. Eng. Sci. 18:365-75
- Gifford, F., Jr. 1957a. J. of Meteorol. 14: 410-14
- 1957b. J. of Metheorol. 14: 475-76
- Glassman, I., Eberstein, I. J. 1963. AIAA J. 1: 1424-26
- Hawthorne, W. R., Weddell, D. S., Hottel, H. C. 1949. 3rd Symp. Combust.,
Flame, Explos, Phenom., 266-88. Baltimore: Williams and Wilkins.
- Hill, J. C. 1970. Phys. Fluids 13: 1394-96
- Hill, J. C. 1976. Homogeneous Turbulent Mixing with Chemical Reaction.
Annual Rev. Fluid Mech. Volume 8
- Hilst, G. R., Donaldson, C. du P., Teske, M., Contiliano, R., Freiberg, J.
1973. A Coupled Two-Dimensional Diffusion and Chemistry Model for
Turbulent and Inhomog. Mixed Reaction Systems. U.S. Environ-
mental Protection Agency EPA-R4-73-016c.(PB-234193/1G1)
- Hinze, J. O. 1959. Turbulence. McGraw-Hill, New York. 586 pp.
- Johnston, H.S., Crosby, J. H. 1954. J. Chem. Phys. 22: 689-92
- Kattan, A., Adler, R. J. 1967. AIChE J. 13: 580-85
- 1972. Chem. Eng. Sci. 27: 1013-28
- Keeler, R. N., Peterson, E. E., Prausnitz, J. M. 1965. AIChE J. 11: 221-27
- Kraichnan, R. H. 1970. Phys. Fluids 13:22-31
- Lamb, R. G. 1972a. Stat. of Marked Particle Conc. in Turb. Fl., Part I:
General Theory; Unpublished Notes.
- 1972b. Stat. of Marked Particle Conc. in Turb. Fl., Part II:
The Case of Chemically Reactive Particles: Unpublished Notes.

- Lamb, R. G. 1973. Atmos. Environ. 7: 257-63
- , Chen, W. H., Seinfeld, J. H. 1975. Numerico-Empirical Analyses of Atmospheric Diffusion Theories. J. Atm. Sci. 32:1794-1807.
- Lee, J. 1966 Phys. Fluids 9: 363-72
- 1973. Quart. Appl. Math. 31: 155-76
- Lee, E. S. 1968. Quasilinearization and Invariant Imbedding.
Academic Press, New York.
- Libby, P. A. 1972. Combust. Sci. Technol. 6: 23-28
- Lin, C. -H. 1971. Stochastic Chemical Reaction with Molecular Diffusion.
MS theses. SUNY, Stony Brook, New York. 73 pp.
- , O'Brien, E. E. 1972. Astronaut. Acta 17: 771-81
- Lindberg, W. R., Thompson, R. O. R. Y. 1974. Quart. J. Roy. Met. Soc.
100: 608-23
- Mao, K. W. 1969. Chemical Reactions with Turbulent Mixing. PhD thesis.
Carnegie-Mellon Univ.
- , Toor, H. L. 1970. AIChE J. 16: 49-52
- , Toor, H. L. 1971. Ind. Eng. Chem. Fundam. 10: 192-97
- McCarthy, H. E. 1970. Second-Order Chemical Reactions with Turbulent
Mixing. PhD thesis. Univ. of Delaware, Newark, Delaware. 168 pp.
- McKelvey, K. N. 1968. Turbulent Mixing with Chemical Reaction. PhD thesis.
Ohio State Univ., Columbus, Ohio. 123 pp.
- Miyairi, Y., Kamiwano, M., Yamamoto, K. 1971. Int. Chem. Eng. 11: 344-52
- Miyawaki, O., Tsujikawa, H., Uraguchi, Y. 1974. J. Chem. Eng. Jap.
7: 52-56

- Monin, A. S., Yaglom, A. M. 1971. Statistical Fluid Mechanics. MIT Press, Cambridge. 769 pp.
- O'Brien, E. E. 1966. Phys. Fluids 9: 1561-65
- 1968. Phys. Fluids 11: 1883-88
- 1971. Phys. Fluids 14: 1326-31
- 1974. Advan. Geophys. 18B: 341-48
- , Eng, R. M. 1970. Phys Fluids 13: 1393-94
- , Francis, G. C. 1962. J. Fluid Mech. 13: 369-82
- , Lin, C. -H. 1972. Phys Fluids 15: 931-33
- Orszag, S. A., Patterson, G. S., Jr., 1972. Phys. Rev. Lett. 28: 76-79
- Pao, Y. H. 1964. AIAA J. 2: 1550-59
- Patterson, G. S., Jr., Corrsin, S. 1966. in Dynamics of Fluids and Plasmas. S.I. Pai, ed. Academic Press, New York. p.275
- Rao, D. P., Dunn, I. J. 1970. Chem. Eng. Sci. 25: 1275-82
- , Edwards, L. L. 1971. AIChE J. 17: 1264-65
- Riley, J. J. 1973. Phys. Fluids 16: 1160-61
- , Corrsin, S. 1971. Preprint, Conf. on Air Poll. Meteorology, Rayleigh, N. C., p.16.
- , Patterson, G. S., Jr. 1974. Phys Fluid 17: 292-97
- Romberg, G. 1974. Int. J. Heat Mass Transfer 17: 1163-79
- Seinfeld, J. H. 1975. Air Pollution: Physical and Chemical Fundamentals. McGraw-Hill, New York. 523 pp.

- Shea, J. R., III, 1976. A Chemical Reaction in A Turbulent Jet.
PhD thesis. Calif. Inst. of Tech., Pasadena, Calif.
(Submitted July 1, 1975)
- Tennekes, H., Lumley, J. L. 1972. A First Course in Turbulence.
MIT Press, Cambridge.
- Thompson, R. O. R. Y. 1971. Quart. J. Roy. Met. 97: 93-98
- Toor, H. L. 1962. AICHE J. 8: 70-78
- , 1969. Ind. Eng. Chem. Fundam. 8: 655-59
- , Singh, M. 1973. Ind. Eng. Chem. Fundam. 12: 448-51
- Forrest, R. S., Ranz, W. E. 1970. AICHE J. 16: 930-42
- Uspensky, J. V. 1937. Introduction to Mathematic Probability.
McGraw-Hill, New York. p.265
- Vassilatos, G., Toor, H. L. 1965. AICHE J. 11: 666-73
- Willis, G. E., Deardorff, J. W. 1975. A Laboratory Model of Diffusion
into the Convective Planetary Boundary Layer. Submitted for
Publication.
- Yieh, H. -N. 1970. Turbulent Mixing with Chemical Reaction. PhD thesis.
Ohio State Univ., Columbus, Ohio. 135 pp.
- Zakanycz, S. 1971. Turbulence and the Mixing of Binary Gases. PhD
thesis. Ohio State Univ., Columbus, Ohio, 172 pp.