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1 Introduction

The idea that the dynamics of a fire might be studied numerically dates back to the beginning of the computer age. Indeed, the fundamental conservation equations governing fluid dynamics, heat transfer, and combustion were first written down over a century ago. Despite this, practical mathematical models of fire (as distinct from controlled combustion) are relatively recent due to the inherent complexity of the problem. Indeed, in his brief history of the early days of fire research, Hoyt Hottel noted “A case can be made for fire being, next to the life processes, the most complex of phenomena to understand” [1].

The difficulties revolve about three issues: First, there are an enormous number of possible fire scenarios to consider due to their accidental nature. Second, the physical insight and computing power necessary to perform all the necessary calculations for most fire scenarios are limited. Any fundamentally based study of fires must consider at least some aspects of bluff body aerodynamics, multi-phase flow, turbulent mixing and combustion, radiative transport, and conjugate heat transfer; all of which are active research areas in their own right. Finally, the “fuel” in most fires was never intended as such. Thus, the mathematical models and the data needed to characterize the degradation of the condensed phase materials that supply the fuel may not be available. Indeed, the mathematical modeling of the physical and chemical transformations of real materials as they burn is still in its infancy.

In order to make progress, the questions that are asked have to be greatly simplified. To begin with, instead of seeking a methodology that can be applied to all fire problems, we begin by looking at a few scenarios that seem to be most amenable to analysis. Hopefully, the methods developed to study these “simple” problems can be generalized over time so that more complex scenarios can be analyzed. Second, we must learn to live with idealized descriptions of fires and approximate solutions to our idealized equations. Finally, the methods should be capable of systematic improvement. As our physical insight and computing power grow more powerful, the methods of analysis can grow with them.

To date, three distinct approaches to the simulation of fires have emerged. Each of these treats the fire as an inherently three dimensional process evolving in time. The first to reach maturity, the “zone” models, describe compartment fires. Each compartment is divided into two spatially homogeneous volumes, a hot upper layer and a cool lower layer. Mass and energy balances are enforced for each layer, with additional models describing other physical processes appended as differential or algebraic equations as appropriate. Examples of such phenomena include fire plumes, flows through doors, windows and other vents, radiative and convective heat transfer, and solid fuel pyrolysis. An excellent description of the physical and mathematical assumptions behind the zone modeling concept is given by Quintiere [2], who chronicles developments through 1983. Model development since then has progressed to the point where documented and supported software implementing these models are widely available [3].

The relative physical and computational simplicity of the zone models has led to their widespread use in the analysis of fire scenarios. So long as detailed spatial distributions of physical properties are not required, and the two layer description reasonably approximates reality, these models are quite reliable. However, by their very nature, there is no way to systematically improve them. The rapid growth of computing power and the corresponding maturing of computational fluid dynamics (CFD), has led to the development of CFD based “field” models applied to fire research problems. Virtually all this work is based on the conceptual framework provided by the Reynolds-averaged form of the governing equations, in particular the $k – \varepsilon$ turbulence model pioneered by Patankar and Spalding [4]. The use of CFD models has allowed the description of fires in complex geometries, and the incorporation of a wide variety of physical phenomena. However, these models have a fundamental limitation for fire applications – the averaging procedure at the root of the model equations. The $k – \varepsilon$ model was developed as a time-averaged approximation to the conservation equations of fluid dynamics. While the precise nature of the averaging time is not specified, it is clearly long enough to require the introduction of large eddy transport coefficients to describe the unresolved fluxes of mass, momentum and energy. This is the root cause of the smoothed appearance of the results of
even the most highly resolved fire simulations. The smallest resolvable length scales are determined by
the product of the local velocity and the averaging time, rather than the spatial resolution of the underlying
computational grid. This property of the $k - \varepsilon$ model is typically exploited in numerical computations by
using implicit numerical techniques to take large time steps.

Unfortunately, the evolution of large eddy structures characteristic of most fire plumes is lost with
such an approach, as is the prediction of local transient events. It is sometimes argued that the averaging
process used to define the equations is an “ensemble average” over many replicates of the same experiment
or postulated scenario. However, this is a moot point in fire research since neither experiments nor real
scenarios are replicated in the sense required by that interpretation of the equations. The application of
“Large Eddy Simulation” (LES) techniques to fire is aimed at extracting greater temporal and spatial fidelity
from simulations of fire performed on the more finely meshed grids allowed by ever faster computers. The
phrase LES refers to the description of turbulent mixing of the gaseous fuel and combustion products with
the local atmosphere surrounding the fire. This process, which determines the burning rate in most fires and
controls the spread of smoke and hot gases, is extremely difficult to predict accurately. This is true not only
in fire research but in almost all phenomena involving turbulent fluid motion. The basic idea behind the
LES technique is that the eddies that account for most of the mixing are large enough to be calculated with
reasonable accuracy from the equations of fluid dynamics. The hope (which must ultimately be justified by
appeal to experiments) is that small-scale eddy motion can either be crudely accounted for or ignored.

The equations describing the transport of mass, momentum, and energy by the fire induced flows must
be simplified so that they can be efficiently solved for the fire scenarios of interest. The general equations of
fluid dynamics describe a rich variety of physical processes, many of which have nothing to do with fires.
Retaining this generality would lead to an enormously complex computational task that would shed very
little additional insight on fire dynamics. The simplified equations, developed by Rehm and Baum [5], have
been widely adopted by the larger combustion research community, where they are referred to as the “low
Mach number” combustion equations. They describe the low speed motion of a gas driven by chemical heat
release and buoyancy forces.

The low Mach number equations are solved numerically by dividing the physical space where the fire
is to be simulated into a large number of rectangular cells. Within each cell the gas velocity, temperature,
etc., are assumed to be uniform; changing only with time. The accuracy with which the fire dynamics can
be simulated depends on the number of cells that can be incorporated into the simulation. This number
is ultimately limited by the computing power available. Present day desktop computers limit the number
of such cells to at most a few million. This means that the ratio of largest to smallest eddy length scales
that can be resolved by the computation (the “dynamic range” of the simulation) is roughly 100 $\sim$ 200.
Unfortunately, the range of length scales that need to be accounted for if all relevant fire processes are to be
simulated is roughly $10^4 \sim 10^5$ because combustion processes take place at length scales of 1 mm or less,
while the length scales associated with building fires are of the order of meters or tens of meters. The form
of the numerical equations discussed below depends on which end of the spectrum one wants to capture
directly, and which end is to be ignored or approximated.
2 Hydrodynamic Model

An approximate form of the Navier-Stokes equations appropriate for low Mach number applications is used in the model. The approximation involves the filtering out of acoustic waves while allowing for large variations in temperature and density [5]. This gives the equations an elliptic character, consistent with low speed, thermal convective processes. The computation can either be treated as a Direct Numerical Simulation (DNS), in which the dissipative terms are computed directly, or as a Large Eddy Simulation (LES), in which the large-scale eddies are computed directly and the sub-grid scale dissipative processes are modeled. The choice of DNS vs. LES depends on the objective of the calculation and the resolution of the computational grid. If, for example, the problem is to simulate the flow of smoke through a large, multi-room enclosure, it is not possible to resolve the combustion and transport processes directly. However, for small-scale combustion experiments, it is possible to compute the transport directly and the combustion processes to some extent.

2.1 Conservation Equations

First, consider the conservation equations of mass, momentum and energy for a thermally-expandable, multi-component mixture of ideal gases [5]:

Conservation of Mass
\[ \frac{\partial \rho}{\partial t} + \nabla \cdot (\rho \mathbf{u}) = 0 \]  

Conservation of Species
\[ \frac{\partial}{\partial t}(\rho Y_i) + \nabla \cdot (\rho \mathbf{u} Y_i) = \nabla \cdot (\rho D_i) + \nabla Y_i + W_i'''' \]  

Conservation of Momentum
\[ \rho \left( \frac{\partial \mathbf{u}}{\partial t} + (\mathbf{u} \cdot \nabla) \mathbf{u} \right) + \nabla p = \rho \mathbf{g} + \mathbf{f} + \nabla \cdot \mathbf{\tau} \]  

Conservation of Energy
\[ \frac{\partial}{\partial t}(\rho h) + \nabla \cdot \rho h \mathbf{u} - \frac{D\rho}{Dt} = q'''' - \nabla \cdot \mathbf{q}_r + \nabla \cdot k \nabla T + \nabla \cdot \sum_l h_l (\rho D_l) \nabla Y_l \]  

Note that the external force on the fluid, represented by the term \( \mathbf{f} \) in Eq. (3), consists of the drag exerted by water droplets emanating from sprinklers. The energy driving the system is represented by the heat release rate \( q'''' \) in Eq. (4). The term \( D\rho/Dt = \partial \rho/\partial t + \mathbf{u} \cdot \nabla \rho \) is a material derivative. All other symbols are listed in the Nomenclature (Section 9).

2.2 State, Mass and Energy Equations

The conservation equations are supplemented by an equation of state relating the thermodynamic quantities density, pressure and enthalpy; \( \rho, p \) and \( h \). The pressure is decomposed into three components
\[ p = p_0 - \rho \omega g z + \tilde{p} \]  

The first term on the right hand side is the “background” pressure, the second is the hydrostatic contribution, and the third is the flow-induced perturbation pressure. For most applications, \( p_0 \) is constant. However, if
the enclosure is tightly sealed, \( p_0 \) is allowed to increase (or decrease) with time as the pressure within the enclosure rises due to thermal expansion or falls due to forced ventilation. Also, if the height of the domain is on the order of a kilometer, \( p_0 \) can no longer be assumed constant and must be considered a function of the altitude [6].

The purpose of decomposing the pressure is that for low-Mach number flows, it can be assumed that the temperature and density are inversely proportional, and thus the equation of state can be approximated [5]

\[
p_0 = \rho T \mathcal{R} \sum (Y_i/M_i) = \rho T \mathcal{R} / M
\]  

The pressure \( p \) in the state and energy equations is replaced by the background pressure \( p_0 \) to filter out sound waves that travel at speeds that are much faster than typical flow speeds expected in fire applications. The low Mach number assumption serves two purposes. First, the filtering of acoustic waves means that the time step in the numerical algorithm is bound only by the flow speed as opposed to the speed of sound, and second, the modified state equation leads to a reduction in the number of dependent variables in the system of equations by one. The energy equation (4) is never explicitly solved, but its source terms are included in the expression for the flow divergence, an important quantity in the analysis to follow.

A further assumption about the thermodynamic variables is that the constant-pressure specific heat of the \( i \)th species \( c_{p,i} \) is assumed to be independent of temperature. Under this assumption, the enthalpy can be written as:

\[
h = \sum T \sum c_{p,i} Y_i = T \sum c_{p,i} Y_i
\]  

The specific heat for each species can be expressed in terms of the number of internal degrees of freedom \( \nu_i \) active in that molecule.

\[
c_{p,i} = \left( \frac{2 + \nu_i}{2} \right) \frac{\mathcal{R}}{M_i} = \left( \frac{\gamma_i}{\gamma_i - 1} \right) \frac{\mathcal{R}}{M_i}
\]  

If the ratio of specific heats \( \gamma_i \) for each species is assumed to be that of a diatomic molecule (\( \nu = 5, \gamma = 7/5 \)), the equation of state can be rewritten in the form\(^1\)

\[
p_0(t) = \frac{\gamma - 1}{\gamma} \rho h
\]  

Taking the material derivative of Eq. (9) and using the mass and energy conservation equations, the divergence of the velocity field, \( \nabla \cdot \mathbf{u} \), can be written in terms of the thermodynamic quantities

\[
\nabla \cdot u = \frac{\gamma - 1}{\gamma p_0} \left( q'' - \nabla \cdot q_r + \nabla \cdot k \nabla T + \nabla \cdot \sum T c_{p,i} T \rho D \nabla Y_i - \frac{1}{\gamma - 1} \frac{dp_0}{dt} \right)
\]  

If the enclosure is tightly sealed, the background pressure \( p_0 \) can no longer be assumed constant due to the increase (or decrease) in mass and thermal energy within the enclosure. The evolution equation for the pressure is found by integrating Eq. (10) over the entire domain \( \Omega \)

\[
\frac{dp_0}{dt} = \frac{\gamma - 1}{V} \left( \int q'' dV + \int \sum c_{p,i} T \rho D \nabla Y_i \cdot dS \right) - \frac{\gamma p_0}{V} \int \nabla \cdot \mathbf{u} \cdot dS
\]  

where \( V \) is the volume of the enclosure.

\(^1\)The basis of this approximation is that nitrogen will be the dominant species in most fire scenarios.
2.3 The Momentum Equation

The momentum equation is simplified by subtracting off the hydrostatic pressure gradient from the momen-
tum equation (3), and then dividing by the density to obtain

\[
\frac{\partial \mathbf{u}}{\partial t} - \mathbf{u} \times \mathbf{\omega} + \frac{1}{2} \nabla |\mathbf{u}|^2 + \frac{1}{\rho} \nabla \tilde{p} = \frac{1}{\rho} (\rho - \rho_\infty) \mathbf{g} + \mathbf{f} + \nabla \cdot \mathbf{\tau}
\]  

(12)

To simplify this equation further, a substitution is made

\[
\nabla \mathcal{H} \approx \frac{1}{2} \nabla |\mathbf{u}|^2 + \frac{1}{\rho} \nabla \bar{\rho}
\]  

(13)

The basis for this approximation is seen in the evolution equation for the circulation, obtained by integrating

\[
\frac{d \Gamma}{dt} = \oint \frac{1}{\rho} (\nabla \tilde{p} + (\rho - \rho_\infty) \mathbf{g} + \nabla \cdot \mathbf{\tau}) \cdot d\mathbf{x}
\]  

(14)

There are three sources of vorticity: the baroclinic torque due to the non-alignment of the density and
pressure gradients, buoyancy due to horizontal density gradients, and viscosity. Buoyancy is the dominant
source of vorticity, and the approximation above is equivalent to neglecting the baroclinic torque.³

Neglecting the baroclinic torque simplifies the elliptic partial differential equation obtained by taking
the divergence of the momentum equation

\[
\nabla^2 \mathcal{H} = -\left(\frac{\partial (\nabla \cdot \mathbf{u})}{\partial t} - \nabla \cdot \mathbf{F}\right) ; \quad \mathbf{F} = -\mathbf{u} \times \mathbf{\omega} - \frac{1}{\rho} (\rho - \rho_\infty) \mathbf{g} + \mathbf{f} + \nabla \cdot \mathbf{\tau}
\]  

(15)

The linear algebraic system arising from the discretization of Eq. (15) has constant coefficients and can be
solved to machine accuracy by a fast, direct (i.e. non-iterative) method that utilizes fast Fourier transforms.

No-flux or forced-flow boundary conditions are specified by asserting that

\[
\frac{\partial \mathcal{H}}{\partial n} = -F_n - \frac{\partial u_n}{\partial t}
\]  

(16)

where \(F_n\) is the normal component of \(\mathbf{F}\) at the vent or solid wall, and \(\partial u_n/\partial t\) is the prescribed rate of change
in the normal component of velocity at a forced vent. Initially, the velocity is zero everywhere. At open
external boundaries the pressure-like term \(\mathcal{H}\) is prescribed, depending on whether the flow is outgoing or
incoming

\[
\mathcal{H} = \frac{|\mathbf{u}|^2}{2} \quad \text{outgoing}
\]
\[
\mathcal{H} = 0 \quad \text{incoming}
\]  

(17)

The outgoing boundary condition assumes that the pressure perturbation \(\tilde{p}\) is zero at an outgoing boundary
and that \(\mathcal{H}\) is constant along streamlines. The incoming boundary condition assumes that \(\mathcal{H}\) is zero infinitely
far away.

²Note the use of the vector identity \((\mathbf{u} \cdot \nabla) \mathbf{u} = \frac{1}{2} \nabla |\mathbf{u}|^2 - \mathbf{u} \times \mathbf{\omega}.

³An option exists in the code to restore the baroclinic torque by decomposing the pressure term

\[
\frac{\nabla \tilde{p}}{\rho} = \frac{\nabla \tilde{p}}{\bar{\rho}} + \left(\frac{1}{\rho} - \frac{1}{\bar{\rho}}\right) \nabla \bar{\rho}
\]

and evaluating the second term on the right hand side at the previous time step. The expression \(\bar{\rho}\) is an average density, equal to \(2\rho_{\text{min}}\rho_{\text{max}}/(\rho_{\text{min}} + \rho_{\text{max}})\). For most large-scale applications, the baroclinic torque is relatively small compared to buoyancy.
2.4 Diffusive Terms (LES)

The viscous stress tensor in the momentum equation is given by

$$\tau = \mu \left( 2 \text{ def } u - \frac{2}{3} (\nabla \cdot u) I \right)$$  \hspace{1cm} (18)

where $I$ is the identity matrix and the deformation tensor is defined

$$\text{def } u \equiv \frac{1}{2} \left[ \nabla u + (\nabla u)^T \right] = \begin{pmatrix}
\frac{\partial u}{\partial x} & \frac{1}{2} \left( \frac{\partial u}{\partial y} + \frac{\partial v}{\partial x} \right) & \frac{1}{2} \left( \frac{\partial u}{\partial z} + \frac{\partial w}{\partial x} \right) \\
\frac{1}{2} \left( \frac{\partial v}{\partial x} + \frac{\partial u}{\partial y} \right) & \frac{\partial v}{\partial y} & \frac{1}{2} \left( \frac{\partial v}{\partial z} + \frac{\partial w}{\partial y} \right) \\
\frac{1}{2} \left( \frac{\partial w}{\partial x} + \frac{\partial u}{\partial z} \right) & \frac{1}{2} \left( \frac{\partial w}{\partial y} + \frac{\partial v}{\partial z} \right) & \frac{\partial w}{\partial z}
\end{pmatrix}$$  \hspace{1cm} (19)

In the numerical model, there are two options for treating the dynamic viscosity $\mu$. For a Large Eddy Simulation (LES) where the grid resolution is not fine enough to capture the mixing processes at all relevant scales, a sub-grid scale model for the viscosity is applied. Following the analysis of Smagorinsky [7], the viscosity can be modeled as

$$\mu_{\text{LES}} = \rho (C_s \Delta)^2 \left( 2 (\text{def } u) \cdot (\text{def } u) - \frac{2}{3} (\nabla \cdot u)^2 \right)^{\frac{1}{2}}$$  \hspace{1cm} (20)

where $C_s$ is an empirical constant, $\Delta$ is a length on the order of the size of a grid cell, and the deformation term is related to the Dissipation Function

$$\Phi \equiv \tau \cdot \nabla u \equiv \mu \left( 2 (\text{def } u) \cdot (\text{def } u) - \frac{2}{3} (\nabla \cdot u)^2 \right)$$  \hspace{1cm} (21)

$$= \mu \left[ 2 \left( \frac{\partial u}{\partial x} \right)^2 + 2 \left( \frac{\partial v}{\partial y} \right)^2 + 2 \left( \frac{\partial w}{\partial z} \right)^2 + \\
\left( \frac{\partial v}{\partial x} + \frac{\partial u}{\partial y} \right)^2 + \left( \frac{\partial w}{\partial y} + \frac{\partial v}{\partial z} \right)^2 + \left( \frac{\partial u}{\partial z} + \frac{\partial w}{\partial x} \right)^2 - \frac{2}{3} \left( \frac{\partial u}{\partial x} + \frac{\partial v}{\partial y} + \frac{\partial w}{\partial z} \right)^2 \right]$$  \hspace{1cm} (22)

The dissipation function is the rate at which kinetic energy is transferred to thermal energy. It is a source term in the energy conservation equation that is usually neglected because it is small – an approximation consistent with the low Mach number equations.

In an LES calculation, the thermal conductivity and material diffusivity are related to the turbulent viscosity by

$$k_{\text{LES}} = \frac{\mu_{\text{LES}} c_p}{\text{Pr}} ; \quad (\rho D)_{l,\text{LES}} = \frac{\mu_{\text{LES}}}{\text{Sc}}$$  \hspace{1cm} (24)

The Prandtl number Pr and the Schmidt number Sc are assumed to be constant for a given scenario.

There have been numerous refinements of the original Smagorinsky model [8, 9, 10], but it is difficult to assess the improvements offered by these newer schemes. There are two reasons for this. First, the structure of the fire plume is so dominated by the large-scale resolvable eddies that even a constant eddy viscosity gives results almost identical to those obtained using the Smagorinsky model [11]. Second, the lack of precision in most large-scale fire test data makes it difficult to assess the relative accuracy of each model. The Smagorinsky model with constant $C_s$ produces satisfactory results for most large-scale applications where boundary layers are not well resolved.
2.5 Diffusive Terms (DNS)

For a Direct Numerical Simulation (DNS), the viscosity, thermal conductivity and material diffusivity are approximated from kinetic theory. The viscosity of the $l$th species is given by

$$\mu_l = \frac{26.69 \times 10^{-7} (M_l T)^{\frac{1}{2}}}{\sigma_l^2 \Omega_v} \text{ kg m s} \quad (25)$$

where $\sigma_l$ is the Lennard-Jones hard-sphere diameter (Å) and $\Omega_v$ is the collision integral, an empirical function of the temperature $T$. The thermal conductivity of the $l$th species is given by

$$k_l = \frac{\mu_l c_p,l}{Pr} \text{ J m K} \quad (26)$$

where the Prandtl number $Pr$ is 0.7. The viscosity and thermal conductivity of a gas mixture are given by

$$\mu_{DNS} = \sum_l Y_l \mu_l ; \quad k_{DNS} = \sum_l Y_l k_l \quad (27)$$

The binary diffusion coefficient of the $l$th species diffusing into the $m$th species is given by

$$D_{lm} = \frac{2.66 \times 10^{-7} T^{3/2}}{M_{lm}^2 \sigma_{lm}^2 \Omega_D} \text{ m}^2 \text{ s} \quad (28)$$

where $M_{lm} = 2(1/M_l + 1/M_m)^{-1}$, $\sigma_{lm} = (\sigma_l + \sigma_m)/2$, and $\Omega_D$ is the diffusion collision integral, an empirical function of the temperature $T$ [12]. It is assumed that nitrogen is the dominant species in any combustion scenario considered here, thus the diffusion coefficient in the species mass conservation equations is that of the given species diffusing into nitrogen

$$(\rho D)_{l,DNS} = \rho D_{l0} \quad (29)$$

where species 0 is nitrogen.
3 Combustion

There are two types of combustion models used in FDS. The choice depends on the resolution of the underlying grid. For a DNS calculation where the diffusion of fuel and oxygen can be modeled directly, a global one-step, finite-rate chemical reaction is most appropriate. However, in an LES calculation where the grid is not fine enough to resolve the diffusion of fuel and oxygen, a mixture fraction-based combustion model is used.

3.1 Mixture Fraction Combustion Model

The mixture fraction combustion model is based on the assumption that large-scale convective and radiative transport phenomena can be simulated directly, but physical processes occurring at small length and time scales must be represented in an approximate manner. The nature of the approximations employed are necessarily a function of the spatial and temporal resolution limits of the computation, as well as our current (often limited) understanding of the phenomena involved.

The actual chemical rate processes that control the combustion energy release are often unknown in fire scenarios. Even if they were known, the spatial and temporal resolution limits imposed by both present and foreseeable computer resources places a detailed description of combustion processes beyond reach. Thus, the model adopted here is based on the assumption that the combustion is mixing-controlled. This implies that all species of interest can be described in terms of a mixture fraction \( Z(x,t) \). The mixture fraction is a conserved quantity representing the fraction of material at a given point that originated as fuel. The relations between the mass fraction of each species and the mixture fraction are known as “state relations”. The state relation for the oxygen mass fraction provides the information needed to calculate the local oxygen mass consumption rate. The form of the state relation that emerges from classical laminar diffusion flame theory is a piecewise linear function. This leads to a “flame sheet” model, where the flame is a two dimensional surface embedded in a three dimensional space. The local heat release rate is computed from the local oxygen consumption rate at the flame surface, assuming that the heat release rate is directly proportional to the oxygen consumption rate, independent of the fuel involved. This relation, originally proposed by Huggett [13], is the basis of oxygen calorimetry.

Start with the most general form of the combustion reaction

\[ \nu_F \text{ Fuel} + \nu_O \text{ O}_2 \rightarrow \sum_i \nu_{Pi} \text{ Products} \]  

(30)

The numbers \( \nu_i \) are the stoichiometric coefficients for the overall combustion process that reacts fuel “F” with oxygen “O” to produce a number of products “P”. The stoichiometric equation (30) implies that the mass consumption rates for fuel and oxidizer are related as follows:

\[ \frac{\dot{m}''_F}{\nu_F M_F} = \frac{\dot{m}''_O}{\nu_O M_O} \]  

(31)

The mixture fraction \( Z \) is defined as:

\[ Z = \frac{sY_F - (Y_O - Y_O^\infty)}{sY_F^I + Y_O^\infty} ; \quad s = \frac{\nu_O M_O}{\nu_F M_F} \]  

(32)

By design, it varies from \( Z = 1 \) in a region containing only fuel to \( Z = 0 \) where the oxygen mass fraction takes on its undepleted ambient value, \( Y_O^\infty \). Note that \( Y_F^I \) is the fraction of fuel in the fuel stream. The quantities \( M_F \) and \( M_O \) are the fuel and oxygen molecular weights, respectively. The mixture fraction satisfies the conservation law

\[ \rho \frac{DZ}{Dt} = \nabla \cdot \rho D \nabla Z \]  

(33)
obtained from a linear combination of the fuel and oxygen conservation equations. The assumption that the chemistry is “fast” means that the reactions that consume fuel and oxidizer occur so rapidly that the fuel and oxidizer cannot co-exist. The requirement that fuel and oxidizer simultaneously vanish defines a flame surface as:

$$Z(x,t) = Z_f; \quad Z_f = \frac{Y^\infty_O}{sY^\infty_f + Y^\infty_O}$$  \hspace{1cm} (34)

The assumption that fuel and oxidizer cannot co-exist leads to the “state relation” between the oxygen mass fraction $$Y_O$$ and $$Z$$

$$Y_O(Z) = \begin{cases} Y^\infty_O(1-Z/Z_f) & Z < Z_f \\ 0 & Z > Z_f \end{cases}$$  \hspace{1cm} (35)

State relations for both reactants and products can be derived by considering the following ideal reaction of a hydrocarbon fuel:

$$C_xH_y + \eta(x+y/4) (O_2 + 3.76 N_2) \rightarrow \max(0, 1 - \eta) C_xH_y + \min(1, \eta) x CO_2 + \min(1, \eta) (y/2) H_2O + \max(0, \eta - 1) (x+y/4) O_2 + \eta(x+y/4)3.76 N_2$$  \hspace{1cm} (36)

Here $$\eta$$ is a parameter ranging from 0 (all fuel with no oxygen) to infinity (all oxygen with no fuel). A correspondence between $$\eta$$ and $$Z$$ is obtained by applying the definition of $$Z$$ (Eq. 32) to the left hand side of Eq. (36). Mass fractions of the products of the infinitely fast reaction (including excess fuel or oxygen) can be obtained from the right hand side of Eq. (36). State relations for the ideal reaction of propane and air is shown in Fig. 1.

An expression for the local heat release rate can be derived from the conservation equations and the state relation for oxygen. The starting point is Huggett’s [13] relationship for the heat release rate as a function
of the oxygen consumption
\[
\dot{q}_m = \Delta H_0 \dot{m}_O''
\]  
(37)

Here, \(\Delta H_0\) is the heat release rate per unit mass of oxygen consumed. The oxygen mass conservation equation
\[
\rho \frac{D Y_O}{D t} = \nabla \cdot \rho D \nabla Y_O + \dot{m}_O''
\]  
(38)
can be transformed into an expression for the local heat release rate using the conservation equation for the mixture fraction (33) and the state relation for oxygen \(Y_O(Z)\).
\[
-\dot{m}_O'' = \nabla \cdot \left( \rho D \frac{dY_O}{dZ} \nabla Z \right) - \frac{dY_O}{dZ} \rho D \nabla Z = \rho D \frac{d^2 Y_O}{dZ^2} \left| \nabla Z \right|^2
\]  
(39)

Neither of these expressions for the local oxygen consumption rate is particularly convenient to apply numerically because of the discontinuity of the derivative of \(Y_O(Z)\) at \(Z = Z_f\). However, an expression for the oxygen consumption rate per unit area of flame sheet can be derived from Eq. (39)
\[
-\dot{m}_O' = \frac{dY_O}{dZ} \bigg|_{Z<Z_f} \rho D \nabla Z \cdot n
\]  
(40)

In the numerical algorithm, the local heat release rate is computed by first locating the flame sheet, then computing the local heat release rate per unit area, and finally distributing this energy to the grid cells cut by the flame sheet. In this way, the ideal, infinitely thin flame sheet is smeared out over the width of a grid cell, consistent with all other gas phase quantities.

The above procedure for determining the local heat release rate works well for calculations in which the fire is adequately resolved. A measure of how well the fire is resolved is given by the nondimensional expression \(D^*/\delta x\), where \(D^*\) is a characteristic fire diameter
\[
D^* = \left( \frac{Q}{\rho_c c_p T_\infty \sqrt{g}} \right)^{\frac{3}{2}}
\]  
(41)

and \(\delta x\) is the nominal size of a grid cell\(^4\). The quantity \(D^*/\delta x\) can be thought of as the number of computational cells spanning the characteristic (not necessarily the physical) diameter of the fire. The more cells spanning the fire, the better the resolution of the calculation. For fire scenarios where \(D^*\) is small relative to the physical diameter of the fire, and/or the numerical grid is relatively coarse, the stoichiometric surface \(Z = Z_f\) will underestimate the observed flame height [14]. It has been found empirically that a good estimate of flame height can be found for crude grids if a different value of \(Z\) is used to define the combustion region
\[
\frac{Z_{f,\text{eff}}}{Z_f} = \min \left( 1, \ C \frac{D^*}{\delta x} \right)
\]  
(42)

Here \(C\) is an empirical constant to be used for all fire scenarios. As the resolution of the calculation increases, the \(Z_{f,\text{eff}}\) approaches the ideal value, \(Z_f\). The benefit of the expression is that it provides a quantifiable measure of the grid resolution that takes into account not only the size of the grid cells, but also the size of the fire.

\(^4\)The characteristic fire diameter is related to the characteristic fire size via the relation \(Q^* = (D^*/D)^{5/2}\), where \(D\) is the physical diameter of the fire.
3.2 Finite-rate Reaction (DNS)

In a DNS calculation, the diffusion of fuel and oxygen can be modeled directly, thus it is possible to implement a relatively simple one-step chemical reaction. Consider the reaction of oxygen and a hydrocarbon fuel

\[ \nu_{C_xH_y} C_xH_y + \nu_{O_2} O_2 \rightarrow \nu_{CO_2} CO_2 + \nu_{H_2O} H_2O \]  

(43)

The reaction rate is given by the expression

\[ \frac{d[C_xH_y]}{dt} = -B [C_xH_y]^a [O_2]^b e^{-E/RT} \]  

(44)

Suggested values of \( B, E, a \) and \( b \) for various hydrocarbon fuels are given in Refs. [15, 16]. It should be understood that the implementation of any of these one-step reaction schemes is still very much a research exercise because it is not universally accepted that combustion phenomena can be represented by such a simple mechanism. Efforts are currently underway to determine in what cases a one-step reaction mechanism provides a valid description of the combustion process.
4 Thermal Radiation

The Radiative Transport Equation (RTE) for a non-scattering gas is

\[ s \cdot \nabla I_\lambda(x, s) = \kappa(x, \lambda) [I_b(x) - I(x, s)] \]  

(45)

where \( I_\lambda(x, s) \) is the radiation intensity at wavelength \( \lambda \), \( I_b(x) \) is the source term given by the Planck function, \( s \) is the unit normal direction vector and \( \kappa(x) \) is the absorption coefficient. In practical simulations the spectral dependence can not be solved accurately. Instead, the radiation spectrum is divided into a relatively small number of bands, and a separate RTE is derived for each band. The limits of the bands are selected to give an accurate representation of the most important radiation bands of CO\(_2\) and water. The band specific RTE’s are now

\[ s \cdot \nabla I_n(x, s) = \kappa_n(x) [I_{b,n}(x) - I(x, s)], \quad n = 1...N \]  

(46)

where \( I_n \) is the intensity integrated over the band \( n \), and \( \kappa_n \) is the appropriate mean absorption coefficient inside the band. The source term can be written as a fraction of the blackbody radiation

\[ I_{b,n} = F_n(\lambda_{min}, \lambda_{max}) \sigma T^4 / \pi \]  

(47)

where \( \sigma \) is the Stefan-Boltzmann constant. The calculation of factors \( F_n \) is explained in Ref. [17]. When the intensities corresponding to the bands are known, the total intensity is calculated by summing over all the bands

\[ I(x, s) = \sum_{n=1}^{N} I_n(x, s) \]  

(48)

From a series of numerical experiments it was found that six bands are usually enough (\( N = 6 \)). If the absorption of the fuel is known to be important, separate bands can be reserved for fuel, and the total number of bands is increased to ten (\( N = 10 \)). For simplicity, the fuel is assumed to be CH\(_4\). The limits of the bands are shown in Table 1.

**TABLE 1: Limits of the spectral bands.**

<table>
<thead>
<tr>
<th>9 Band Model</th>
<th>1</th>
<th>2</th>
<th>3</th>
<th>4</th>
<th>5</th>
<th>6</th>
<th>7</th>
<th>8</th>
<th>9</th>
</tr>
</thead>
<tbody>
<tr>
<td>Major Species</td>
<td>Soot</td>
<td>CO(_2)</td>
<td>H(_2)O, Soot</td>
<td>Soot</td>
<td>CO(_2)</td>
<td>H(_2)O</td>
<td>H(_2)O</td>
<td>Soot</td>
<td>Soot</td>
</tr>
<tr>
<td>( \nu ) (1/cm)</td>
<td>10000</td>
<td>3800</td>
<td>3400</td>
<td>2800</td>
<td>2400</td>
<td>2174</td>
<td>1429</td>
<td>1160</td>
<td>1000</td>
</tr>
<tr>
<td>( \lambda ) (( \mu m ))</td>
<td>1.00</td>
<td>2.63</td>
<td>2.94</td>
<td>3.57</td>
<td>4.17</td>
<td>4.70</td>
<td>7.00</td>
<td>8.62</td>
<td>10.0</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>6 Band Model</th>
<th>1</th>
<th>2</th>
<th>3</th>
<th>4</th>
<th>5</th>
<th>6</th>
</tr>
</thead>
<tbody>
<tr>
<td>Major Species</td>
<td>Soot</td>
<td>CO(_2)</td>
<td>H(_2)O, Soot</td>
<td>CH(_4)</td>
<td>Soot</td>
<td>CO(_2)</td>
</tr>
</tbody>
</table>

Even with a reasonably small number of bands, the solution of \( N \) RTE’s is very time consuming. Fortunately, in most large-scale fire scenarios soot is the most important combustion product controlling the thermal radiation from the fire and hot smoke. As the radiation spectrum of soot is continuous, it is possible to assume that the gas behaves as a gray medium. The spectral dependence is lumped into one absorption coefficient (\( N = 1 \)) and the source term is given by the blackbody radiation intensity

\[ I_b(x) = \sigma T(x)^4 / \pi \]  

(49)

In optically thin flames, where the amount of soot is small compared to the amount of CO\(_2\) and water, the gray gas assumption may produce significant overpredictions of the emitted radiation.
For the calculation of the gray or band-mean absorption coefficients, $\kappa_n$, a narrow-band model, RADCAL [18], has been implemented in FDS. At the start of a simulation the absorption coefficient(s) are tabulated as a function of mixture fraction and temperature. During the simulation the local absorption coefficient is found by table-lookup.

In calculations of limited spatial resolution, the source term, $I_b$, in the RTE requires special treatment in the neighborhood of the flame sheet because the temperatures are smeared out over a grid cell and are thus considerably lower than one would expect in a diffusion flame. Because of its dependence on the temperature raised to the fourth power, the source term must be modeled in those grid cells cut by the flame sheet. Elsewhere, there is greater confidence in the computed temperature, and the source term can assume its ideal value there

$$\kappa I_b = \begin{cases} \kappa \sigma T^4 / \pi & \text{Outside flame zone} \\ \chi_r q''' / 4\pi & \text{Inside flame zone} \end{cases}$$

Here, $q'''$ is the chemical heat release rate per unit volume and $\chi_r$ is the local fraction of that energy emitted as thermal radiation. Note the difference between the prescription of a local $\chi_r$ and the resulting global equivalent. For a small fire ($D < 1$ m), the local $\chi_r$ is approximately equal to its global counterpart. However, as the fire increases in size, the global value will typically decrease due to a net re-absorption of the thermal radiation by the increasing smoke mantle.

The boundary condition for the radiation intensity leaving a gray diffuse wall is given as

$$I_w(s) = \varepsilon I_{bw} + \left(1 - \varepsilon / \pi \right) \int_{s^\prime \cdot \mathbf{n}_w < 0} I_w(s^\prime) |s^\prime \cdot \mathbf{n}_w| d\Omega$$

where $I_w(s)$ is the intensity at the wall, $\varepsilon$ is the emissivity, and $I_{bw}$ is the black body intensity at the wall.

The radiative transport equation (46) is solved using techniques similar to those for convective transport in finite volume methods for fluid flow [19], thus the name given to it is the Finite Volume Method (FVM). To obtain the discretized form of the RTE, the unit sphere is divided into a finite number of solid angles. In each grid cell a discretized equation is derived by integrating equation (45) over the cell $ijk$ and the control angle $\delta\Omega_l$, to obtain

$$\int_{V_{ijk}} \int_{\Omega_l} s \cdot \nabla I_n(x,s) dV d\Omega = \int_{V_{ijk}} \kappa_n(x) \left[I_{b,n}(x) - I_n(x,s)\right] dV d\Omega$$

The volume integral on the left hand side is replaced by a surface integral over the cell faces using the divergence theorem. Assuming that the radiation intensity $I(x,s)$ is constant on each of the cell faces, the surface integral can be approximated by a sum over the cell faces. More detail on the discretization and solution of the RTE can be found in Section 7.

The radiant heat flux vector $\mathbf{q}_r$ is defined

$$\mathbf{q}_r(x) = \int s I(x,s) d\Omega$$

The radiative loss term in the energy equation is

$$-\nabla \cdot \mathbf{q}_r(x) = \kappa(x) \left[U(x) - 4\pi I_b(x)\right] ; \quad U(x) = \int_{4\pi} I(x,s) d\Omega$$

In words, the net radiant energy gained by a grid cell is the difference between that which is absorbed and that which is emitted.
5 Thermal Boundary Conditions

The type of thermal boundary condition applied at any given surface depends on whether that surface is to heat up and burn, whether the burning rate will simply be prescribed, or whether there is to be any burning at all.

5.1 Convective Heat Transfer to Walls

The heat fluxes to a solid surface consist of gains and losses from convection and radiation. The radiative flux at the surface is obtained from the boundary condition for the radiation equation, Eq. (51).

The calculation of the convective heat flux depends on whether one is performing a Direct Numerical Simulation (DNS) or a Large Eddy Simulation (LES). In a DNS calculation, the convective heat flux to a solid surface \( q''_c \) is obtained directly from the gas temperature gradient at the boundary

\[ q''_c = -k \frac{dT}{dx} \]  

(55)

where \( x \) is the coordinate pointing into the solid. In an LES calculation, the convective heat flux to the surface is obtained from a combination of natural and forced convection correlations

\[ q''_c = h \Delta T \text{ W/m}^2 \; ; \; h = \max \left[ C|\Delta T|^{\frac{1}{3}}, 0.037 \frac{k}{L} \left( \frac{|u|}{v} \right)^{0.8} Pr^{\frac{3}{5}} \right] \text{ W/m}^2/K \]  

(56)

where \( \Delta T \) is the difference between the wall and the gas temperature, \( C \) is the coefficient for natural convection (1.43 for a horizontal surface and 0.95 for a vertical surface) [20], \( L \) is a characteristic distance related to the size of the plate, \( k \) is the thermal conductivity of the gas, and \( v \) is the kinematic viscosity of the gas.

5.2 Pyrolysis Model, Thermally-Thick Solid

If the surface material is assumed to be thermally-thick, a one-dimensional heat conduction equation for the material temperature, \( T_s(x,t) \), is applied in the direction \( x \) pointing into the air/solid interface \( (x = 0) \)

\[ \rho_s c_s \frac{\partial T_s}{\partial t} = k_s \frac{\partial^2 T_s}{\partial x^2} \; ; \; -k_s \frac{\partial T_s}{\partial x}(0,t) = q''_c + q''_r \]  

(57)

where \( \rho_s, c_s \) and \( k_s \) are the (constant) density, specific heat and conductivity of the material; and \( q''_c \) and \( q''_r \) are the convective and radiative heat fluxes at the surface. If the material is assumed to ignite and burn at some prescribed temperature, \( T_p \), the boundary conditions of Eq. (57) change

\[ T_s(0,t) = T_p \; ; \; q''_p = q''_c + q''_r + k_s \frac{\partial T_s}{\partial x}(0,t) \]  

(58)

Here, \( q''_p \) is energy available for pyrolyzing fuel via the simple relationship

\[ \dot{m}'' = \frac{q''_p}{\Delta H_v} \]  

(59)

where \( \dot{m}'' \) is the mass loss rate of fuel and \( \Delta H_v \) is the heat of vaporization.
5.3 Pyrolysis Model, Thermally-Thin Solid

If the surface material is assumed to be thermally-thin, that is, its temperature is assumed uniform across its width, its temperature, $T_s(t)$, is governed by its density, specific heat and thickness $\delta$

$$\frac{dT_s}{dt} = \frac{q''_c + q''_r}{\rho_s c_s \delta}$$

In this case, the individual values of the parameters $\rho_s$, $c_s$ and $\delta$ are not as important as their product. At a prescribed pyrolysis temperature, $T_p$

$$T_s = T_p ; \quad q''_p = q''_c + q''_r$$

where the convective and radiative fluxes are summed over the front and the back of the thin fuel.

5.4 Pyrolysis Model, Liquid Fuels

The rate at which liquid fuel evaporates when burning is a function of the liquid temperature and the concentration of fuel vapor above the pool surface. Equilibrium is reached when the partial pressure of the fuel vapor above the surface equals the Clausius-Clapeyron pressure

$$p_{cc} = p_0 \exp \left[ - \frac{h_v M_f}{R} \left( \frac{1}{T_s} - \frac{1}{T_b} \right) \right]$$

where $h_v$ is the heat of vaporization, $M_f$ is the molecular weight, $T_s$ is the surface temperature, and $T_b$ is the boiling temperature of the fuel.

For simplicity, the liquid fuel itself is treated like a thermally-thick solid for the purpose of computing the heat conduction. There is no computation of the convection of the liquid within the pool.
6 Sprinklers

Simulating the effects of a sprinkler spray involves a number of pieces: predicting activation, computing the droplet trajectories and tracking the water as it drips onto the burning commodity.

6.1 Sprinkler Activation

The temperature of the sensing element of a given sprinkler is estimated from the differential equation put forth by Heskestad and Bill [21], with the addition of several terms to account for radiative heating and cooling by water droplets in the gas stream from previously activated sprinklers

\[ \frac{dT_l}{dt} = \frac{\sqrt{|u|}}{RTI} (T_g - T_l) - \frac{C}{RTI} (T_l - T_m) - \frac{C_2}{RTI} \beta |u| \]  

Here \( T_l \) is the link temperature, \( T_g \) is the gas temperature in the neighborhood of the link, \( T_m \) is the temperature of the sprinkler mount (assumed ambient), and \( \beta \) is the volume fraction of (liquid) water in the gas stream. The sensitivity of the detector is characterized by the value of RTI. The amount of heat conducted away from the link by the mount is indicated by the “C-Factor”, \( C \). The constant \( C_2 \) has been empirically determined by DiMarzo [22] to be \( 6 \times 10^6 \, \text{K}/(\text{m/s})^2 \), and its value is relatively constant for different types of sprinklers.

6.2 Sprinkler Droplet Size Distribution

Once activation is predicted, a sampled set of spherical water droplets is tracked from the sprinkler to either the floor or the burning commodity. In order to compute the droplet trajectories, the initial size and velocity of each droplet must be prescribed. This is done in terms of random distributions. The initial droplet size distribution of the sprinkler spray is expressed in terms of its Cumulative Volume Fraction (CVF), a function that relates the fraction of the water volume (mass) transported by droplets less than a given diameter. Researchers at Factory Mutual have suggested that the CVF for an industrial sprinkler may be represented by a combination of log-normal and Rosin-Rammler distributions [23]

\[
F(d) = \begin{cases} 
\frac{1}{\sqrt{2\pi}} \int_0^d \frac{1}{\sigma'} e^{-\frac{[\ln(d'/d_m)]^2}{2\sigma'^2}} \, dd' \quad (d \leq d_m) \\
1 - e^{-0.693 \left( \frac{d}{d_m} \right)^\gamma} \quad (d_m < d)
\end{cases}
\]  

(64)

where \( d_m \) is the median droplet diameter (i.e. half the mass is carried by droplets with diameters of \( d_m \) or less), and \( \gamma \) and \( \sigma' \) are empirical constants equal to about 2.4 and 0.6, respectively. The median drop diameter is a function of the sprinkler orifice diameter, operating pressure, and geometry. Research at Factory Mutual has yielded a correlation for the median droplet diameter [24]

\[
\frac{d_m}{D} \propto \text{We}^{-\frac{1}{\gamma}}
\]  

(65)

where \( D \) is the orifice diameter of the sprinkler. The Weber number, the ratio of inertial forces to surface tension forces, is given by

\[
\text{We} = \frac{\rho_w U^2 D}{\sigma_w}
\]  

(66)
Eq. (65) appears to be independent of flow rate and operating pressure. Three different sprinklers were tested in their study with orifice diameters of 16.3 mm, 13.5 mm, 12.7 mm and the constants were approximately 4.3, 2.9, 2.3, respectively. The strike plates of the two smaller sprinklers were notched, while that of the largest sprinkler was not [24].

In the numerical algorithm, the size of the sprinkler droplets are chosen to mimic the Rosin-Rammler/log-normal distribution. A Probability Density Function (PDF) for the droplet diameter is defined

\[
f(d) = F'(d) d^3 / \int_0^\infty F'(d') d'd'
\]  

Droplet diameters are randomly selected by equating the Cumulative Number Fraction of the droplet distribution with a uniformly distributed random variable \(U\)

\[
U(d) = \int_0^d f(d') dd'
\]

Figure 2 displays typical Cumulative Volume Fraction and Cumulative Number Fraction functions.

Every droplet from a given sprinkler is not tracked. Instead, a sampled set of the droplets is tracked. Typically, 1,000 droplets per sprinkler per second are tracked (50 droplets every 0.05 s, depending on user preference). If the droplet diameters are selected randomly, too much CPU time is wasted tracking smaller droplets, and the larger droplets, which carry most of the water, are under-represented in the sampled population. To remedy this problem, the uniform variable \(U\) is transformed by a function \(g\) of the form

\[
g(U) = U^{1/n} \quad (n > 1)
\]

Now the sampled population will contain a greater number of larger droplets, but whose weighting factors will be reduced by a factor

\[
g'(U) = \frac{1}{n} U^{1/n-1}
\]
The procedure for selecting droplet sizes is as follows: Suppose water is leaving the sprinkler at a mass flow rate of $\dot{m}$. Suppose also that the time interval for droplet insertion into the numerical simulation is $\delta t$, and the number of droplets inserted each time interval is $N$. Choose $N$ uniformly distributed random numbers between 0 and 1, call them $U_i$, map them according to Eq. (69) into a transformed set of random numbers biased towards the larger droplet sizes, obtain $N$ droplet diameters $d_i$ based on the given droplet size distribution (Eq. (68), and then compute a weighting constant $C$ from the mass balance

$$\dot{m} \delta t = C \sum_{i=1}^{N} g'(U_i) \frac{4}{3} \pi \rho_w \left( \frac{d_i}{2} \right)^3$$

(71)

The mass and heat transferred from each droplet will be multiplied by the weighting factor $C g'(U_i)$.

6.3 Sprinkler Droplet Trajectory in Air

For a sprinkler spray, the force term $f$ in Eq. (3) represents the momentum transferred from the water droplets to the gas. It is obtained by summing the force transferred from each droplet in a grid cell and dividing by the cell volume

$$f = \frac{1}{2} \sum \rho C_D \pi r_d^2 (u_d - u) |u_d - u| \delta x \delta y \delta z$$

(72)

where $C_D$ is the drag coefficient, $r_d$ is the droplet radius, $u_d$ is the velocity of the droplet, $u$ is the velocity of the gas, $\rho$ is the density of the gas, and $\delta x \delta y \delta z$ is the volume of the grid cell. The trajectory of an individual droplet is governed by the equation

$$\frac{d}{dt}(m_d u_d) = m_d g - \frac{1}{2} \rho C_D \pi r_d^2 (u_d - u)|u_d - u|$$

(73)

where $m_d$ is the mass of the droplet. The drag coefficient is a function of the local Reynolds number

$$C_D = \begin{cases} 
24/\text{Re} & \text{Re} < 1 \\
24 \left(1 + 0.15 \text{Re}^{0.687}\right) / \text{Re} & 1 < \text{Re} < 1000 \\
0.44 & 1000 < \text{Re}
\end{cases}$$

(74)

$$\text{Re} = \frac{\rho |u_d - u| 2r_d}{\mu}$$

(75)

where $\mu$ is the dynamic viscosity of air.

6.4 Sprinkler Droplet Transport on a Surface

When a water droplet hits a solid horizontal surface, it is assigned a random horizontal direction and moves at a fixed velocity until it reaches the edge, at which point it drops straight down at the same fixed velocity. This “dripping” velocity has been measured to be on the order of 0.5 m/s [26]. Penetration of water into porous materials is handled very crudely by assigning a fraction of the water droplets that strike a solid horizontal surface to go straight through the solid at a slow velocity. Neither the fraction nor the velocity has been validated.

6.5 Mass and Energy Transfer from Droplets

The evaporation of water droplets is handled semi-empirically. A water droplet suspended in air will evaporate as a function of the droplet equilibrium vapor mass fraction, the local gas phase vapor mass fraction,
the heat transfer to the droplet, and the droplet’s motion relative to the gas. A correlation for the mass loss rate of a droplet that involves these parameters is given here [25]

$$\frac{dm_d}{dt} = -2\pi r_d \cdot Sh \cdot \rho \cdot D \cdot (Y_d - Y_g)$$  \hspace{1cm} (76)

The subscripts $d$ and $g$ refer to the droplet and gas, respectively, $m_d$ is the droplet mass, $D$ is the diffusion coefficient for water vapor into air, $Y$ is the water vapor mass fraction, and $Sh$ is the droplet Sherwood number, given by a correlation involving the Reynolds and Schmidt numbers

$$Sh = 2 + 0.6 \cdot Re^{1/2} \cdot Sc^{1/2}$$ \hspace{1cm} (77)

The vapor mass fraction of the gas, $Y_g$, is obtained from the overall set of mass conservation equations and the vapor mass fraction of the droplet is obtained from the Clausius-Clapeyron equation

$$X_d = \exp \left[ \frac{h_v \cdot M_w}{R} \left( \frac{1}{T_b} - \frac{1}{T_d} \right) \right] ; \hspace{1cm} Y_d = \frac{X_d}{X_d(1 - M_a/M_w) + M_a/M_w}$$ \hspace{1cm} (78)

where $X_d$ is the droplet water vapor volume fraction, $h_v$ is the heat of vaporization, $M_w$ is the molecular weight of water, $M_a$ is the molecular weight of air, $R$ is the gas constant, $T_b$ is the boiling temperature of water and $T_d$ is the droplet temperature.

In addition to calculating the mass transfer due to evaporation, the transfer of energy must also be calculated. The droplet heats up due to the convective heat transfer across the surface of the droplet minus the energy required to evaporate water

$$m_d \cdot c_{p,w} \cdot \frac{dT_d}{dt} = A_d \cdot h_d \cdot (T_g - T_d) - \frac{dm_d}{dt} \cdot h_v$$ \hspace{1cm} (79)

Here $c_{p,w}$ is the specific heat of water, $A_d = 4\pi r_d^2$ is the surface area of the droplet, $h_d$ is the heat transfer coefficient, given by

$$h_d = \frac{Nu \cdot k}{2r_d} ; \hspace{1cm} Nu = 2 + 0.6 \cdot Re^{1/2} \cdot Pr^{1/3}$$ \hspace{1cm} (80)

$Nu$ is the Nusselt number, $k$ is the thermal conductivity of air, and the Prandtl number, $Pr$, is about 0.7 for air. The Sherwood number, $Sh$, is analoguous to the Nusselt number, with the Schmidt number about 0.6 compared to the 0.7 for the Prandtl number.

Finally, the exchange of mass and energy between the droplets and the gas results in an additional term that must be added to the expression for the divergence, Eq. (10)

$$\nabla \cdot \mathbf{u} = \ldots + \frac{R}{\gamma p_0} \left( \rho \sum \left( \frac{Y_i}{M_i} \right) \frac{\partial T}{\partial t} + \frac{T}{M_w} \cdot \dot{m}_w''' \right)$$ \hspace{1cm} (81)

where $\dot{m}_w'''$ is the water evaporation rate per unit volume. The liquid water droplets are assumed to occupy no volume, simplifying the analysis.

### 6.6 Fire Suppression by Water

The above two sections describe heat transfer from a droplet of water to a hot gas, a hot solid, or both. Although there is some uncertainty in the values of the respective heat transfer coefficients, the fundamental physics are fairly well understood. However, when the water droplets encounter burning surfaces, simple heat transfer correlations become more difficult to apply. The reason for this is that the water is not only cooling the surface and the surrounding gas, but it is also changing the pyrolysis rate of the fuel. If the
surface of the fuel is planar, it is possible to characterize the decrease in the pyrolysis rate as a function of the decrease in the total heat feedback to the surface. Unfortunately, most fuels of interest in fire applications are multi-component solids with complex geometry at scales unresolvable by the computational grid.

To date, most of the work in this area has been performed at Factory Mutual. An important paper on the subject is by Yu et al. [27]. The authors consider dozens of rack storage commodity fires of different geometries and water application rates, and characterize the suppression rates in terms of a few global parameters. Their analysis yields an expression for the total heat release rate from a rack storage fire after sprinkler activation

\[ Q = Q_0 e^{-k(t-t_0)} \]  

where \( Q_0 \) is the total heat release rate at the time of application \( t_0 \), and \( k \) is a fuel-dependent constant. For the FMRC Standard Plastic commodity \( k \) is given as

\[ k = 0.716 \dot{m}_w'' - 0.0131 \text{ s}^{-1} \]  \hspace{1cm} (83)

where \( \dot{m}_w'' \) is the flow rate of water impinging on the box tops, divided by the area of exposed surface (top and sides). It is expressed in units of kg/m\(^2\)/s. For the Class II commodity, \( k \) is given as

\[ k = 0.536 \dot{m}_w'' - 0.0040 \text{ s}^{-1} \]  \hspace{1cm} (84)

Unfortunately, this analysis is based on global water flow and burning rates. Equation (82) accounts for both the cooling of non-burning surfaces as well as the decrease in heat release rate of burning surfaces. In the FDS model, the cooling of unburned surfaces and the reduction in the heat release rate are computed locally, thus it is awkward to apply a global suppression rule. However, the exponential nature of suppression by water is observed both locally and globally, thus it is assumed that the local burning rate of the fuel can be expressed in the form [26]

\[ \dot{m}_f''(t) = \dot{m}_{f,0}(t) e^{-\int k(t) dt} \]  \hspace{1cm} (85)

Here \( \dot{m}_{f,0}(t) \) is the burning rate per unit area of the fuel when no water is applied and \( k(t) \) is a linear function of the local water mass per unit area, \( m_w'' \), expressed in units of kg/m\(^2\),

\[ k(t) = a m_w''(t) \text{ s}^{-1} \]  \hspace{1cm} (86)

Note that \( a \) is an empirical constant.
7 Numerical Method

This section presents the details of the numerical algorithm. First the equations that are being solved are presented. Each of the conservation equations emphasize the importance of the velocity divergence and vorticity fields, as well as the close relationship between the thermally expandable fluid equations \[5\] and the Boussinesq equations for which the authors have developed highly efficient solution procedures \[28, 29\]. All spatial derivatives are approximated by second order central differences and the flow variables are updated in time using an explicit second order predictor-corrector scheme.

7.1 Simplified Equations

Regardless of whether one is performing an LES or a DNS calculation, the overall solution algorithm is the same. The equations derived in Section 2 that are to be solved numerically are listed again here.

Conservation of Mass

\[
\frac{\partial \rho}{\partial t} + \mathbf{u} \cdot \nabla \rho = -\rho \nabla \cdot \mathbf{u} \tag{87}
\]

Conservation of Species

\[
\frac{\partial \rho Y_l}{\partial t} + \mathbf{u} \cdot \nabla \rho Y_l = -\rho Y_l \nabla \cdot \mathbf{u} + \nabla \cdot \rho D \nabla Y_l + W''_l \tag{88}
\]

Conservation of Momentum

\[
\frac{\partial \mathbf{u}}{\partial t} + \mathbf{u} \times \mathbf{\omega} + \nabla H = \frac{1}{\rho} ((\rho - \rho_\infty)\mathbf{g} + \mathbf{f} + \nabla \cdot \mathbf{\tau}) \tag{89}
\]

Divergence Constraint

\[
\nabla \cdot \mathbf{u} = \gamma \frac{1}{\gamma - 1} \left( q''' - \nabla \cdot \mathbf{q}_p + \nabla \cdot k \nabla T + \nabla \cdot \sum_l c_{p,l} T \rho D \nabla Y_l - \frac{1}{\gamma - 1} \frac{dp_0}{dt} \right) \tag{90}
\]

Equation of State

\[
p_0(t) = \rho T \mathcal{R} \sum_l Y_l / M_l \tag{91}
\]

Notice that the source terms from the energy conservation equation have been incorporated into the divergence and ultimately are involved in the mass conservation equation. The temperature is found from the density and background pressure via the equation of state.

7.2 Temporal Discretization

All calculations start with ambient initial conditions. At the beginning of each time step, the quantities \(\rho^n, Y^n_i, \mathbf{u}^n, H^n_\infty,\) and \(p_0^n\) are known. All other quantities can be derived from them. Note that the superscript \((n+1)_e\) refers to an estimate of the value of the quantities at the \((n+1)\)st time step.

1. The thermodynamic quantities \(\rho, Y_i,\) and \(p_0\) are estimated at the next time step with an explicit Euler step. For example, the density is estimated

\[
\rho^{(n+1)_e} = \rho^n - \delta t (\mathbf{u}^n \cdot \nabla \rho^n + \rho^n \nabla \cdot \mathbf{u}^n) \tag{92}
\]

The divergence \((\nabla \cdot \mathbf{u})^{(n+1)_e}\) is formed from these estimated thermodynamic quantities. The normal velocity components at boundaries that are needed to form the divergence are assumed known.
2. A Poisson equation for the pressure is solved with a direct solver

\[ \nabla^2 p^n = -\frac{(\nabla \cdot u)^{(n+1)}_e - (\nabla \cdot u)^n}{\delta t} - \nabla \cdot F^n \] (93)

Note that the vector \( F \) contains the convective, diffusive and force terms of the momentum equation. These will be described in detail below. Then the velocity is estimated at the next time step

\[ u^{(n+1)}_e = u^n - \delta t (F^n + \nabla p^n) \] (94)

Note that the divergence of the estimated velocity field is identically equal to the estimated divergence \( (\nabla \cdot u)^{(n+1)}_e \) that was derived from the estimated thermodynamic quantities. The time step is checked at this point to ensure that

\[ \delta t < \min \left( \frac{\delta x}{u}, \frac{\delta y}{v}, \frac{\delta z}{w} \right) \] (95)

If the time step is too large, it is reduced so that it satisfies the CFL condition and the procedure starts from the beginning of the time step. If the time step satisfies the stability condition, the procedure continues.

3. The thermodynamic quantities \( \rho, Y_i, \) and \( p_0 \) are corrected at the next time step. For example, the density is corrected

\[ \rho^{n+1} = \frac{1}{2} \left( \rho^n + \rho^{(n+1)}_e - \delta t (u^{(n+1)}_e \cdot \nabla \rho^{(n+1)}_e + p^{(n+1)}_e \nabla \cdot u^{(n+1)}_e) \right) \] (96)

The divergence \( (\nabla \cdot u)^{(n+1)} \) is derived from the corrected thermodynamic quantities.

4. The pressure is recomputed using estimated quantities

\[ \nabla^2 p^{(n+1)}_e = -\frac{2(\nabla \cdot u)^{n+1} - (\nabla \cdot u)^{(n+1)}_e - (\nabla \cdot u)^n}{\delta t} - \nabla \cdot F^{(n+1)}_e \] (97)

The velocity is then corrected

\[ u^{n+1} = \frac{1}{2} \left[ u^n + u^{(n+1)}_e - \delta t \left( F^{(n+1)}_e + \nabla p^{(n+1)}_e \right) \right] \] (98)

Note again that the divergence of the corrected velocity field is identically equal to the corrected divergence.

7.3 Spatial Discretization

Spatial derivatives in the governing equations are written as second order accurate finite differences on a rectilinear grid. The overall domain is a rectangular box that is divided into rectangular grid cells. Each cell is assigned indices \( i, j \) and \( k \) representing the position of the cell in the \( x, y \) and \( z \) directions, respectively. Scalar quantities are assigned in the center of each grid cell, thus \( \rho_{ijk}^n \) is the density at the \( n \)th time step in the center of the cell whose indices are \( i, j \) and \( k \). Vector quantities like velocity are assigned at cell faces, thus the \( x \) component of velocity \( u \) is defined at the faces whose normals are parallel to the \( x \)-axis, the \( y \) component \( v \) is defined at the faces whose normals are parallel to the \( y \)-axis, and the \( z \) component \( w \) is defined at the faces whose normals are parallel to the \( z \)-axis. The quantity \( u_{ijk}^n \) is the \( x \) component of velocity at the forward pointing face of the \( ijk \)th cell; \( u_{i-1,j,k}^n \) is at the backward pointing face of the \( ijk \)th cell.
7.4 Large Eddy vs. Direct Numerical Simulation

The major difference between an LES and a DNS calculation is the form of the viscosity, and the thermal and material diffusivities. For a Large Eddy Simulation, the dynamic viscosity is defined at cell centers

\[ \mu_{ijk} = \rho_{ijk} (C_s \Delta)^2 |S| \]  

(99)

where \( C_s \) is an empirical constant, \( \Delta = (\delta x \delta y \delta z)^\frac{1}{3} \), and

\[ |S|^2 = 2 \left( \frac{\partial u}{\partial x} \right)^2 + 2 \left( \frac{\partial v}{\partial y} \right)^2 + \left( \frac{\partial w}{\partial z} \right)^2 + \left( \frac{\partial u}{\partial y} + \frac{\partial v}{\partial x} \right)^2 + \left( \frac{\partial u}{\partial z} + \frac{\partial w}{\partial x} \right)^2 + \left( \frac{\partial v}{\partial z} + \frac{\partial w}{\partial y} \right)^2 - \frac{2}{3} (\nabla \cdot \mathbf{u})^2 \]  

(100)

The quantity \( |S| \) consists of second order spatial differences averaged at cell centers. The thermal conductivity and material diffusivity of the fluid are related to the viscosity by

\[ k_{ijk} = \frac{c_{p,0} \mu_{ijk}}{\text{Pr}} \]

; \( (\rho D)_{ijk} = \frac{\mu_{ijk}}{\text{Sc}} \)  

(101)

where \( \text{Pr} \) is the Prandtl number and \( \text{Sc} \) is the Schmidt number, both assumed constant. Note that the specific heat \( c_{p,0} \) is that of the dominant species of the mixture. Based on simulations of smoke plumes, \( C_s \) is 0.14, \( \text{Pr} \) and \( \text{Sc} \) are 0.2. There is no rigorous justification for these choices.

The dynamic viscosity, thermal conductivity and diffusion coefficients for a DNS calculation are defined at cell centers

\[ \mu_{ijk} = \sum_l Y_{l,ijk} \mu_l(T_{ijk}) \]  

(102)

\[ k_{ijk} = \sum_l Y_{l,ijk} k_l(T_{ijk}) \]  

(103)

\[ D_{l,ijk} = D_{l0} (T_{ijk}) \]  

(104)

where the values for each individual species are approximated from kinetic theory [12]. The term \( D_{l0} \) is the binary diffusion coefficient for species \( l \) diffusing into the predominant species 0, usually nitrogen. It is often the case that the numerical grid is too coarse to resolve steep gradients in flow quantities when the temperature is near ambient. However, as the temperature increases and the diffusion coefficients increase in value, the situation improves. As a consequence, there is a provision in the numerical algorithm to place a lower bound on the viscous coefficients to avoid numerical instabilities at temperatures close to ambient.
7.5 The Mass Transport Equations

Due to the low Mach number approximation being used in the model, the mass and energy equations are combined by way of the divergence. The divergence of the flow field contains much of the fire-specific source terms described above.

7.5.1 Convective and Diffusive Transport

The density at the center of the $i,j,k$th cell is updated in time with the following predictor-corrector scheme. In the predictor step, the density at the $(n+1)$st time level is estimated based on information at the $n$th level

$$\frac{\rho_{ijk}^{(n+1)} - \rho_{ijk}^n}{\delta t} + (\mathbf{u} \cdot \nabla \rho)_{ijk}^{n} = -\rho_{ijk}^n (\nabla \cdot \mathbf{u})_{ijk}^n$$

(105)

Following the prediction of the velocity and background pressure at the $(n+1)$st time level, the density is corrected

$$\frac{\rho_{ijk}^{(n+1)} - \frac{1}{2} \left( \rho_{ijk}^n + \rho_{ijk}^{(n+1)} \right)}{\frac{1}{2} \delta t} + (\mathbf{u} \cdot \nabla \rho)_{ijk}^{(n+1)} = -\rho_{ijk}^{(n+1)} (\nabla \cdot \mathbf{u})_{ijk}^{(n+1)}$$

(106)

The species conservation equations are differenced the same way

$$\frac{(\rho Y_l)_{ijk}^{(n+1)} - (\rho Y_l)_{ijk}^n}{\delta t} + (\mathbf{u} \cdot \nabla (\rho Y_l))_{ijk}^n = -(\rho Y_l)_{ijk}^n (\nabla \cdot \mathbf{u})_{ijk}^n + (\nabla \cdot \rho D \nabla Y_l)_{ijk}^n + W_{ijk}^{n}$$

(107)

at the predictor step, and

$$\frac{(\rho Y_l)_{ijk}^{(n+1)} - \frac{1}{2} \left( (\rho Y_l)_{ijk}^n + (\rho Y_l)_{ijk}^{(n+1)} \right)}{\frac{1}{2} \delta t} + (\mathbf{u} \cdot \nabla (\rho Y_l))_{ijk}^{(n+1)} = -(\rho Y_l)_{ijk}^{(n+1)} (\nabla \cdot \mathbf{u})_{ijk}^{(n+1)} + (\nabla \cdot \rho D \nabla Y_l)_{ijk}^{(n+1)} + W_{ijk}^{n}$$

(108)

at the corrector step.

The convective terms are written as upwind-biased differences in the predictor step and downwind-biased differences in the corrector step. In the expressions to follow, the symbol ± means + in the predictor step and − in the corrector step. The opposite is true for ⋅.

$$\mathbf{u} \cdot \nabla \rho_{ijk} = \frac{1}{2} \left( \begin{array}{c} \epsilon_u u_{i,j,k} \rho_{i+1,j,k} - \rho_{i,j,k} + \epsilon_u u_{i-1,j,k} \rho_{i,j,k} - \rho_{i-1,j,k} \\
\epsilon_v v_{i,j,k} \rho_{i,j+1,k} - \rho_{i,j,k} + \epsilon_v v_{i,j-1,k} \rho_{i,j,k} - \rho_{i,j-1,k} \\
\epsilon_w w_{i,j,k} \rho_{i,j,k+1} - \rho_{i,j,k} + \epsilon_w w_{i,j,k-1} \rho_{i,j,k} - \rho_{i,j,k-1} \end{array} \right)$$

(109)

$$\mathbf{u} \cdot \nabla \rho Y_l_{ijk} = \frac{1}{2} \left( \begin{array}{c} \epsilon_u u_{i,j,k} (\rho Y_l)_{i+1,j,k} - (\rho Y_l)_{i,j,k} + \epsilon_u u_{i-1,j,k} (\rho Y_l)_{i,j,k} - (\rho Y_l)_{i-1,j,k} \\
\epsilon_v v_{i,j,k} (\rho Y_l)_{i,j+1,k} - (\rho Y_l)_{i,j,k} + \epsilon_v v_{i,j-1,k} (\rho Y_l)_{i,j,k} - (\rho Y_l)_{i,j-1,k} \\
\epsilon_w w_{i,j,k} (\rho Y_l)_{i,j,k+1} - (\rho Y_l)_{i,j,k} + \epsilon_w w_{i,j,k-1} (\rho Y_l)_{i,j,k} - (\rho Y_l)_{i,j,k-1} \end{array} \right)$$

(110)

Note that without the inclusion of the ε’s, these are simple central difference approximations. The ε’s are local CFL numbers, $\epsilon_u = u \delta t / \delta x$, $\epsilon_v = v \delta t / \delta y$, and $\epsilon_w = w \delta t / \delta z$, where the velocity components are those.
that immediately follow. Their role is to bias the differencing upwind. Where the local CFL number is near unity, the difference becomes nearly fully upwinded. Where the local CFL number is much less than unity, the differencing is more centralized [30].

The divergence in both the predictor and corrector step is discretized

$$\nabla \cdot \mathbf{u}_{ijk} = \frac{\gamma - 1}{\gamma p_0} \left( q_{ijk}'' + (\nabla \cdot k \nabla T)_{ijk} + \sum_l (\nabla \cdot T c_{pl} \rho D \nabla Y_l)_{ijk} - \frac{1}{\gamma - 1} \left( \frac{d p_0}{dt} \right) \right)$$  \hspace{1cm} (111)

The thermal and material diffusion terms are pure central differences, with no upwind or downwind bias, thus they are differenced the same way in both the predictor and corrector steps

$$\nabla \cdot k \nabla T_{ijk} = \frac{1}{\delta x} \left[ k_{i+\frac{1}{2}jk} T_{i+1,jk} - T_{ijk} - k_{i-\frac{1}{2}jk} T_{i-1,jk} \right] + \frac{1}{\delta y} \left[ k_{i,j+\frac{1}{2}jk} T_{i,j+1,ik} - T_{ijk} - k_{i,j-\frac{1}{2}jk} T_{i,j-1,ik} \right] + \frac{1}{\delta z} \left[ k_{ij,k+\frac{1}{2}jk} T_{ij,k+1,ik} - T_{ijk} - k_{ij,k-\frac{1}{2}jk} T_{ij,k-1,ik} \right]$$  \hspace{1cm} (112)

$$\nabla \cdot c_{pl} T \rho D \nabla Y_l_{ijk} = \frac{c_{pl}}{\delta x} \left[ T_{i+\frac{1}{2},jk} \rho D_{i+\frac{1}{2},jk} Y_{i+1,ijk} - Y_{ij} - T_{i-\frac{1}{2},jk} \rho D_{i-\frac{1}{2},jk} Y_{i-1,ijk} \right] + \frac{c_{pl}}{\delta y} \left[ T_{i,j+\frac{1}{2},jk} \rho D_{i,j+\frac{1}{2},jk} Y_{i,j+1,ijk} - Y_{ij} - T_{i,j-\frac{1}{2},jk} \rho D_{i,j-\frac{1}{2},jk} Y_{i,j-1,ijk} \right] + \frac{c_{pl}}{\delta z} \left[ T_{i,j,k+\frac{1}{2},jk} \rho D_{i,j,k+\frac{1}{2},jk} Y_{i,j,k+1,ijk} - Y_{ij} - T_{i,j,k-\frac{1}{2},jk} \rho D_{i,j,k-\frac{1}{2},jk} Y_{i,j,k-1,ijk} \right]$$  \hspace{1cm} (113)

The temperature is extracted from the density via the equation of state

$$T_{ijk} = \frac{P_0}{\rho_{ijk} \mathcal{R} \sum_{l=0}^{N} (Y_{i,l,ijk}/M_l)}$$  \hspace{1cm} (115)

Because only species 1 through $N$ are explicitly computed, the summation is rewritten

$$\sum_{l=0}^{N} \frac{Y_{i,l,ijk}}{M_l} = \frac{1}{M_0} + \sum_{l=1}^{N} \left( \frac{1}{M_l} - \frac{1}{M_0} \right) Y_{i,l}$$  \hspace{1cm} (116)

In isothermal calculations involving multiple species, the density can be extracted from the average molecular weight

$$\rho_{ijk} = \frac{P_0}{T_0 \mathcal{R} \sum_{l=0}^{N} Y_{i,l,ijk}/M_l}$$  \hspace{1cm} (117)

Again, because only species 1 through $N$ are explicitly computed, this expression can be written

$$\rho_{ijk} = \frac{M_0 P_0}{T_0 \mathcal{R}} + \sum_{l=1}^{N} \left( \frac{1}{M_l} - \frac{1}{M_0} \right) (\rho Y_{i,l,ijk})$$  \hspace{1cm} (118)
7.5.2 Heat Release Rate (Mixture Fraction)

Energy from combustion is released into those grid cells through which the flame sheet \((Z = Z_f)\) passes. The analytical expression for the heat release rate per unit area of flame sheet is

\[
\dot{q}'' = \Delta H \frac{dY_O}{dZ} \bigg|_{Z < Z_f} (\rho D) \nabla Z \cdot \mathbf{n}
\]  

(119)

where \(\mathbf{n}\) is the outward facing unit normal. Note that both \(dY_O/dZ\) and \(\nabla Z \cdot \mathbf{n}\) are negative. To convert the analytical expression for the HRR per unit area into a discretized expression for the HRR per unit volume, all cells through which the flame sheet passes must be identified. Then, the normal derivative of \(Z\) must be computed component by component. For example, suppose in grid cell \(ijk\) the flame sheet passes between the cell center and the cell face pointing in the positive \(x\) direction, as determined from a linear interpolation of \(Z\) at the center of cell \(ijk\) and \(Z\) at the center of cell \(i+1, jk\). The contribution to the heat release rate per unit volume in the \(ijk\)th cell from the gradient of \(Z\) in the \(x\) direction is

\[
\dot{q}'''_{ijk} = -\Delta H \frac{dY_O}{dZ} \bigg|_{Z < Z_f} (\rho D) \frac{\partial Z}{\partial x} \frac{\partial Y_O}{\partial x} \frac{\partial Y_F}{\partial z} \delta x \delta y \delta z
\]  

(120)

A similar expression can be derived for all other possible cuts through the cell by the flame sheet.

7.5.3 Heat Release Rate (Finite-Rate Reaction)

In a DNS calculation (usually), a one-step, finite-rate reaction of a hydrocarbon fuel is assumed

\[
\nu C_x H_y + \nu O_2 \rightarrow \nu CO_2 + \nu H_2 O
\]  

(121)

For each grid cell, at the start of a time step where \(t = t^n\) and \(Y^n_{C_x H_y, ijk} \equiv Y_F(t^n)\) and \(Y^n_{O_2, ijk} \equiv Y_O(t^n)\), the following ODE is solved numerically with a 2nd order Runge-Kutta scheme

\[
\frac{dY_F}{dt} = -\frac{B \rho^n_{ijk}}{M_F} \frac{Y_F(t)^a Y_O(t)^b e^{-E/RT_{ijk}}}{M_O M_F} 
\]  

(122)

\[
\frac{dY_O}{dt} = -\frac{\nu O M_O}{\nu F M_F} \frac{dY_F}{dt}
\]  

(123)

The temperature \(T_{ijk}\) and density \(\rho_{ijk}\) are fixed at their values at time \(t^n\) and the ODE is iterated from \(t^n\) to \(t^{n+1}\) in about 100 time steps. The pre-exponential factor \(B\), the activation energy \(E\), and the exponents \(a\) and \(b\) are input parameters. The average heat release rate over the entire time step is given by

\[
\dot{q}_{ijk}^{n+1} = \Delta H \frac{Y_F(t^n) - Y_F(t^{n+1})}{\delta t}
\]  

(124)

where \(\delta t = t^{n+1} - t^n\). The species mass fractions are adjusted at this point in the calculation (before the convection and diffusion update)

\[
Y_{i,j,k}^n = Y_i(t^n) - \frac{\nu_i M_i}{\nu_F M_F} (Y_F(t^n) - Y_F(t^{n+1}))
\]  

(125)
7.6 Thermal Radiation

The discretized RTE is derived by integrating equation (45) over the grid cell $ijk$ and the control angle $\delta \Omega^l$:

$$
\int_{\Omega^l} \int_{V_{ijk}} s \cdot \nabla I(x,s) dV d\Omega = \int_{\Omega^l} \int_{V_{ijk}} \kappa(x) [I_b(x) - I(x,s)] dV d\Omega
$$

(126)

The volume integral on the left hand side is replaced by a surface integral over the cell faces using the divergence theorem. Assuming that the radiation intensity $I(x,s)$ is constant on each of the cell faces, the surface integral can be approximated by a sum over the cell faces. Assuming further that $I(x,s)$ is constant within the volume $V_{ijk}$ and over the angle $\delta \Omega^l$ we obtain

$$
\sum_{m=1}^{6} A_m I_{m}^l \int_{\Omega^l} (s \cdot n_m) d\Omega = \kappa_{ijk} \left[ I_{b,ijk} - I_{l,ijk} \right] V_{ijk} \delta \Omega^l
$$

(127)

where

$I_{l,ijk}$ radiant intensity in direction $l$
$I_m^l$ radiant intensity at cell face $m$
$I_{b,ijk}$ radiant blackbody Intensity in cell
$\delta \Omega^l$ solid angle corresponding to direction $l$
$V_{ijk}$ volume of cell $ijk$
$A_m$ area of cell face $m$
$n_m$ unit normal vector of the cell face $m$

It must be noticed, that while the intensity is assumed constant within the angle $\delta \Omega^l$, its direction covers the angle $\delta \Omega^l$ exactly.

In Cartesian coordinates$^5$, the normal vectors $n_m$ are the base vectors of the coordinate system and the integrals over the solid angle do not depend on the physical coordinate, but the direction only. The intensities on the cell boundaries, $I_m^l$, are calculated using a first order upwind scheme. If the physical space is swept in the direction $s^l$, the intensity $I_{l,ijk}$ can be directly solved from an algebraic equation. This makes the numerical solution of the FVM very fast. Iterations are needed only to account for the reflective boundaries. However, this is seldom necessary in practice, because of the small time step set by the flow solver.

The spatial discretization for the RTE solver is the same as for the fluid solver. The coordinate system used to discretize the solid angle is shown in Figure 3. The discretization of the solid angle is done by dividing first the polar angle, $\theta$, into $N_\theta$ bands, where $N_\theta$ is an even integer. Each $\theta$-band is then divided into $N_\phi(\theta)$ parts in the azimuthal ($\phi$) direction. $N_\phi(\theta)$ must be divisible by 4. The numbers $N_\theta$ and $N_\phi(\theta)$ are chosen to give the total number of angles $N_\Omega$ as close to the value defined by the user as possible. $N_\Omega$ is calculated as

$$
N_\Omega = \sum_{i=1}^{N_\theta} N_\phi(\theta_i)
$$

(128)

The distribution of the angles is based on empirical rules that try to produce equal solid angles $\delta \Omega^l = 4\pi/N_\Omega$. The number of $\theta$-bands is

$$
N_\theta = 1.17 \frac{N_\Omega^{1/2.26}}{\Omega}
$$

(129)

rounded to the nearest even integer. The number of $\phi$-angles on each band is

$$
N_\phi(\theta) = \max \left\{ 4, 0.5N_\Omega \left[ \cos(\theta^-) - \cos(\theta^+) \right] \right\}
$$

(130)

$^5$In the axisymmetric case equation (127) becomes a little bit more complicated, as the cell face normal vectors $n_m$ are not always constant. However, the computational efficiency can still be retained.
rounded to the nearest integer that is divisible by 4. \( \theta^- \) and \( \theta^+ \) are the lower and upper bounds of of the \( \theta \)-band, respectively. The discretization is symmetric with respect to the planes \( x = 0 \), \( y = 0 \), and \( z = 0 \). This symmetry has three important benefits: First, it avoids the problems caused by the fact that first order upwind scheme, used to calculate intensities on the cell boundaries, is more diffusive in non-axial directions than axial. Second, the treatment of the mirror boundaries becomes very simple, as will be shown later. Third, it avoids so called “overhang” situations, where \( \mathbf{s} \cdot \mathbf{i} \), \( \mathbf{s} \cdot \mathbf{j} \) or \( \mathbf{s} \cdot \mathbf{k} \) changes sign inside the control angle. These “overhangs” would make the resulting system of linear equations more complicated.

In the axially symmetric case these “overhangs” can not be avoided, and a special treatment, developed by Murthy and Mathur [31], is applied. In these cases \( N_\phi (\theta_i) \) is kept constant, and the total number of angles is \( N_\Omega = N_\theta \times N_\phi \). In addition, the angle of the vertical slice of the cylinder is chosen to be same as \( \delta \phi \).

The cell face intensities, \( I^l_{ij} \), appearing on the left hand side of (127) are calculated using a first order upwind scheme. Consider, for example, a control angle having a direction vector \( \mathbf{s} \). If the radiation is traveling in the positive \( x \)-direction, \( i.e. \mathbf{s} \cdot \mathbf{i} \geq 0 \), the intensity on the upwind side, \( I^l_{i-1,jk} \), and the intensity on the downwind side is the intensity in the cell itself \( I^l_{ij} \).

On a rectilinear grid, the normal vectors \( \mathbf{n}_m \) are the base vectors of the coordinate system and the integrals over the solid angle can be calculated analytically. Equation (127) can be simplified

\[
d^l_{ijk} I^l_{ijk} = a^l_{ix} I^l_{ix} + a^l_{iy} I^l_{iy} + a^l_{iz} I^l_{iz} + b^l_{ijk}
\]

where

\[
a^l_{ij} = A_x |D^l_x| + A_y |D^l_y| + A_z |D^l_z| + \kappa_{ijk} V_{ijk} \delta \Omega^l
\]

\[
a^l_{ix} = A_x |D^l_x|
\]

\[
a^l_{iy} = A_y |D^l_y|
\]

\[
a^l_{iz} = A_z |D^l_z|
\]

\[
b^l_{ijk} = \kappa_{ijk} I^l_{b,ijk} V_{ijk} \delta \Omega^l
\]

\[
\delta \Omega = \int_{\Omega^l} d\Omega = \int_{\delta \phi} \int_{\delta \theta} \sin \theta \, d\theta \, d\phi
\]
\[ D_x' = \int_{\Omega'} (s' \cdot i) d\Omega \\
= \int_{\delta \phi} \int_{\delta \theta} (s' \cdot i) \sin \theta \ d\theta d\phi \\
= \int_{\delta \phi} \int_{\delta \theta} \cos \phi \sin \theta \sin \theta \ d\theta d\phi \\
= \frac{1}{2} (\sin \phi^+ - \sin \phi^-) \left[ \Delta \theta - (\cos \theta^+ \sin \theta^+ - \cos \theta^- \sin \theta^-) \right] \]

\[ D_y' = \int_{\Omega'} (s' \cdot j) d\Omega \\
= \int_{\delta \phi} \int_{\delta \theta} \sin \phi \sin \theta \sin \theta \ d\theta d\phi \\
= \frac{1}{2} (\cos \phi^- - \cos \phi^+) \left[ \Delta \theta - (\cos \theta^+ \sin \theta^+ - \cos \theta^- \sin \theta^-) \right] \]

\[ D_z' = \int_{\Omega'} (s' \cdot k) d\Omega \\
= \int_{\delta \phi} \int_{\delta \theta} \cos \theta \sin \theta \sin \theta \ d\theta d\phi \\
= \frac{1}{2} \Delta \phi \left[ (\sin \theta^+)^2 - (\sin \theta^-)^2 \right] \]

Here \( i, j \) and \( k \) are the base vectors of the Cartesian coordinate system. \( \theta^+, \theta^-, \phi^+ \) and \( \phi^- \) are the upper and lower boundaries of the control angle in the polar and azimuthal directions, respectively, and \( \Delta \theta = \theta^+ - \theta^- \) and \( \Delta \phi = \phi^+ - \phi^- \). The solution method of (131) is based on an explicit marching sequence [32]. The marching direction depends on the propagation direction of the radiation intensity. As the marching is done in the “downwind” direction, the “upwind” intensities in all three spatial directions are known, and the intensity \( I_{ijk}^l \) can be solved directly. Iterations may be needed only with the reflective walls and optically thick situations. Currently, no iterations are made.

The boundary condition on a solid wall is given as

\[ I_{w}^l = \varepsilon \frac{\sigma T_w^4}{\pi} + \frac{1 - \varepsilon}{\pi} \sum_{D_n' < 0} I_{w}^l' |D_n'| \]

where \( D_n' = \int_{\Omega'} (s \cdot n_w) d\Omega \). The constraint \( D_n' < 0 \) means that only the “incoming” directions are taken into account when calculating the reflection. The radiative heat flux on the wall is

\[ q_w = \sum_{l=1}^{N_\Omega} I_w^l \int_{\delta \Omega} (s \cdot n_w) d\Omega = \sum_{l=1}^{N_\Omega} I_w^l D_n^l \]

where the coefficients \( D_n^l \) are equal to \( \pm D_{x}^l, \pm D_{y}^l \) or \( \pm D_{z}^l \), and can be calculated for each wall element beforehand.

The open boundaries are treated as black walls, where the incoming intensity is the black body intensity of the ambient temperature. On mirror boundaries the intensities leaving the wall are calculated from the incoming intensities using a predefined connection matrix.

\[ I_{w,ijk}^l = I_{ij}' \]

Computationally intensive integration over all the incoming directions is avoided by keeping the solid angle discretization symmetric \( x, y \) and \( z \) planes. The connection matrix associates one incoming direction \( l' \) to each mirrored direction on each wall cell.
The local incident radiation intensity is

\[ U_{ijk} = \sum_{l=1}^{N_\Omega} I_{ijk}^l \delta \Omega^l \]  

(144)

### 7.7 Thermal and Material Boundary Conditions

Four types of thermal boundary conditions are applied at solid surfaces. The first, and simplest, is an adiabatic boundary condition that states that there is no temperature gradient normal to the surface. It is implemented by assigning to the grid cell that is embedded in the solid (the ghost cell) the same temperature as the first cell in the gas (the gas cell).

The second type of boundary condition is where the solid surface has a prescribed temperature (usually this prescribed temperature is a function of time).

The third type of boundary condition assumes the solid to be thermally-thin. The surface temperature is \( T_{\text{surf}} \) where

\[ \alpha \delta_n \frac{\partial T_{\text{surf}}}{\partial n} = q_{\text{conv}}' \]

(145)

where \( T_{\text{surf}} \) is the wall temperature, \( \bar{n} \) is the time step used when updating the thermal boundary conditions (usually greater than the hydrodynamic time step \( \delta t \)), and \( \alpha, r, \delta \) are the input density, specific heat and thickness of the wall. In a DNS calculation where the boundary layer is resolved, the convective flux to the wall is given by

\[ q_{\text{conv}}'' = -k \frac{T_{\text{gas}} - T_{\text{w}}}{\delta n/2} \]  

(146)

where \( \delta n \) is the size of a grid cell in the normal direction to the wall. In an LES calculation where the boundary layer is not resolved,

\[ q_{\text{conv}}'' = C |T_{\text{gas}} - T_{\text{w}}|^\delta (T_{\text{gas}} - T_{\text{w}}) \quad \text{W/m}^2 \]  

(147)

where \( C \) is an empirical coefficient (0.95 for vertical surface; 1.43 for horizontal), and \( T_{\text{gas}} \) is the temperature of the gas in the cell bordering the wall.

The fourth type of thermal boundary condition is for a thermally-thick solid. In this case, a one-dimensional heat transfer calculation is performed at each boundary cell designated as thermally-thick. The width of the solid \( \delta \) is partitioned into \( N \) cells, clustered near the front face. The cell boundaries are located at points \( x_i \)

\[ x_i = f(\xi) = \delta \xi + \frac{1 - \xi}{\delta \xi} \xi_i \]  

(148)

where \( 0 \leq i \leq N, \xi_i = i \delta \xi, \delta \xi = \delta / N, \) and \( 0 < s \leq 1 \) is a measure of the degree of clustering of the cells at the front face. The width of each cell is \( \delta x_i = \xi_i^{\prime}(\xi_i^{\prime - 1}) \delta \xi, 1 \leq i \leq N \) where \( \xi_i^{\prime - 1} = (i - \frac{1}{2}) \delta \xi. \) The temperature at the center of the \( i \)th cell is denoted \( T_{s_i}. \) These temperatures are updated in time using an implicit Crank-Nicholson scheme

\[
\frac{T_{s,i+1}^{n+1} - T_{s,i}^{n}}{\delta t} = \frac{\Delta}{2\delta x_i} \left( \frac{T_{s,i+1}^{n+1} - T_{s,i}^{n}}{\delta x_{s,i+\frac{1}{2}}} - \frac{T_{s,i+1}^{n} - T_{s,i}^{n-1}}{\delta x_{s,i-\frac{1}{2}}} + \frac{T_{s,i+1}^{n+1} - T_{s,i}^{n+1}}{\delta x_{s,i+\frac{1}{2}}} - \frac{T_{s,i-1}^{n+1} - T_{s,i}^{n+1}}{\delta x_{s,i-\frac{1}{2}}} \right) \]  

(149)

for \( 1 \leq i \leq N. \) The boundary condition is discretized

\[ -k \frac{T_{s,i+\frac{1}{2}}^{n+1} - T_{s,i}^{n+1}}{\delta x_{s,i}^{\frac{1}{2}}} = q_{s,i}'' - 4 \varepsilon \sigma T_{s,i}^{n+1} \left( T_{s,i+\frac{1}{2}}^{n+1} - T_{s,i}^{n+1} \right) \]  

(150)
where \( T_{s, \frac{1}{2}} = (T_{s, 1} + T_{s, 0})/2 \) is the temperature at the front face. Notice that the radiative emission term has been linearized

\[
T_{s, \frac{1}{2}}^{(n+1)^4} - T_{s, \frac{1}{2}}^{n^4} \approx 4\left(\frac{T_{s, \frac{1}{2}}^{n+1} - T_{s, \frac{1}{2}}^n}{\frac{T_{s, \frac{1}{2}}^n - T_{s, \frac{1}{2}}^{n+1}}{4}}\right)
\]

(151)

The wall temperature is defined \( T_w \equiv T_{s, \frac{1}{2}} = (T_{s, 0} + T_{s, 1})/2 \).

Regardless of how the wall temperature is determined, there are two ways of coupling the wall temperature with the fluid calculation. Gas phase temperatures are defined at cell centers; the wall is defined at the boundary of the bordering gas phase cell and a “ghost” cell inside the wall. As far as the gas phase calculation is concerned, the normal temperature gradient at the wall is expressed in terms of the temperature difference between the “gas” cell and the “ghost” cell. The wall temperature affects the gas phase calculation through the prescription of the ghost cell temperature. This ghost cell temperature has no physical meaning on its own. Only the difference between ghost and gas cell temperatures matters, for this defines the heat transfer to the wall. In a DNS calculation, the wall temperature is assumed to be an average of the ghost cell temperature and the temperature of the first cell in the gas, thus the ghost cell temperature is defined

\[
T_{\text{ghost}} = 2T_w - T_{\text{gas}}
\]

(152)

For an LES calculation, the heat lost to the boundary is equated with an empirical expression

\[
k \frac{T_{\text{gas}} - T_{\text{ghost}}}{\delta n} = C |T_{\text{gas}} - T_w|^\frac{1}{\gamma} (T_{\text{gas}} - T_w)
\]

(153)

where \( \delta n \) is the distance between the center of the ghost cell and the center of the gas cell. This equation is solved for \( T_{\text{ghost}} \), so that when the conservation equations are updated, the amount of heat lost to the wall is equivalent to the empirical expression on the right hand side. Note that \( T_{\text{ghost}} \) is purely a numerical construct. It does not represent the temperature within the wall, but rather establishes a temperature gradient at the wall consistent with the empirical correlation.

At solid walls there is no transfer of mass, thus the boundary condition for the \( l \)th species at a wall is simply

\[
Y_{l, \text{ghost}} = Y_{l, \text{gas}}
\]

(154)

where the subscripts “ghost” and “gas” are the same as above since the mass fraction, like temperature, is defined at cell centers. At forced flow boundaries either the mass fraction \( Y_{l, w} \) or the mass flux \( \dot{m}_{l, w} \) of species \( l \) may be prescribed. Then the ghost cell mass fraction can be derived because, as with temperature, the normal gradient of mass fraction is needed in the gas phase calculation. For cases where the mass fraction is prescribed

\[
Y_{l, \text{ghost}} = 2Y_{l, w} - Y_{l, \text{gas}}
\]

(155)

For cases where the mass flux is prescribed, the following equation must be solved iteratively

\[
\dot{m}_{l, w} = u_n \frac{\rho_{\text{ghost}} Y_{l, \text{ghost}} + \rho_{\text{gas}} Y_{l, \text{gas}}}{2} - \rho D \frac{Y_{l, \text{gas}} - Y_{l, \text{ghost}}}{\delta n} \mp \frac{\delta t u_n^2}{2} \frac{\rho_{\text{gas}} Y_{l, \text{gas}} - \rho_{\text{ghost}} Y_{l, \text{ghost}}}{\delta n}
\]

(156)

where \( \dot{m}_{l, w} \) is the mass flux of species \( l \) per unit area, \( u_n \) is the normal component of velocity at the wall pointing into the flow domain, and \( \delta n \) is the distance between the center of the ghost cell and the center of the gas cell. Notice that the last term on the right hand side is subtracted at the predictor step and added at the corrector step, consistent with the biased upwinding introduced earlier.

Once the temperature and species mass fractions have been defined in the ghost cell, the density in the ghost cell is computed from the equation of state

\[
\rho_{\text{ghost}} = \frac{\rho_0}{\mathcal{R} T_{\text{ghost}} \sum_l (Y_{l, \text{ghost}} M_l)}
\]

(157)
7.8 The Momentum Equation

The three components of the momentum equation are

\[
\begin{align*}
\frac{\partial u}{\partial t} + F_x + \frac{\partial g}{\partial x} &= 0 ; \quad F_x = w \omega_y - v \omega_x - \frac{1}{\rho} \left( f_x + \frac{\partial \tau_{xx}}{\partial x} + \frac{\partial \tau_{xy}}{\partial y} + \frac{\partial \tau_{xz}}{\partial z} \right) \\
\frac{\partial v}{\partial t} + F_y + \frac{\partial g}{\partial y} &= 0 ; \quad F_y = u \omega_z - w \omega_y - \frac{1}{\rho} \left( f_y + \frac{\partial \tau_{yx}}{\partial x} + \frac{\partial \tau_{yy}}{\partial y} + \frac{\partial \tau_{yz}}{\partial z} \right) \\
\frac{\partial w}{\partial t} + F_z + \frac{\partial g}{\partial z} &= 0 ; \quad F_z = v \omega_x - u \omega_z - \frac{1}{\rho} \left( f_z + \frac{\partial \tau_{zx}}{\partial x} + \frac{\partial \tau_{zy}}{\partial y} + \frac{\partial \tau_{zz}}{\partial z} \right)
\end{align*}
\]

The spatial discretization of the momentum equations take the form

\[
\begin{align*}
\frac{\partial u}{\partial t} + F_{x,j,k} + \frac{\mathcal{H}_{i+1,j,k} - \mathcal{H}_{i,j,k}}{\delta x} &= 0 \\
\frac{\partial v}{\partial t} + F_{y,j,k} + \frac{\mathcal{H}_{i,j+1,k} - \mathcal{H}_{i,j,k}}{\delta y} &= 0 \\
\frac{\partial w}{\partial t} + F_{z,j,k} + \frac{\mathcal{H}_{i,j,k+1} - \mathcal{H}_{i,j,k}}{\delta z} &= 0
\end{align*}
\]

where \( \mathcal{H}_{i,j,k} \) is taken at center of cell \( i,j,k \), \( u_{i,j,k} \) and \( F_{x,j,k} \) are taken at the side of the cell facing in the forward \( x \) direction, \( v_{i,j,k} \) and \( F_{y,j,k} \) at the side facing in the forward \( y \) direction, and \( w_{i,j,k} \) and \( F_{z,j,k} \) at the side facing in the forward \( z \) (vertical) direction. In the definitions to follow, the components of the vorticity \( (\omega_x, \omega_y, \omega_z) \) are located at cell edges pointing in the \( x, y \) and \( z \) directions, respectively. The same is true for the off-diagonal terms of the viscous stress tensor: \( \tau_{yx} = \tau_{yz}, \tau_{z} = \tau_{zx}, \) and \( \tau_{xy} = \tau_{yx} \). The diagonal components of the stress tensor \( \tau_{xx}, \tau_{yy}, \) and \( \tau_{zz} \); the external force components \( (f_x, f_y, f_z) \); and the upwinding bias terms \( \varepsilon_u, \varepsilon_v, \) and \( \varepsilon_w \) are located at the respective cell faces.

\[
\begin{align*}
F_{x,j,k} &= \left( \frac{1}{2} \varepsilon_u \omega_{i+1,j,k} - \frac{1}{2} \varepsilon_u \omega_{i-1,j,k} \right) + \frac{1}{2} \varepsilon_u \omega_{i+1,j-1,k} - \frac{1}{2} \varepsilon_u \omega_{i-1,j-1,k} \right) + \frac{1}{2} \varepsilon_u \omega_{i,j+1,k} - \frac{1}{2} \varepsilon_u \omega_{i,j-1,k} \right) + \frac{1}{2} \varepsilon_u \omega_{i,j,k+1} - \frac{1}{2} \varepsilon_u \omega_{i,j,k-1} \right) \right)
\end{align*}
\]

\[
\begin{align*}
F_{y,j,k} &= \left( \frac{1}{2} \varepsilon_u \omega_{i+1,j,k} - \frac{1}{2} \varepsilon_u \omega_{i-1,j,k} \right) + \frac{1}{2} \varepsilon_u \omega_{i+1,j+1,k} - \frac{1}{2} \varepsilon_u \omega_{i-1,j+1,k} \right) + \frac{1}{2} \varepsilon_u \omega_{i,j+1,k} - \frac{1}{2} \varepsilon_u \omega_{i,j-1,k} \right) + \frac{1}{2} \varepsilon_u \omega_{i,j,k+1} - \frac{1}{2} \varepsilon_u \omega_{i,j,k-1} \right) \right)
\end{align*}
\]

\[
\begin{align*}
F_{z,j,k} &= \left( \frac{1}{2} \varepsilon_u \omega_{i+1,j,k} - \frac{1}{2} \varepsilon_u \omega_{i-1,j,k} \right) + \frac{1}{2} \varepsilon_u \omega_{i+1,j+1,k} - \frac{1}{2} \varepsilon_u \omega_{i-1,j+1,k} \right) + \frac{1}{2} \varepsilon_u \omega_{i,j+1,k} - \frac{1}{2} \varepsilon_u \omega_{i,j-1,k} \right) + \frac{1}{2} \varepsilon_u \omega_{i,j,k+1} - \frac{1}{2} \varepsilon_u \omega_{i,j,k-1} \right) \right)
\end{align*}
\]

\[
\begin{align*}
\omega_{x,j,k} &= \frac{w_{i,j+1,k} - w_{i,j,k}}{\delta y} - \frac{v_{i,j+1,k+1} - v_{i,j,k+1} - v_{i,j,k}}{\delta z}
\end{align*}
\]
\[
\omega_{ij,k} = \frac{u_{ij,k+1} - u_{ij,k}}{\delta z} - \frac{W_{i+1,j,k} - W_{ij,k}}{\delta x} 
\]
\[
\omega_{i,j,k} = \frac{v_{i+1,j,k} - v_{ij,k}}{\delta y} - \frac{u_{i,j+1,k} - u_{ij,k}}{\delta x} 
\]
\[
\tau_{xx,ijk} = \mu_{ijk} \left( 2 \frac{u_{ijk} - u_{i-1,j,k}}{\delta x} - \frac{2}{3} (\nabla \cdot \mathbf{u})_{ijk} \right) = \mu_{ijk} \left( \frac{4}{3} (\nabla \cdot \mathbf{u})_{ijk} - 2 \frac{v_{ijk} - v_{ij,j-1,k}}{\delta y} - 2 \frac{W_{ij,k} - W_{ij,k-1}}{\delta z} \right) \tag{170}
\]
\[
\tau_{yy,ijk} = \mu_{ijk} \left( 2 \frac{v_{ijk} - v_{i,j-1,k}}{\delta y} - \frac{2}{3} (\nabla \cdot \mathbf{u})_{ijk} \right) = \mu_{ijk} \left( \frac{4}{3} (\nabla \cdot \mathbf{u})_{ijk} - 2 \frac{u_{ijk} - u_{i-1,j,k}}{\delta x} - 2 \frac{W_{ij,k} - W_{ij,k-1}}{\delta z} \right) \tag{171}
\]
\[
\tau_{zz,ijk} = \mu_{ijk} \left( 2 \frac{W_{ij,k} - W_{ij,k-1}}{\delta z} - \frac{2}{3} (\nabla \cdot \mathbf{u})_{ijk} \right) = \mu_{ijk} \left( \frac{4}{3} (\nabla \cdot \mathbf{u})_{ijk} - 2 \frac{u_{ijk} - u_{i-1,j,k}}{\delta x} - 2 \frac{v_{ijk} - v_{ij,j-1,k}}{\delta y} \right) \tag{172}
\]
\[
\tau_{xy,ijk} = \tau_{yx,ijk} = \mu_{i+\frac{1}{2},j+\frac{1}{2},k} \left( \frac{u_{ij,k+1} - u_{ij,k}}{\delta y} + \frac{v_{i+1,j,k} - v_{ij,k}}{\delta x} \right) \tag{173}
\]
\[
\tau_{xz,ijk} = \tau_{zx,ijk} = \mu_{i+\frac{1}{2},j+\frac{1}{2},k} \left( \frac{u_{ij,k+1} - u_{ij,k}}{\delta z} + \frac{W_{i+1,j,k} - W_{ij,k}}{\delta x} \right) \tag{174}
\]
\[
\tau_{yz,ijk} = \tau_{zy,ijk} = \mu_{i+\frac{1}{2},j+\frac{1}{2},k} \left( \frac{v_{ij,k+1} - v_{ij,k}}{\delta z} + \frac{W_{ij,j+1,k} - W_{ij,k}}{\delta y} \right) \tag{175}
\]
\[
\varepsilon_u = \frac{u \delta t}{\delta x} \tag{176}
\]
\[
\varepsilon_v = \frac{v \delta t}{\delta y} \tag{177}
\]
\[
\varepsilon_w = \frac{w \delta t}{\delta z} \tag{178}
\]

The variables \( \varepsilon_u, \varepsilon_v, \) and \( \varepsilon_w \) are local CFL numbers evaluated at the same locations as the velocity component immediately following them, and serve to bias the differencing of the convective terms in the upwind direction. The subscript \( i + \frac{1}{2} \) indicates that a variable is an average of its values at the \( i \)th and the \( (i+1) \)th cell. The divergence defined in Eq. (111) is identically equal to the divergence defined by
\[
(\nabla \cdot \mathbf{u})_{ijk} = \frac{u_{ij,k} - u_{i-1,j,k}}{\delta x} + \frac{v_{ij,k} - v_{i,j-1,k}}{\delta y} + \frac{w_{ij,k} - W_{ij,k-1}}{\delta z} \tag{179}
\]

The equivalence of the two definitions of the divergence is a result of the form of the discretized equations, the time-stepping scheme, and the direct solution of the Poisson equation for the pressure.

### 7.8.1 Force Terms

The external force term components, in addition to including the effects of buoyancy, may also include the drag force from sprinkler droplets.

\[
f_{x,ijk} = \frac{1}{2} \sum \rho Cp \pi r_d^2 (u_d - u_{ijk}) |\mathbf{u}_d - \mathbf{u}| \frac{1}{\delta x \delta y \delta z} \left( \rho_{i+\frac{1}{2},j,k} - \rho_\infty \right) g_x \tag{180}
\]
\[
f_{y,ijk} = \frac{1}{2} \sum \rho Cp \pi r_d^2 (v_d - v_{ijk}) |\mathbf{u}_d - \mathbf{u}| \frac{1}{\delta x \delta y \delta z} \left( \rho_{i,j+\frac{1}{2},k} - \rho_\infty \right) g_y \tag{181}
\]
\[
f_{z,ijk} = \frac{1}{2} \sum \rho Cp \pi r_d^2 (w_d - w_{ijk}) |\mathbf{u}_d - \mathbf{u}| \frac{1}{\delta x \delta y \delta z} \left( \rho_{i,j+k+\frac{1}{2}} - \rho_\infty \right) g_z \tag{182}
\]

where \( \mathbf{g} = (g_x, g_y, g_z) \) is the gravity vector, \( r_d \) is the radius of a droplet, \( \mathbf{u} = (u_d, v_d, w_d) \) the velocity of a droplet, \( C_p \) the drag coefficient, and \( \delta x \delta y \delta z \) the volume of the \( ijk \)th cell. The summations represent all droplets within a grid cell centered about the \( x, y, \) and \( z \) faces of a grid cell respectively.
7.8.2  Time Step

The time step is determined by the CFL condition, and in cases of high viscosity, a parabolic stability criterion typical of explicit second order accurate schemes

$$\delta t < \min \left( \frac{\delta x}{u_{i,j,k}}, \frac{\delta y}{v_{i,j,k}}, \frac{\delta z}{w_{i,j,k}}, \frac{\rho_{ijk} \delta x^2}{8 \mu_{ijk}}, \frac{\rho_{ijk} \delta y^2}{8 \mu_{ijk}}, \frac{\rho_{ijk} \delta z^2}{8 \mu_{ijk}} \right)$$  \hspace{1cm} (183)

The estimated velocities $u^{(n+1)e}$, $v^{(n+1)e}$ and $w^{(n+1)e}$ are tested at each time step to ensure that the above condition is satisfied. If it is not, then the time step is set to 0.8 of its allowed maximum value and the estimated velocities are recomputed (and checked again). The parabolic stability criterion is only invoked for a DNS calculation.

7.9  The Pressure Equation

The divergence of the momentum equation yields a Poisson equation for the pressure

$$\frac{\partial}{\partial t} \frac{\delta z}{\delta x} \left( \frac{\delta z}{\delta y} \left( \frac{\delta z}{\delta t} (\nabla \cdot \mathbf{u})_{i,j,k} \right) \right) = - \frac{F_{x,i,j,k} - F_{x,i+1,j,k}}{\delta x} - \frac{F_{z,i,j,k} - F_{z,i,j+1,k}}{\delta y} - \frac{F_{z,i,j,k} - F_{z,i+1,j,k}}{\delta z} - \frac{\partial}{\partial \mathbf{e}} (\nabla \cdot \mathbf{u})_{i,j,k}$$  \hspace{1cm} (184)

The lack of a superscript implies that all quantities are to be evaluated at the same time level. This elliptic partial differential equation is solved using a direct (non-iterative) FFT-based solver that is part of a library of routines for solving elliptic PDEs called CRAYFISHPAK [33]. To ensure that the divergence of the fluid is consistent with the definition given in Eq. (10), the time derivative of the divergence is defined

$$\frac{\partial}{\partial t} (\nabla \cdot \mathbf{u})_{i,j,k} = \frac{(\nabla \cdot \mathbf{u})_{i,j,k}^{(n+1)e} - (\nabla \cdot \mathbf{u})_{i,j,k}^{n}}{\delta t}$$  \hspace{1cm} (185)

at the predictor step, and then

$$\frac{\partial}{\partial t} (\nabla \cdot \mathbf{u})_{i,j,k} = \frac{2(\nabla \cdot \mathbf{u})_{i,j,k}^{(n+1)e} - (\nabla \cdot \mathbf{u})_{i,j,k}^{n} - (\nabla \cdot \mathbf{u})_{i,j,k}^{n}}{\delta t}$$  \hspace{1cm} (186)

at the corrector step. The discretization of the divergence was given in Eq. (111).

Direct Poisson solvers are most efficient if the domain is a rectangular region, although other geometries such as cylinders and spheres can be handled almost as easily. For these solvers, the no-flux condition (187) is simple to prescribe at external boundaries. For example, at the floor, $z = 0$, the Poisson solver is supplied with the Neumann boundary condition

$$\frac{\partial H_{j,i,k}}{\partial z} = -F_{z,i,j,k}$$  \hspace{1cm} (187)

However, many practical problems involve more complicated geometries. For building fires, doors and windows within multi-room enclosures are very important features of the simulations. These elements may be included in the overall domain as masked grid cells, but the no-flux condition (187) cannot be directly prescribed at the boundaries of these blocked cells. Fortunately, it is possible to exploit the relatively small changes in the pressure from one time step to the next to enforce the no-flux condition. At the start of a time step, the components of the convection/diffusion term $\mathbf{F}$ are computed at all cell faces that do not correspond to walls. At those cell faces that do correspond to solid walls, prescribe

$$F_n = -\frac{\partial H_n}{\partial n} + \beta u_n$$  \hspace{1cm} (188)
where \( F_n \) is the normal component of \( \mathbf{F} \) at the wall, and \( \beta \) is a relaxation factor empirically determined to be about 0.8 divided by the time step \( \delta t \). The asterisk indicates the most recent value of the pressure. Obviously, the pressure at this particular time step is not known until the Poisson equation is solved. Equation (188) asserts that following the solution of the Poisson equation for the pressure, the normal component of velocity \( u_n \) will be driven closer to zero according to

\[
\frac{\partial u_n}{\partial t} \approx -\beta u_n
\]  

(189)

This is approximate because the true value of the velocity time derivative depends on the solution of the pressure equation, but since the most recent estimate of pressure is used, the approximation is very good. Also, even though there are small errors in normal velocity at solid surfaces, the divergence of each blocked cell remains exactly zero for the duration of the calculation. In other words, the total flux into a given obstruction is always identically zero, and the error in normal velocity is usually at least several orders of magnitude smaller than the characteristic flow velocity. When implemented as part of a predictor-corrector updating scheme, the no-flux condition at solid surfaces is maintained remarkably well.

At open boundaries (say \( i = I \)), \( H \) is prescribed depending on whether the flow is incoming or outgoing

\[
H_{i+\frac{1}{2},jk} = \begin{cases} \frac{(u_{I,j,k} + v_{I,j-\frac{1}{2},k} + w_{I,j-\frac{1}{2},k})}{2} & u_{I,j,k} > 0 \\ 0 & u_{I,j,k} < 0 \end{cases}
\]

(190)

where \( I \) is the index of the last gas phase cell in the \( x \) direction and \( u_{I,j,k} \) is the \( x \) component of velocity at the boundary. The value of \( H \) in the ghost cell is

\[
H_{i+1,jk} = 2H_{i+\frac{1}{2},jk} - H_{i,jk}
\]

(191)

### 7.10 Particle Tracking

Thermal elements are introduced into the flow field as a means of introducing heat and as a way to visualize the flow. The position \( x_p \) of each thermal element is governed by the equations

\[
\frac{dx_p}{dt} = \mathbf{u}
\]

(192)

The thermal element positions are updated according to the same predictor-corrector scheme that is applied to the other flow quantities. Briefly, the position \( x_p \) of a given thermal element is updated according to the two step scheme

\[
\begin{align*}
    x_p^{(n+1)e} &= x_p^n + \delta t \bar{\mathbf{u}}^n \\
    x_p^{n+1} &= \frac{1}{2} \left( x_p^n + x_p^{(n+1)e} + \delta t \bar{\mathbf{u}}^{(n+1)e} \right)
\end{align*}
\]

(193) (194)

where the bar over the velocity vector indicates that the velocity of the fluid is interpolated at the element’s position.
8 Conclusion

The equations and numerical algorithm described in this document form the core of an evolving fire model. As research into specific fire-related phenomena continues, the relevant parts of the model can be improved. Because the model was originally designed to analyze industrial-scale fires, it can be used reliably when the fire size is specified and the building is relatively large in relation to the fire. In these cases, the model predicts flow velocities and temperatures to an accuracy of 10 to 20% compared to experimental measurements. Currently, research is focussed on improving both the gas phase and solid phase descriptions of combustion in the model so that simulations involving fire growth and suppression, especially in residential sized rooms, can be improved.

Any user of the numerical model must be aware of the assumptions and approximations being employed. There are two issues for any potential user to consider before embarking on calculations. First, for both real and simulated fires, the growth of the fire is very sensitive to the thermal properties (conductivity, specific heat, density, burning rate, etc.) of the surrounding materials. Second, even if all the material properties are known, the physical phenomena of interest may not be simulated due to limitations in the model algorithms or numerical grid. Except for those few materials that have been studied to date at NIST, the user must supply the thermal properties of the materials, and then validate the performance of the model with experiments to ensure that the model has the necessary physics included. Only then can the model be expected to predict the outcome of fire scenarios that are similar to those that have actually been tested.
9 Nomenclature

\( A_s \) water droplet surface area
\( B \) pre-exponential factor for Arrhenius reaction
\( C \) Sprinkler C-Factor
\( C_D \) drag coefficient
\( C_s \) Smagorinsky constant (LES)
\( c_p \) constant pressure specific heat
\( D \) diffusion coefficient
\( D^* \) characteristic fire diameter
\( d_m \) median volumetric droplet diameter
\( E \) activation energy
\( f \) external force vector (excluding gravity)
\( g \) acceleration of gravity
\( \mathcal{H} \) total pressure divided by the density
\( h \) enthalpy; heat transfer coefficient
\( h_i \) enthalpy of \( i \)th species
\( I \) radiation intensity
\( I_b \) radiation blackbody intensity
\( k \) thermal conductivity; suppression decay factor
\( M \) molecular weight of the gas mixture
\( M_i \) molecular weight of \( i \)th gas species
\( \dot{m}_f \) fuel mass flux
\( \dot{m}_w \) water mass flux
\( m_w^* \) water mass per unit area
\( \dot{m}_O \) oxygen consumption rate per unit area
\( \text{Nu} \) Nusselt number
\( \text{Pr} \) Prandtl number
\( p \) pressure
\( p_0 \) background pressure
\( \tilde{p} \) pressure perturbation
\( \mathbf{q}_r \) radiative heat flux vector
\( \dot{q}'' \) heat release rate per unit volume
\( \dot{q}_r'' \) radiative flux to a solid surface
\( \dot{q}_c'' \) convective flux to a solid surface
\( \dot{Q} \) total heat release rate
\( \dot{Q}^* \) characteristic fire size
\( \mathcal{R} \) universal gas constant
\( \text{Re} \) Reynolds number
\( r_d \) water droplet radius
\( \text{RTI} \) Response Time Index of sprinkler
\( s \) unit vector in direction of radiation intensity
\( \text{Sc} \) Schmidt number
\( \text{Sh} \) Sherwood number
\( T \) temperature
\( t \) time
\( t_b \) thermal element burn-out time (LES)
\( U \) integrated radiant intensity
\textbf{u} = (u, v, w) \quad \text{velocity vector}

\dot{W}_i'' \quad \text{production rate of } i\text{th species per unit volume}

We \quad \text{Weber number}

\mathbf{x} = (x, y, z) \quad \text{position vector}

X_i \quad \text{volume fraction of } i\text{th species}

Y_i \quad \text{mass fraction of } i\text{th species}

Y_{O^\infty} \quad \text{mass fraction of oxygen in the ambient}

Y_F \quad \text{mass fraction of fuel in the fuel stream}

y_s \quad \text{soot yield}

Z \quad \text{mixture fraction}

\gamma \quad \text{ratio of specific heats}

\Delta H \quad \text{heat of combustion}

\Delta H_O \quad \text{energy released per unit mass oxygen consumed}

\delta \quad \text{wall thickness}

\kappa \quad \text{absorption coefficient}

\mu \quad \text{dynamic viscosity}

\nu_i \quad \text{stoichiometric coefficient, species } i

\Phi \quad \text{dissipation function}

\rho \quad \text{density}

\tau \quad \text{viscous stress tensor}

\chi_r \quad \text{radiative loss fraction}

\sigma \quad \text{Stefan-Boltzmann constant}

\omega = (\omega_x, \omega_y, \omega_z) \quad \text{vorticity vector}
References


