

## 1

## Modeling of Gasifiers: Overview of Current Developments

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*..the broad industrial application of advanced reactor types such as fluidised bed and entrained phase reactors originally discovered and developed from coal gasification.*

*K.H. van Heek [93]*

## 1.1

### Numerical Modeling in Engineering

Recent developments in technology demonstrate that real progress in the field of mechanical/aerospace or chemical engineering can be achieved using numerical<sup>1)</sup> modeling. Basically, experiments are much more expensive than computations. Especially, taking into account possible risks or disasters during tests, simulations become more attractive. However, it should be emphasized that any model is *useless* unless it reproduces the values measured or their basic behavior during real processes. The basic advantages of modeling are as follows [1]:

- Numerical modeling is *cheaper* than experimental investigations.
- Numerical modeling makes it possible to “*see*” or “*access*” processes that are impossible to measure, for example, processes inside particles undergoing heterogeneous combustion or gasification.
- It can be used to *find the optimum* parameters in existing industrial equipment or provide novel designs of next-generation devices.
- Modeling is *not static*. It can be always improved at any time to expand the range of applications.
- Finally, computer codes can be seen as a *reservoir of knowledge*.

The basic disadvantage of modeling is its complexity in terms of understanding all steps and algorithms, which are often closed, for example, commercial software, or programmed without appropriate accompanying commentaries.

1) In most cases, the mathematical equations describing a real engineering problem cannot be solved analytically, and therefore numerical solution is required usually.

Generally, the process of modeling engineering problems can be divided into three basic phases. Each of these phases follow the other, but they can be considered as independent, too. The *first phase* is the development or adaptation of a *mathematical model* for the corresponding engineering problem. Depending on the level of modeling, this model can have the form of partial differential equations (PDE) or just algebraic nonlinear equations. Very often, the mathematical model cannot be solved analytically, and that is why a numerical solution is needed. Hence, the *second phase* includes the construction of an appropriate *numerical model*<sup>2)</sup> based on the mathematical model developed. The *third phase* is defined by the *actual solution* of the numerical model.

At the beginning of computational engineering era 30–40 years ago, a researcher or engineer had to go through all three phases. Now, with significant development of computational hardware (powerful multiprocessor-based desktop computers) and computing software, for example, MATLAB<sup>®</sup> or CFD-based software ANSYS-Fluent<sup>®</sup>, -CFX<sup>®</sup>, STAR-CCM+<sup>®</sup>, or open-source CFD software OpenFOAM<sup>3)</sup>, the first and second phases of modeling have been clubbed into one phase. This phase combines the adaptation of existing models (to choose from a “menu” list) or model development and its implementation into an available software. Hence, the basic challenge is now *not* to program numerical models, but to develop a physical model that is capable of predicting adequately the processes under investigation. In this view, the difference between model development and numerical simulations is that the former requires more professional knowledge and understanding of the physics of processes, whereas the latter, using a commercial software, can be performed by any user after reading a manual.

Applied to chemical and processes engineering, there is tremendous potential for using computational engineering software to design reactors with higher efficiency than existing ones. In this view, computational fluid dynamics (CFD) provides the solution methods and algorithms including their programming for conservation equations applied to any fluid/gas/solid flow problem. In particular, for scaling up of chemical reactors, a reliable fluid dynamic reactor model is of great benefit. Especially, gas–solid reactors, which are very important elements in many energy and chemical conversion processes, can be designed optimally, for example, by increasing the conversion rate of coal in a gasifier, using multiphase CFD. Analysis of the literature (see, e.g., [3]) shows that one of the first multiphase CFD software was developed for modeling fluidized-bed coal gasifiers using the continuum approach [4] and the Euler–Lagrange approach [5]. Finally, the well-known open-source MFIX code [6–9] developed originally by the US Department of Energy (DOE) is widely used for coal gasification modeling [8, 10]. Recently, because of the implementation of coal combustion/gasification multiphase models into different commercial CFD software (e.g., ANSYS-Fluent<sup>®</sup>), many computational works have been published in the

2) A *numerical model* is one in which a final solution can be calculated using a *finite number of basic arithmetic operations* [2].

3) [www.openfoam.org](http://www.openfoam.org)

literature (e.g., see [11–17]) that are devoted to the calculation of chemically reacting multiphase flows in pilot-scale gasifiers [18–24].

### 1.1.1

#### The Role of Direct Numerical Simulation (DNS) in Particulate-Flow Modeling

Generally, numerical simulations of *particulate flows* are based on either the *Euler–Lagrange* or the *Euler–Euler* model. The difference between the two models lies in their different treatment of the movement of the particles (for details see Chapter 3). For a detailed review of existing works devoted to the multiscale modeling of two-phase flows, in particular fluidized beds, we refer the reader to [25–27].

Referring to the modeling of particulate flows (dense<sup>4</sup>) and dilute<sup>5</sup>), the Euler–Lagrange models show greater “physical” resolution of particulate flows in comparison to the Euler–Euler models. In the Euler–Lagrange model, the solid phase is represented by solid particles that obey Newton’s laws of motion, written in the Lagrange space. The gas or liquid phase is treated using an Eulerian type of model represented by the Navier–Stokes equations written in the volume-averaged or direct form depending on the “physical” scale resolution. In the literature devoted to fluidized-bed reactors, this class of models is called *discrete particle models* (DPMs) or *discrete element models* (DEM)s (e.g., see the reviews [25] and [27]).

Following the classification of DPM models presented in [25], DPM-based Euler–Lagrange models can be divided into unresolved (UDPM) or resolved discrete particle models (RDPM) depending on the coupling of the Euler and the Lagrange phases. In the UDPM method, the Eulerian grid is *at least one order of magnitude larger than the size of the particles*. Thus, to model fluid–particle interaction or the particle temperature, one requires closure correlations to describe the impulse exchange between the particle and the fluid or the heat transfer between the particle and the surrounding fluid (see [28] and [29]). Here, it should be noted that, similar to Euler–Euler models, one of the disadvantages of the UDPM is the semiempirical character of the correlation for the drag force and the Nusselt number that are used to calculate the particle trajectories and the particle temperature, respectively. In particular, in [28] it is shown that the drag relation has a significant impact on the accuracy of numerical simulations relating to experimental data. Furthermore, *the UDPM method may generate numerical problems once the volume of the grid cell approaches the volume of the particle*. Thus, in the case of turbulent flow, this method is suitable only in combination with the so-called Reynolds-averaged Navier–Stokes equations (RANS) or with large eddy simulations (LESs) (e.g., see [30] and [31]). The particle–particle interaction within UDPMs is basically handled by two different

4) Dense particulate flows are characterized by high values of the volume fraction of a solid  $\varepsilon > 0.01$  [3].

5) Dilute particulate flows are defined by  $\varepsilon < 0.01$ .

models: a hard-sphere and a soft-sphere model (see Chapter 3). However, recently, the soft-sphere model originally proposed by Cundall and Strack [32] has become more popular for UDPM simulations using a large number of particles ( $N_p > 10^6$ ) [33].

In spite of the significant success of coupled DPM/DEM CFD models in the prediction of fluidized-bed systems, one of the limitations of this class of models is the use of the so-called subgrid *zero equation* (0-D) models for the modeling of hydrodynamic forces acting on the particles, and heat and mass transfer between the particles and the fluid. Applied to heat transfer calculation using UDPM CFD models, the temperature evolution of the particles is basically calculated using a simplified semiempirical model where the effective heat transfer coefficient is calculated using a Nusselt number relation (e.g., Ranz–Marshall equation for a spherical particle [34]). This simplification is justified by the fact that the cell size of an Eulerian grid is larger than the size of the particles. Thus, to model fluid–particle interaction or the particle temperature, one requires closure correlations to describe the momentum exchange between the particle and the fluid or the heat transfer between the particle and the surrounding fluid (see [29]).

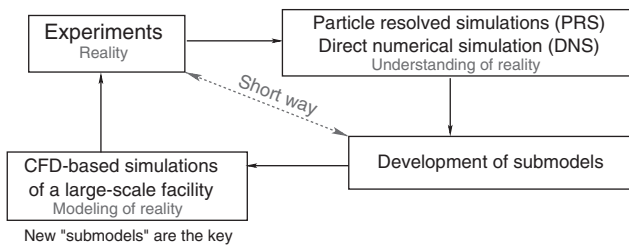
In contrast to the UDPM, the *resolved discrete particle model* (RDPM) uses an Eulerian grid, with cells about one order of magnitude smaller than the size of the particles (see [26] for details). Both the particle–particle and particle–fluid interactions are modeled directly using hard-sphere/soft-sphere models and surface integrals, respectively. From this point of view, in the literature RDPM is often known as the *direct numerical simulation* (DNS) model or *particle-resolved simulation* (PRS). It should be noted that originally the term “DNS” came from turbulence modeling (e.g., see [35]), where it was implied that the size of the smallest turbulent vortices (Kolmogorov scale) is larger than the smallest cell in a computational grid. Applied to simulations of moving particles, the main idea of DNS models is to embed an irregular solid particle/particles into a larger, simple domain and to specify no-slip boundary conditions on the particle boundaries. Thus the fluid flow is computed only between the solid particles. The forces acting on each particle are calculated directly by taking the surface integrals over each particle. Generally, the so-called immersed boundary (IB) method is used for the DNS of particulate flows. For a review of IB methods, we refer the reader to the work by Mittal and Iccarino [36]. Examples of DNS-based models for particulate flows can be found in representative works by Pan *et al.* [37] (isothermal particulate flows) and by Deen *et al.* [38] (nonisothermal particulate flows), where corresponding reviews of the fundamental work in this area are given in detail.

An alternative to the classical DNS Euler–Lagrange models of particulate flows is the combination of the lattice Boltzmann method [39, 40], which is used to solve the fluid flow between the solid particles, and an Euler method, which is applied to solve a convection–diffusion equation for a passive scalar such as the temperature or species concentration (e.g., see [41]).

Applied to the modeling of gas–solid chemically reacting flows in gasifiers or combustors, the UDPM-based Euler–Lagrange models have become well-established tools for macroscale simulations of transport processes, whereas DNS-Euler–Lagrange approaches are used for understanding the micro- and mesoscale processes by resolving single or several chemically reacting particles including intraparticle diffusion of chemical species and heat transfer. In this view, DNS of fluid–particle flows allows the prediction of parameters and the “observation” of processes, which are almost impossible or very expensive to measure in experimental studies. Hence, DNS plays the role of a *numerical experiment*. For example, in the case of nonisothermal gas–solid flows, DNS can deliver the heat transfer coefficient between the fluid and the particles, which can be utilized in the development of closure correlations (submodels) to describe the heat transfer exchange between the particle and the fluid (e.g., see [38]). Hence, new submodels play the role of *scale “bridges”* between microscale (e.g., interfacial phenomena) and macroscale simulations (e.g., reactor-scale simulations). Finally, utilization of the so-called submodels allows one to take into account the multiscale character of gas–solid flows. However, in the development of submodels, the following requirements should be kept in mind:

- *Simple submodels are of great importance* because anybody can understand them and they are basically fast and robust in simulations. However, too simple a model may provide only superficial information.
- At the same time, too sophisticated a submodel may take years to develop and it can cause difficulties in computations (e.g., the convergence problem). Here, it should be noted that, generally, submodels have to be run many times until the macroscale simulation converges.

An example of a such multiscale modeling strategy for particulate flows in chemical reactors is shown in Figure 1.1. The different scales to be modeled in a gasifier are shown in Figure 1.2. The sequential use of all steps shown in the figure may significantly reduce the errors or uncertainties in model development and, hence, enhance the reliability of the final results and models. As an example, the work by Agrawal *et al.* [42] provides a thorough review of a similar multiscale approach.



**Figure 1.1** Principal scheme of a multiscale modeling strategy for particulate flows in a chemical reactor.

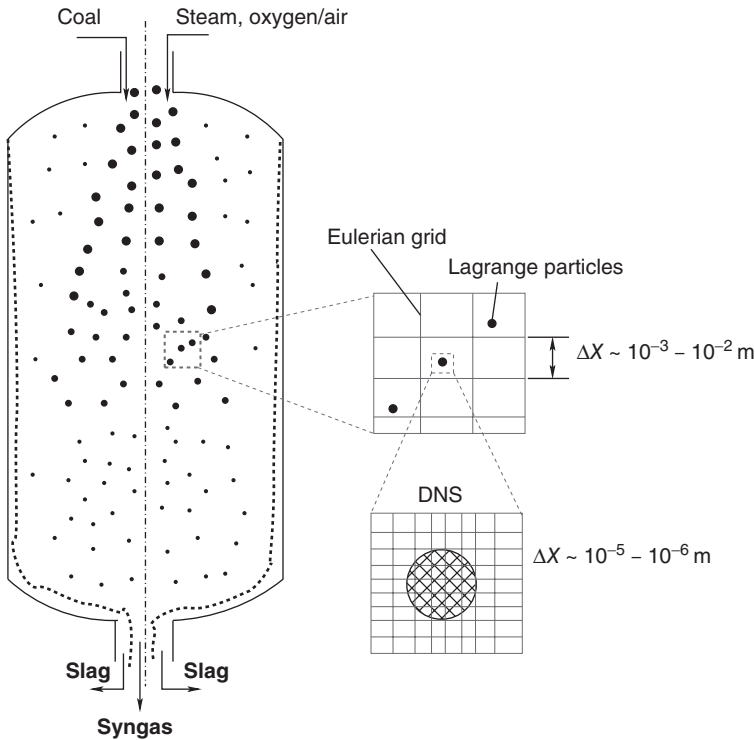
## Summary

DNS using new numerical and postprocessing algorithms has the potential to transform significantly the current model development. In particular, it is advantageous that, compared to advanced/expensive experimental techniques, DNS can enable researchers to have access to the microscale and mesoscale characteristics of chemically reacting gas–solid flows. This strategy allows engineers to make new model designs in a timely and cost-efficient manner. Moreover, carrying out “numerical experiments” for different input parameters can provide a better understanding of the problem to be solved. *And by knowing a “physically accurate” numerical solution, it is possible to find a semiempirical approach to solving the same problem using less computational time and resources.* **In this book, this relatively new approach is illustrated by numerous examples.**

## 1.2

### CFD-based Modeling of Entrained-Flow Gasifiers

Applied to industrial companies engaged in the development and production of industrial-scale gasifiers, CFD – and especially commercial CFD software – has only recently been explored as a powerful tool in designing and optimizing gasifiers and their working parameters. It is evident now that the coupling of CFD with a chemical reaction engineering theory has the potential to reduce the need for expensive and time-consuming large-scale tests. Especially, in the last 20 years significant improvements in the CFD modeling and computational hardware and combustion/gasification model development have made it possible to gain insights into the influence of design variables, coal properties, and processing conditions on the gasifier performance. In the first line, it concerns entrained-flow gasifiers because of their several advantages over fluidized-bed or fixed-bed systems. In particular, entrained-flow gasifiers are becoming popular in the coal conversion into synthetic fuels because they produce higher coal gasification rates and are easier in operation than other reactors due to their simple design and reliability (see Chapters 2 and 11 for details). Moreover, entrained-flow gasifiers are easier to model compared to fluidized-bed and fixed-bed gasifiers because of the dilute particulate flows, where particle collisions can be neglected. The principal scheme of an entrained-flow gasifier including different scales of modeling concepts is shown in Figure 1.2. In an entrained-flow reactor, small ( $O(10^{-4})\text{m}$ ) coal particles (solid or as a slurry) are injected into a moving gaseous medium which enhances the dispersion of particles over the reactor. This effect provides the largest solid–gas reactive surface area, which promotes the chemical reaction between the solid and gas phases. As gaseous medium, oxygen (air) and steam are introduced simultaneously to the coal particles. Near the inlet of fuel in the zone of coal–oxygen mixing, extremely high temperature is to be expected as a result of the relatively high oxygen concentration and the combustion of volatiles produced during the devolatilization of coal. Strictly



**Figure 1.2** Principal scheme of an entrained-flow gasifier and different scales of modeling concepts.

speaking, this phenomenon is very similar to the processes occurring near the inlet of a coal combustor. The heat produced by oxidation of coal supports the endothermic gasification reactions. Theoretically, in an ideal case, gasification processes can be organized in a such way that the heat release from oxidation (exothermic) reactions balances the heat needed for the endothermic gasification reactions. However, in real practice, all chemical reactions may take place simultaneously in a gasifier because of the impact of gas flow and turbulence. In this view, a CFD-based modeling coupled with heterogeneous and homogeneous chemistry is necessary to understand and then to optimize the dynamics of coal conversion under entrained-flow conditions. In general, CFD simulation serves as a preliminary part for complex design studies or to investigate phenomena in a known gasifier setup. A wide range of boundary as well as model conditions are needed to define the proper CFD simulation framework, which requires the understanding of the processes to be modeled. In particular, many physical effects have to be taken into account such as turbulent flow, coal particle conversion reactions, homogeneous chemistry, particle–flow interactions, radiation, and so on. For simplicity, each effect can be subdivided into complex subprocesses in order to be able to develop a final overall model for a numerical investigation of a

reactor. Such module-based principle is often used in computational engineering. Applied to modeling of a gasifiers, the following multiscale phenomena have to be taken into account [43, 44]:

- HHI – heterogeneous and homogeneous chemistry interaction.
- TCI – turbulence–chemistry interaction.
- PTU – particle–turbulence interaction.
- PGI – particle–gas interaction including the following processes:
  - heating and moisture evaporation of coal particles,
  - coal devolatilization and char formation,
  - char oxidation and gasification.
- PWI – particle–wall interaction.
- PPI – particle–particle interaction<sup>6</sup>.

It should be noted that the direct modeling of a gasifier resolving the particles and all turbulence scales (e.g., see [18]), ranging from several meters for the whole reactor to several micrometers for the coal particles (Figure 1.2) is impossible nowadays because of the lack of computing power. For example, to carry out CFD-based particle-resolved simulation of an entrained-flow gasifier with a height of 10 m and a radius of 1 m, we need a grid with more than  $10^{17}$  control volumes assuming an average grid spacing of  $\Delta x = 10^{-5}$  m.

Therefore, recent CFD studies of an entrained-flow gasifier include models for turbulence, radiation heat transfer, coal drying and devolatilization, and char combustion/gasification. Analysis of recent publications [14, 15, 18–22, 45, 46] including a recent review paper [44] shows that, to describe multiscale phenomena in chemically reacting entrained pulverized coal flows, the following mainstream models are used. For a detailed review of the models used for stochastic tracking of particles in an entrained-flow gasifier including its coupling with different turbulence models (SST  $k - \omega$ , standard, and realizable  $k - \epsilon$ , LES), the reader is referred to the work of Kumar and Ghoniem [47].

### 1.2.1

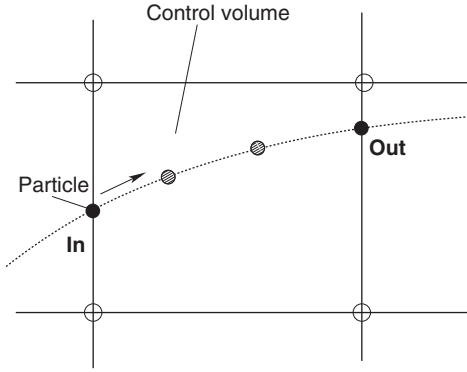
#### Mainstream Computational Submodels

Pulverized coal combustion/gasification is basically modeled as a dilute solid–gas reacting flow utilizing an Eulerian–Lagrangian approach, (e.g., see [43, 44]). The so-called particle-source-in-cell method [48] is used to calculate the interaction between a moving particle and gas, governed by mass, momentum, energy, and species conservation through various particle source terms. To illustrate the main idea of this method, we write the mass conservation equation for the gas phase as follows [48]:

$$\frac{\partial}{\partial t}(\rho) + \frac{\partial}{\partial x_i}(\rho u_i) = -\frac{1}{V_{cv}} \sum_{i=1}^{n_p} \frac{(m_{p,i,out} - m_{p,i,in})}{\Delta t} \quad (1.1)$$

6) Because of the low values of volume fraction of the solid, PPI can be neglected in the modeling of entrained-flow gasifiers.





**Figure 1.3** Illustration of mass, heat, and momentum exchange between continuous (gas) and discrete (particle) phases.

where  $m_{p,i,\text{out}}$  is the mass of the  $i$ th particle at the cell exit [kg],  $m_{p,i,\text{in}}$  is mass of the  $i$ th particle at the cell entry [kg], and  $n_p$  is the number of particles inside the control volume  $V_{\text{cv}}$  (see Figure 1.3).

The momentum transfer from the continuous phase to the discrete phase is computed by summing the change in momentum of each particle passing through a control volume:

$$\mathbf{F}_{g-s} = \sum_{i=1}^{n_p} \left[ \frac{18 \mu C_D Re}{24 \rho_p d_p^2} (\mathbf{u}_{p,i} - \mathbf{u}) + \mathbf{F}_{\text{other}} \right]. \quad (1.2)$$

The change in thermal energy of each particle passing through a control volume is given by

$$\begin{aligned} \dot{Q}_{\text{cv}} = & \sum_{i=1}^{n_p} \frac{1}{\Delta t} [(m_{p,i,\text{in}} - m_{p,i,\text{out}}) (-\Delta h_{fg} + \Delta h_{\text{devot}} + \Delta h_{\text{het}})] - \\ & - \sum_{i=1}^{n_p} \frac{1}{\Delta t} \left[ m_{p,i,\text{out}} \int_{T_{\text{ref}}}^{T_{p,\text{out}}} c_p dT - m_{p,i,\text{in}} \int_{T_{\text{ref}}}^{T_{p,\text{in}}} c_p dT \right] \end{aligned} \quad (1.3)$$

where  $C_D$  is the drag force coefficient,  $\mu$  is the viscosity of the gas,  $\rho_p$  is the density of the particle,  $d_p$  is the diameter of the particle,  $Re$  is the *relative* Reynolds number,  $\mathbf{u}_p$  is the velocity of the particle,  $\mathbf{u}$  is the velocity of the gas phase,  $T_{p,\text{out}}$  is the temperature of the  $i$ th particle at the cell exit,  $T_{p,\text{in}}$  is the temperature of the  $i$ th particle at the cell entry, and  $\Delta h_{fg}$ ,  $\Delta h_{\text{devot}}$ ,  $\Delta h_{\text{het}}$  are the enthalpies of moisture evaporation, devolatilization, and heterogeneous reactions, respectively.

### 1.2.1.1 Particle Conversion

Generally, the rates of particle conversion processes such as drying, devolatilization, and gasification are heterogeneous reactions and are slow compared to the turbulence timescale [15]. In this case, the conversion fluxes are calculated using the mean gas properties.

- For the prediction of a particle drying, basically, one uses the so-called surface-based model, which assumes that the moisture content is located on the particle surface [1, 44]. Thus, the drying can be described by utilizing a theory used for droplet evaporation. Energy conservation equation for the particle during heating and drying has the form

$$m_p c_p \frac{dT_p}{dt} = \underbrace{A_p \alpha (T_\infty - T_p)}_{\text{conv.-diffus.}} + \underbrace{A_p \epsilon_s \sigma (T_\infty^4 - T_p^4)}_{\text{radiation}} - \dot{m} \cdot \Delta h_{fg} \quad (1.4)$$

where  $A_p$  is the particle surface area.

If  $T_p < T_{\text{boil}}$ , the moisture flux can be calculated using the semiempirical relation for interfacial species mass balance, given by

$$\dot{m}'' = Y_{\text{H}_2\text{O}}^* \dot{m}'' + \underbrace{\rho_g \beta (Y_{\text{H}_2\text{O}}^* - Y_{\text{H}_2\text{O},\infty})}_{\text{conv.-diffus.}} \quad (1.5)$$

where  $Y_{\text{H}_2\text{O}}^*$  is the interfacial mass fraction of steam,  $Y_{\text{H}_2\text{O},\infty}$  is the steam mass fraction in a control volume of the CFD grid,  $Y_{\text{H}_2\text{O}}^* = \frac{P_{\text{sat}}}{P} \frac{MW_{\text{H}_2\text{O}}}{MW_{\text{mix}}}$ , and  $P_{\text{sat}}$  is the vapor pressure.

If  $T_p > T_{\text{boil}}$ , the evaporation rate is governed by heat transfer, and therefore the quasi-steady-state model for droplet evaporation can be used [49]:

$$\dot{m}'' = \frac{\dot{m}}{A_p} = \frac{\alpha}{c_{p,g}} \ln(1 + B_q) \quad (1.6)$$

where

$$B_q = \frac{c_{p,g} (T_\infty + T_{\text{boil}})}{\Delta h_{fg} - \frac{\sigma \epsilon_s (T_\infty^4 + T_{\text{boil}}^4)}{\dot{m}''}} \quad (1.7)$$

with  $T_\infty$  being the temperature of gas in a control volume of the CFD grid.

The heat transfer coefficient  $\alpha$  is defined as follows:

$$\alpha = \frac{Nu \lambda}{d_p}, \quad Nu = 2 + 0.6 Re_p^{1/3} Pr^{1/3}. \quad (1.8)$$

Here, the Nusselt number is calculated using the Ranz–Marshall relation [34]. For a detailed review of the basic models, see Chapter 5.

- At high ambient temperatures, after the drying is completed<sup>7)</sup> the particle temperature increases. As a result of the thermal decomposition of organic compounds inside a coal particle, the so-called volatile matter “leaves” the particle. This process is called *devolatilization*. Devolatilization kinetics and yields are **strongly** dependent on the heating rate, the ambient gas, and the ambient pressure (e.g., see the recent two-dimensional CFD-based simulation of a coal particle ignition [51]).

Some of the most usable devolatilization models are as follows [1]:

<sup>7)</sup> For large particles,  $d_p > 10^{-3}$  m, drying and devolatilization can occur simultaneously. But in the case surface-based drying model, this is not valid anymore [50].

- The single-global reaction rate model [52]. The thermal decomposition rate of dry coal particles is described as

$$-\frac{dm_p}{dt} = -\dot{m}_p = k [m_p - m_{p,0} \cdot (1 - f_{v,0}) (1 - f_{w,0})] \quad (1.9)$$

where  $m_p$  and  $m_{p,0}$  are the current and initial particle mass,  $f_{v,0}$  is the mass fraction of volatiles on a dry basis, and  $f_{w,0}$  is the mass fraction of moisture *initially* present in the coal particle as received. The rate constant  $k$  has the form

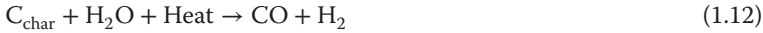
$$k = A_k \cdot \exp \left[ -\frac{E_d}{R_u T_p} \right] \quad (1.10)$$

where  $R_u$  is the ideal gas constant, and  $T_p$  is the particle temperature.

- The multiple-reaction model [53].
- The Kobayashi model (for details, see [1]).
- The CPD (chemical percolation devolatilization model [54]. This model characterizes the devolatilization behavior of *rapidly heated coal based on the physical and chemical transformations* of the coal structure.

Basically the volatile matter contained in the coal are assumed to be composed of CO, CO<sub>2</sub>, H<sub>2</sub>, CH<sub>4</sub>, H<sub>2</sub>O, and C<sub>x</sub>H<sub>y</sub> as a heavy fraction. For a detailed review of the basic models, see Chapter 10.

- Generally, to predict the char consumption by gasification/combustion, the so-called Baum and Street model [55] is used. Smith [56] generalized this approach for simplified multispecies surface reactions represented by three simple global heterogeneous reactions:



This model belongs to the class of one-film models and considers that heterogeneous reactions take place on the surface of the particle (in most cases a sphere). In the literature, this approach is referred to as the *diffusion kinetic single film* (DKSF) or the *kinetic/diffusion* model. The species O<sub>2</sub>, CO<sub>2</sub>, and H<sub>2</sub>O are considered to react heterogeneously with char after diffusion to the particle surface through the boundary layer. The kinetic/diffusion-limited rate model uses harmonic average weighting between diffusion and kinetic defined rates:

$$k_i^S = \frac{k_{\text{diff},i} \cdot k_{\text{kin},i}}{k_{\text{diff},i} + k_{\text{kin},i}} \quad (1.14)$$

where  $k_{\text{diff},i}$  and  $k_{\text{kin},i}$  are the diffusion and kinetic rate constants for the  $i$ th reaction, respectively:

$$k_{\text{diff},i} = C_i \frac{[(T_p + T_\infty) / 2]^{0.75}}{d_p}, \quad (1.15)$$

$$k_{\text{kin},i} = A_E T^n \exp \left( \frac{-E_A^i}{(R_u T_p)} \right) \quad (1.16)$$

where  $C_i$  is the overall mass diffusion-limited constant [44]:

$$C_i = \frac{\nu_i MW_C \overline{MW}}{MW_i R T_0^{1.75}} \cdot Sh \cdot D_{i,0} \cdot \frac{P_0}{P} \quad (1.17)$$

where  $\nu_i$  is the stoichiometric coefficient in the  $i$ th reaction, and  $\overline{MW}$  is the average molecular weight of the gas mixture in a control volume of the CFD grid.

The constant  $C_i$  depends on a heterogeneous reaction [44]. In the CFD software ANSYS-Fluent<sup>®</sup> [57], this constant has a default value of about  $10^{-12} \text{ s K}^{-0.75}$ .

Taking into account basic heterogeneous surface reactions, the carbon consumption rate has the form

$$\dot{m}_C = A_p \sum_{i=1}^3 P_{i,g} k_i^S. \quad (1.18)$$

For a detailed review of the basic models and description of the new intrinsic-based model, see Chapter 10.

### 1.2.1.2 Turbulence–Chemistry Interaction

In industrial-scale gasifiers/combustors, the gas flow inside a reactor is always turbulent, which makes the numerical modeling a nontrivial task. The so-called DNS cannot be used for the whole gasifier (see discussion in previous section). In this case, turbulence models (e.g., RANS) have to be applied to account for the effect of turbulence on the transport processes including chemistry–turbulence interaction. Applied to CFD-based modeling of gasifiers, the impact of turbulence on the homogenous chemistry has been modeled using the so-called eddy dissipation model (EDM) and the eddy dissipation concept (EDC) models coupled with RANS or LES.

The EDM model [58] assumes that the chemical reaction is faster than the timescale of the turbulence mixing of the species, which is governed by the large eddy-mixing time,  $k/\epsilon$ , as originally proposed by Spalding [59]. Thus, a homogeneous chemical reaction is supposed to occur instantaneously when the reactants are brought into contact. This assumption makes it unnecessary to use finite-rate kinetics. An enhanced version of the EDM takes the finite-rate chemistry into account. Finally, the smaller reaction rate given by the Arrhenius rate and turbulent mixing rate is chosen for homogeneous reactions (e.g., see [21]). The important tuning parameters in this model are the so-called Magnussen's empirical constants A and B (default: A = 4.0, B = 0.5), for the reactant and the product, respectively. Their variation can significantly change the final results (e.g., see [44]). A significant limitation of this model is that only two reactions can be considered whereas, in fact, there are different Arrhenius rates for a multistep mechanism [57].

As the next extension of the EDM model for the case of *multistep chemical kinetics*, the so-called EDC model was used in many works on combustion/gasification [24, 60, 61]. The EDC model is based on the original work of Magnussen [62].

It should be noted that in the EDC model a scalar equation is solved for each chemical species. Thus, in comparison to the EDM model, the EDC model needs a relatively high calculation time for integrating the chemistry.

An alternative and very promising model for coal combustion/gasification is the flamelet model [63], where the fuel and the oxidizer are supplied separately to the reaction zone. The distinguishing feature of this model as applied to CFD-based combustion modeling is that there is no need to calculate scalar equations for the species. Instead of chemical species transport equations, one needs to solve only two equations for the mean mixture fraction and the mixture variance. In particular, recently Prieler *et al.* [61] carried out a CFD analysis of an 11.5 kW lab-scale furnace with oxygen–natural gas combustion for a high-temperature process using three different TCI models: two-step EDM, EDC with 17 species and 46 reversible reactions, and the steady laminar flamelet model (SFM). It was shown that EDM was unable to predict the oxygen–fuel combustion correctly. In contrast to the EDM results, temperatures calculated using EDC and SFM showed close agreement with the measured data in the furnace. However, using SFM the computational time was decreased from 3 weeks needed for EDC model to 4 days on an 8 CPU-core computer.

Finally, it should be noted that the so-called advanced TCI models, such as flamelet or PDF models, for coal combustion/gasification are in the development phase. However, some promising results obtained using flamelet-based models have been published recently [51, 64].

## 1.2.2

### Review of CFD-related Works

Next, we present a brief review of the recent literature devoted to CFD-based modeling of entrained-flow gasifiers and related processes. Here, the focus is on entrained-flow coal gasification since 1990. A comprehensive review of the basic works devoted to the modeling and simulations of entrained-flow gasifiers published before 1990 can be found in [65, 66].

This short review is divided into the analysis of CFD-based works that used noncommercial software and the commercial ANSYS-Fluent® software.

#### 1.2.2.1 Noncommercial Software

At the beginning of 1990, Sijerčić and Hanjalć developed a two-dimensional code for the modeling of an entrained-flow gasifier [13, 67, 68]. The gas phase was described in the Eulerian frame and the discrete phase in the Lagrangian frame, taking into account heat and mass transfer exchange between the phases using the particle-source-in-cell method [48]. Four heterogeneous chemical reactions of coal were considered in a kinetic–diffusion regime accounting for the impact of particle velocity on the heat and mass transfer between a particle and a gas using the Ranz–Marshall relation. However, concentrations of chemical species on the particle surface were neglected in the surface-based burnout model. The distinguishing feature of the model of Sijerčić and Hanjalć is the use of a transport

equation for the particle number density  $N_p$  to calculate the particle concentration field necessary for the prediction of radiative heat transfer coefficients. The code was validated against published experimental data for the BCURA reactor [69]. More information concerning the BCURA reactor is given in the next section.

One of the first published results on the three-dimensional simulation of an industrial-scale 200 tpd (tons per day) two-stage air-blown entrained-flow coal gasifier was by Chen and coworkers [14–16]. An extended coal–gas mixture fraction model with the “multi solids progress variables method” was utilized to simulate the gasification reaction and the reactant mixing process. The model tracked 11 500 particle trajectories, and a  $21 \times 21 \times 62$  grid mesh was used. It was shown that the three different zones, namely the devolatilization, the combustion, and the gasification zones, have complex contours in the gasifier. Moreover, it was demonstrated that turbulent fluctuations in the volatile and the char–oxygen reaction have a significant impact on the temperature and gas composition.

In 2013, Abani and Ghoniem [22] published one of the first LES calculations of a lab-scale entrained-flow gasifier (BYU gasifier) operating at atmospheric pressure. They used the open-source CFD software *OpenFOAM* with a standard kinetic/diffusion approach. The total computational mesh size consisted of  $0.33 \times 10^6$  cells. The Rosin–Rammler distribution [70] (with a distribution index of 3.5) was used to represent the variation in particle size with a minimum diameter of 10  $\mu\text{m}$  and a maximum diameter of 80  $\mu\text{m}$ . The LES/RANS results showed that in the combustion zone RANS calculation overpredicted the mixing rate, which led to higher combustion temperatures and it did not capture accurately the unsteady characteristics of the two-phase mixing in the gasification zone, which, on the other hand, was important for modeling char consumption. LES calculation resulted in a longer combustion zone and a more uniform species distribution in the gasification zone. The overall results of the LES simulations showed a more accurate prediction of the scalar fields compared to similar RANS calculations.

#### 1.2.2.2 Commercial Software

With significant development and progress in the commercial CFD software ANSYS-Fluent<sup>®</sup>, several papers on entrained-flow gasifiers have been published recently. Silaen and Wang [19] effectively employed the DPM-CFD gasification model available in ANSYS-Fluent<sup>®</sup> to investigate the influence of different submodels on gasification performance including five turbulence models, four devolatilization models, and three solid coal sizes. Three-dimensional simulations were carried out using the following RANS turbulence models: Standard  $k - \epsilon$ , RNG  $k - \epsilon$ , Standard  $k - \omega$  Model, SST  $k - \omega$  Model, and Reynolds Stress Model (RSM). The results showed that the standard  $k - \epsilon$  and the RSM turbulence models gave consistent results. Concerning devolatilization rates, chemical percolation devolatilization (CPD) and the single-rate models reproduced more moderate results and the devolatilization rates were not as slow as those of the Kobayashi model.

Recently, Lu and Wang [45] carried out investigations on three-dimensional simulations of a two-stage slagging-type entrained-flow gasifier (operating

pressure 24 atm, 1700 tpd, 190 MW energy output) using five different radiation models available in the ANSYS-Fluent<sup>®</sup> software [57]: discrete transfer radiation model (DTRM), P-1 radiation model, Rosseland radiation model, surface-to-surface (S2S) radiation model, and discrete ordinates (DO) radiation model. The commercial software ANSYS-Fluent<sup>®</sup>-Version 12. was utilized. The computational grid consisted of  $1.1 \times 10^6$  unstructured tetrahedral cells. For TCI modeling, both EDM and finite-rate models were used to calculate the reaction rates. It was shown that the P-1 model was more robust and stable in predicting the syngas temperature and composition compared to the other four models used. However, the P-1 model resulted in the lowest temperature of the inner wall of the gasifier. The DO and DTRM models took about twice the CPU time as the other radiation models.

The assumptions that surface-based heterogeneous kinetics does not adequately represent the gasification process advocate that investigations in the area of heterogeneous reactions submodels should be the focus for gasification reactor modeling [p.94] [65]. In this context, Australian researchers of the Cooperative Research Centre for Coal in Sustainable Development (CCSD) incorporated new submodels in a two-dimensional RANS simulation in the CFD software ANSYS-Fluent<sup>®</sup> for better predictions of the drying, pyrolysis, and heterogeneous coal gas–char reactions. The results of numerical investigations were compared with experimental data [71–74]. In particular, in a series of conference papers, Hla *et al.* coupled successfully the intrinsic heterogeneous reaction rates at elevated pressure with a model proposed by Laurendeau [75]. The intrinsic character of coal was accounted for by the random pore model developed by Bhatia and Perlmutter [76]. The simulation results indicated that the used models could predict a more realistic image of the gasifier performance. Not only the trends could be reproduced but also good agreements of the experimental data for different types of coal were reached for CO, CO<sub>2</sub>, and H<sub>2</sub> species concentrations along the axis. Good agreement with experimental data was reached especially with anthracite coals. However, it was found out that the boundary conditions (e.g., wall temperature) had a great impact on the final results concerning their agreement with experiments [73].

Kumar and Ghoniem [77] modified the DKSF submodel by Baum and Street using an additional term characterizing a moving flame front (MFF) introduced by Zhang *et al.* [78]. The overall results showed that the use of the MFF model gave more accurate results reflecting better physics of particle burn-up history. The main idea of MFF model is to vary the flame front radius up to several (up to 50) particle radii to fit the burnout curve to the experimental data. However, it is well known that the ratio between the flame radius and the radius of a carbon particle oxidizing in an O<sub>2</sub>-based atmosphere cannot exceed the value of 2. This value can be derived using the classical two-film model (see Chapters 6 and 8).

It can be seen that significant progress was achieved in the commercial CFD software ANSYS-Fluent<sup>®</sup> concerning the prediction of heat and fluid flow in pulverized coal jets (e.g., see the comparison of different CFD software [79]) and in entrained-flow gasifiers. On the other hand, the development of improved

submodels describing particles conversion was slower and thus the submodels developed in the 1980s are still basically to describe the particle–gas interaction in gasifiers.

Only recently, new, advanced submodels developed in this century received more attention in CFD-related predictions of chemically reacting flows in gasifiers and pulverized coal combustors. For instance, Vascellari *et al.* published a series of papers devoted to CFD-based simulation of pulverized coal MILD combustion [80] and the BYU entrained-flow gasifier [81] using advanced coal/char-conversion submodels [23, 24]. In particular, numerical simulations carried out in [80] revealed that the use of new virtual homogeneous-zone single-film submodel (H-zone model), originally developed by Schulze *et al.* [82], produces results that are closer to the experimental data in comparison to the standard Baum and Street burnout submodel. The distinguishing feature of the H-zone model [82] is the coupling of homogeneous CO oxidation reaction with heterogeneous gasification reactions for the calculation of particle temperature and carbon conversion rate. Additionally, this new “surface-based” subgrid model considers a detailed description of the transport phenomena in the proximity of the particles under convective environmental conditions. Further developments of this char-conversion submodel is presented in Chapter 10.

Recently, Vascellari *et al.* [24] implemented a single  $N$ th-order reaction (SNOR) model originally developed by Liu and Niksa [83] into the ANSYS-Fluent<sup>®</sup> software using the user-defined function (UDF). This model is an intrinsic-based model which takes into account random pore evolution and char density changes. The CBK/E [84] and CBK/G [83] models for char oxidation and gasification, respectively, were used for calibrating the SNOR kinetic model. Turbulence was modeled using the realizable  $k - \varepsilon$  approach coupled with the EDC model accounting for the TCI in combination with a detailed kinetic mechanism (for details, see [24]). Radiation was modeled via the P1 model available in ANSYS-Fluent<sup>®</sup>. Comparison with the experimental data for the BYU entrained-flow gasifier [81] showed good agreement for gas composition and carbon conversion. However, the main disadvantage of the SNOR submodel is the need for calibration with the CBK/G model. Moreover, for the calculation of mass conversion rates for each heterogeneous reaction, the model uses an empirical factor which accounts for the physical evolution effects such as char density changes and pore evolution. This factor is a function of the char-conversion rate  $X$ , which has the form of a fifth-order polynomial correlation for oxidation and gasification reactions separately. The char density was calculated as a function of  $X$  according to [83]

$$\rho_c = \rho_{c,0} (1 - X)^{\alpha_n}, \quad X = 1 - \frac{m_c}{m_{c,0}} \quad (1.19)$$

where  $\alpha_n$  is an empirical model parameter.

It can be seen that this model does not account for the simultaneous change of particle density and particle diameter. It is a well-known fact that, during the oxidation of char, the so-called diffusion-controlled regimes govern the char conversion, where the particle diameter changes instead of the density [75, 85, 86].



From this point of view, further developments of intrinsic-based submodels are needed to avoid the use of many unphysical input empirical model parameters and to account for intraparticle diffusion and heat transfer coefficients into such models.

### Summary

The analysis of the literature shows that in most recent simulations the commercial CFD code ANSYS-Fluent<sup>®</sup> was utilized using more advanced submodels implemented via UDF. As discussed at the beginning of this chapter, those software packages greatly reduce the effort in developing better strategies for modeling the complex physics in gasifiers. Despite the rapid growth in the availability and speed of computer technologies, there is only a slow transition from 2D to 3D RANS or even LES calculations. The recent research focus has been on improving the turbulent nature of two-phase flows and incorporating and validating new submodels to account for the intrinsic nature of gasification. Abani *et al.* [22] demonstrated that a good estimation of the unsteady characteristics of the turbulent flow field can yield a better description of the combustion and gasification processes. The works by Hla *et al.*, Kumar *et al.*, and Vascellari *et al.* [24] use heterogeneous reaction models for CFD in their gasifier simulations and showed the intrinsic behavior of coal. There is a great need for new developments in this area because most of the presented CFD predictions are based on nonintrinsic combustion assumptions that do not capture accurately the gasification behavior of coal. In the future, slag behavior and CFD gas–particle interaction (dense particle flows) need to be the new focus points for further research. In addition to new CFD tools, there is a great need for validation cases from char to high-ash and high-volatile yield coals under varying operation conditions such as high pressures to be able to accomplish future developments.

Finally, it should be emphasized that any successful application of a CFD software requires good understanding of the models and assumptions that will be used in the simulations. However, in many cases commercial CFD codes are black boxes, where it is impossible to “read” the model and equations in the code. From this point of view, it is extremely important to validate the software before actual studies can be carried out. At the same time, the parametric runs can help understand the basic assumptions in the model used.

## 1.3

### Benchmark Tests for CFD Modeling

A review of recent works devoted to CFD-based modeling of entrained-flow gasifiers revealed the importance of models and software validation against experimental data published in the literature. In the following section, we analyze experimental data for lab-scale gasifiers published in the open literature. It should be stated that the proximate and ultimate analyses are based on either as

received (ar) or as dry and ash-free (daf) state. Those input data are important to characterize, among others, devolatilization processes that have a great impact on the overall carbon consumption.

### 1.3.1

#### British Coal Utilization Research Association Reactor (BCURA)

The BCURA pilot-scale combustion reactor is an air-blown furnace which was operated at ambient pressure. The system could process more than 9 tpd of coal. A detailed description of the experimental setup and results are documented in the article by Gibson and Morgan [69] and Baker *et al.* [87]. A first approach of the mathematical model of the reactor was proposed by Field in his book (see Appendix U [88]). Several reports have been published by BCURA that are not part of this review and may provide further details and experimental results. For further information, see the references in [69].

The BCURA pilot-scale combustor consists of a horizontal cylindrical chamber with two inlets for coal, and primary and secondary air. The reactor has a height of 6.1 m and a diameter of 1.1 m. All basic geometric parameters are given in Figure 1.4.

The presented setup and results are taken from the experiment “Flame 49” [69]. In Tables 1.1 and 1.2, the boundary conditions are given. The inner wall temperature of the chamber is a complex function of the estimated heat loss and depends on the used models (e.g., radiation model) and boundary conditions. The outer wall temperature can be assumed with 400 K. The injection speed of the particles may be estimated with  $21.9 \text{ m s}^{-1}$  if you consider that the fluid and particle flow field are equal at the entry point.

The proximate and ultimate analysis are given in Table 1.3. The used coal has a high-ash and a fixed-carbon content and can be classified as a low-rank bituminous coal. No data is available for heterogeneous kinetics and only limited experimental data for the BCURA rig are documented in literature. Experimental results of “Flame 49” are illustrated in Figures 1.5 and 1.6. The measured overall heat loss adds up to 1350 kW. A simplified contour plot of temperature isolines is shown in Figure 1.6. A flame zone appears at approximately 1 m [69].

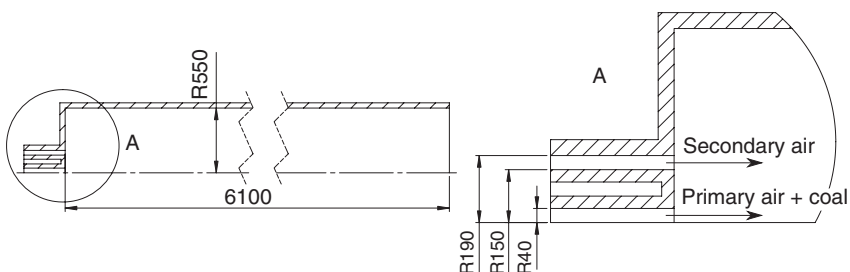


Figure 1.4 Geometry of the BCURA rig (in mm).

**Table 1.1** Boundary conditions [69].

BC	Value
$\dot{m}_1$	0.104 kg/s
$T_1$	373 K
$\dot{m}_2$	0.822 kg/s
$T_2$	626 K
$T_{\text{wall}}$	$T_{\text{wall}}(\dot{Q}_{\text{loss}})^b$
$p_{\text{operation}}$	1 bar
$\epsilon_{\text{wall}}$	0.7
$\dot{Q}_{\text{loss}}$	1350 kW

<sup>a</sup> ...  $X_{\text{N}_2}/X_{\text{O}_2} = 0.79/0.21$

<sup>b</sup> ...  $T_{\text{wall\_min}} = 400$  K

**Table 1.2** Simulation setup parameters [69].

BC	Value
$\dot{m}_{\text{Fuel}}$	0.086 kg/s
X-Velocity	21.9 m/s
R-Velocity	0 m/s
$T_{\text{fuel}}$	373 K
Min. diameter	2e-6 m
Max. diameter	200e-6 m
Mean diameter	43e-6 m
Spread par.	1.0 <sup>b</sup>

<sup>a</sup> ...  $\Delta u \approx u_{\text{air}} - u_{\text{Coal}} = 0$

<sup>b</sup> ... assumed

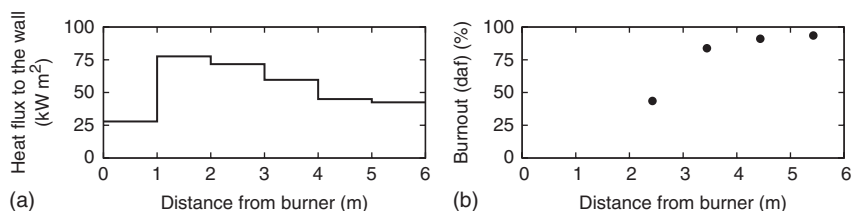
**Table 1.3** Coal properties [69].

Moist	Proximate analysis (wt%)			Ultimate analysis (daf, wt%)					HHV (ar) (Mj/kg)
	FC	VM	Ash	C	H	O	S	N	
4.10	53.59	32.01	10.30	80.60	5.14	11.59	1.86	0.81	27.9

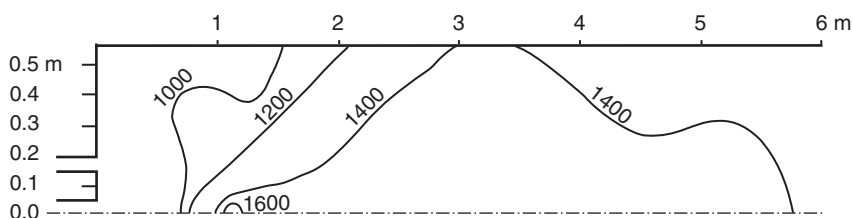
### 1.3.2

#### Brigham Young University Reactor (BYU)

The BYU reactor is an oxygen-fed lab-scale gasifier (0.6 tpd) operating at ambient pressure (Tables 1.4–1.6). This experimental rig is very well documented. Therefore, it is well suited as a validation setup. But you need to be aware that only highly volatile coals were considered in the past surveys. Therefore, a good



**Figure 1.5** Measured data of “Flame 49” with (a) heat flux through the walls and (b) burnout of the coal along the axis. Graphs based on the data taken from [69].



**Figure 1.6** Contour plots of the measured temperature field of Flame 49 (in K). Graphs based on the data in [69].

**Table 1.4** Boundary conditions for the BYU reactor [81].

BC	Value
$\dot{m}_1$	7.290 g/s
$T_1$	367 K
$X_{O_2}/X_{Ar}/X_{H_2O}$	0.850/0.126/0.024
$\dot{m}_2$	1.840 g/s
$T_2$	450 K
$X_{H_2O}$	1
$T_{wall}$	unkown
$p_{operation}$	1 bar

devolatilization prediction for this reactor model is necessary. Four types of coals were investigated, and experimental results in the axial and radial directions for different species concentration have been reported by Brown *et al.* [81]. Additional information on conducted BYU experiments can be found, among others, in the journal papers of Soelberg, Smoot, and Smith *et al.* [89, 90].

The experimental rig consists of six horizontally oriented sections with a length of 305 mm and one section of 153 mm which is partially illustrated in Figure 1.7. The effective length of the reactor chamber is specified as 1890 mm. A tube-in-tube configuration separates the primary from the secondary inlet streams with a diameter of 4.6 and 28.6 mm, respectively; for details see [81].

**Table 1.5** Injection conditions for the BYU reactor [81].

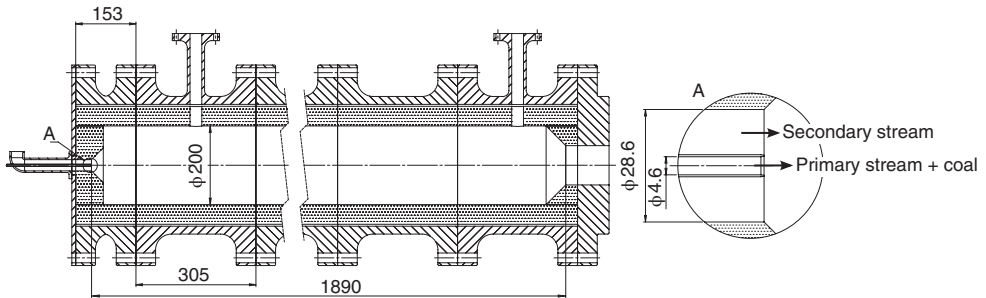
BC	Value
$\dot{m}_{\text{fuel}}$	6.634 g/s
X-Velocity	50.6 m/s
R-Velocity	0 m/s
$T_{\text{fuel}}$	367 K
Min. diameter	3e-6 m
Max. diameter	35e-6 m
Mean diameter	80e-6 m
Spread par.	1.0 <sup>b</sup>

<sup>a</sup> ...  $\Delta u \approx u_{\text{gas}} - u_{\text{fuel}} = 0$

<sup>b</sup> ... assumed

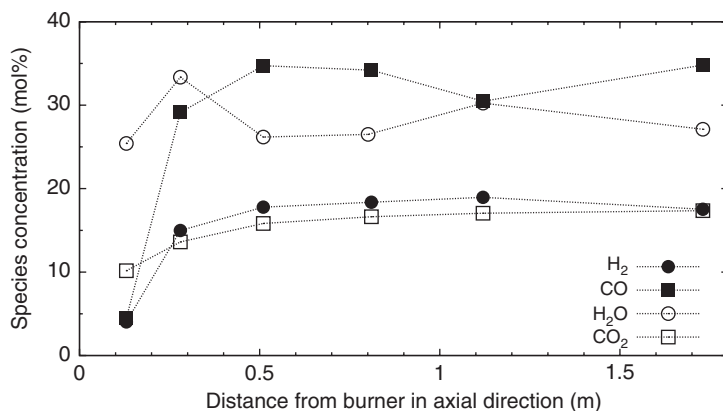
**Table 1.6** Coal properties Utah bituminous coal [81].

Proximate analysis (wt%)				Ultimate analysis (daf, wt%)					HHV (db)
Moist	FC	VM	Ash	C	H	O	S	N	Mj/kg
2.4	43.7	45.6	8.3	77.60	6.56	13.88	1.42	0.55	29.8

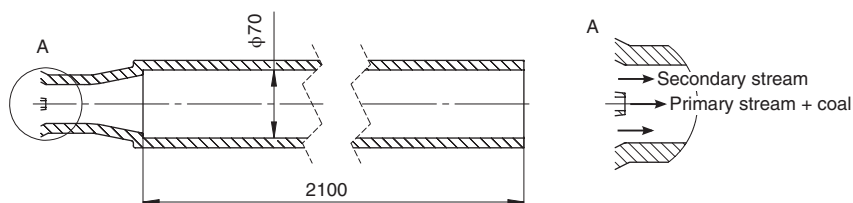
**Figure 1.7** Geometry of the BYU rig (in mm) [81].

Several experiments have been conducted by the BYU. The presented boundary conditions and obtained results focus only on experiments with Utah bituminous coal. Other coals (Wyoming subbituminous, North Dakota lignite, and Illinois No. 6 bituminous coals) and the corresponding kinetic parameters for the heterogeneous reaction are documented by Brown *et al.* [81]. Wall temperatures were not mentioned directly in this work, but it can be assumed that rig walls are nonadiabatic (e.g., see [23, 81]).

Several parametric studies for the BYU lab-scale reactor were performed [89]. Molar concentrations of CO, CO<sub>2</sub>, H<sub>2</sub>, and H<sub>2</sub>O in the radial and axial direction are documented for the Utah bituminous coal, and an example is illustrated in



**Figure 1.8** Molar concentrations of CO, CO<sub>2</sub>, H<sub>2</sub>, and H<sub>2</sub>O along the axis for Utah bituminous coal. Graph based on the data in [81].



**Figure 1.9** Geometry of the PEFR rig (in mm).

Figure 1.8 for the axial distribution of different species. The exit temperatures are also measured to have an additional parameter to fit the used boundary conditions. For the Utah coal, the exit gas temperature is estimated between 1350 and 1400 K. Further information on the experimental results of Utah coal is presented by Soelberg *et al.* [90]. He has included contour plots for the obtained species concentrations in the BYU reactor.

### 1.3.3

#### Pressurized Entrained-Flow Reactor (PEFR)

The PEFR is a small lab-scale reactor (0.1 tpd) and is part of a project developed by the Cooperative Research Centre for Coal in Sustainable Development (CCSD) in cooperation with the Commonwealth Scientific and Industrial Research Organization (CSIRO) during the late 1990s. The special feature of this reactor is the directly measured high-pressure heterogeneous intrinsic kinetics of the used Australian coals at 20 bar. Hla *et al.* have described all parameters for setting up a CFD calculation [72, 73]. More information concerning the gasification behavior of Australian coals can be found in the article by Harris *et al.* [71] and supplemented in the research reports of the CCSD [74, 91, 92].

**Table 1.7** Simulations boundary conditions for the PEFR reactor [74].

BC	Value
$\dot{m}_1$	1.680 g/s
$T_1$	298.15 K
$X_{N_2}$	1.0
$\dot{m}_2$	15.385 g/s
$T_2$	1275 K
$X_{N_2}/X_{O_2}$	0.973/0.027
$T_{\text{wall\_inlet}}$	Adiabat
$T_{\text{wall\_reactor}}$	1673 K
$P_{\text{operation}}$	20 bar

**Table 1.8** Setup parameters for the PEFR reactor [74].

BC	Value
$\dot{m}_{\text{fuel}}$	0.511 g/s
X-Velocity <sup>a</sup>	1.8 m/s
R-Velocity	0 m/s
Temperature	298.15 K
Min. diameter	20e-6 m
Max. diameter	250e-6 m
Mean diameter	177e-6 m
Spread par. <sup>b</sup>	1.12

<sup>a</sup>...  $\Delta \mathbf{u} \approx \mathbf{u}_1 - \mathbf{u}_{\text{fuel}} = 0$

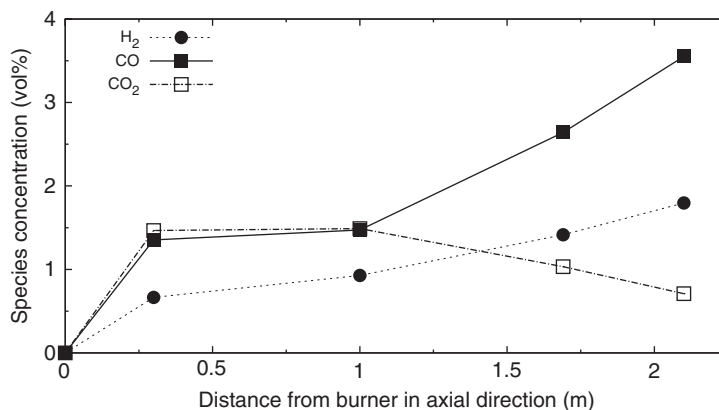
<sup>b</sup>... assumed

**Table 1.9** Coal properties of CRC252 coal.

Proximate analysis (wt%)				Ultimate analysis (daf, wt%)					HHV (db) (MJ/kg)
Moist	FC	VM	Ash	C	H	O	S	N	
10.7	39.11	38.85	10.34	78.1	5.9	14.4	0.5	1.1	25.7 <sup>a</sup>

<sup>a</sup>... value taken from Harris *et al.* [71].

The gasification reactor consists of a horizontal cylindrical chamber with two inlets for coal and primary and secondary gas streams. The vertically oriented reaction chamber is 2100 mm long and has a diameter of 70 mm. No information on the nozzle geometry is available in the literature. The basic geometric parameters are shown in Figure 1.9. The presented boundary conditions are taken from experiments for the coal type CRC252 and are listed in the articles of Hla and



**Figure 1.10** Distribution of the different species along the axis of CRC252 in the PEFR. Graph based on the data in [72].

Harris *et al.* (for details, see Tables 1.7–1.9) [72, 73]. The particle size distributions are documented in the research report of Harris *et al.* (see Table 1.8) [92].

Carbon conversion, particle diameter, and species molar concentration along the reactor axis have been described by Hla *et al.* [72] for six types of coal. The described Australian coals are characterized by an identification number, for example, CRC 281. The given proximate and ultimate analyses indicate a broad selection of different types of coal ranging from high-ash, high-volatile, and anthracite coals. In Figure 1.10, the axial distributions of the different species are illustrated for the gasification experiments of the coal CRC 252. The lines in Figure 1.10 should illustrate the general trends of the measured data. Further information is given in [72–74].

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