

# Modeling Gas Effects in a Bubbling Fluidized Bed Reactor for Biomass Pyrolysis

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## Abstract

Fast pyrolysis of biomass in a fluidized bed reactor is typically conducted in a nitrogen gas environment. Recycling product gas can improve the economics of operating such a system by reducing reliance on pure process streams.

# 1 Introduction

Fast pyrolysis is a versatile method for thermochemical conversion of solid biomass into liquid bio-oil which can be used for bio-fuel and high-value chemical production. Bio-oil is commonly generated in bubbling fluidized bed and circulating fluidized bed reactor systems in which biomass particles rapidly devolatilize in the absence of oxygen into mixtures of light gases, condensable bio-oil vapors, and solid char [4, 5, 16]. Since biomass pyrolysis normally occurs in a non-oxidizing environment, the fluidization gas (carrier gas) is often pure nitrogen [16]. To maximize bio-oil yields, the reactor typically operates at temperatures near 500°C and must maintain particle residence times up to 10 seconds and gas residence times less than 2 seconds [5]. Deviations from these conditions can result in significant production and quality penalties, therefore optimal reactor design and control become crucial to achieving commercially viable bio-oil production.

To improve the economic possibilities of biomass fast pyrolysis systems, char can be burned for process heat while recycled pyrolysis gas can assist with fluidization [4, 14]. The major generated components of pyrolysis gas are CO, CO<sub>2</sub>, CH<sub>4</sub>, H<sub>2</sub>, and other light hydrocarbons [1, 26]. Several experiments investigated the effects of these gases on reactor conditions and pyrolysis yields [14, 17, 26] but modeling the effects of the different gases was not discussed.

There are several reactor models that investigate the hydrodynamics and conversion of biomass at fast pyrolysis conditions [21, 20, 15, 23, 24]. These models assume the carrier gas is pure nitrogen which is a typical scenario for biomass fast pyrolysis. The authors are not aware of any models in the biomass pyrolysis literature that investigate the effects of a carrier gas other than pure nitrogen.

This paper uses engineering correlations, reduced-order modeling techniques, and CFD simulations to model a bubbling fluidized bed pyrolysis reactor. The scope of this study is to evaluate different fluidization gases and their effects on the hydrodynamics and biomass conversion in a fluidized bed reactor operating at fast pyrolysis conditions.

# 2 Experimental apparatus

The NREL 2FBR reactor system thermochemically converts biomass feedstock at fast pyrolysis conditions. The system is comprised of two bubbling fluidized bed (BFB) reactors where the first reactor is for biomass fast pyrolysis and the second reactor is for vapor phase upgrading. Modeling activities discussed in this paper refer to the BFB pyrolysis reactor.

An overview of the NREL 2FBR system is shown in Figure 1, components of the pyrolysis reactor are detailed in Figure 2, while dimensions and typical operating conditions of the pyrolysis unit are given in Figure 3. Sand is used as the dominant heat transfer medium in the pyrolyzer. Biomass particles are fed to the reactor via a screw auger and nitrogen is used as the fluidization/carrier

gas. More information about the NREL 2FBR biomass pyrolysis system is available elsewhere [11, 22].

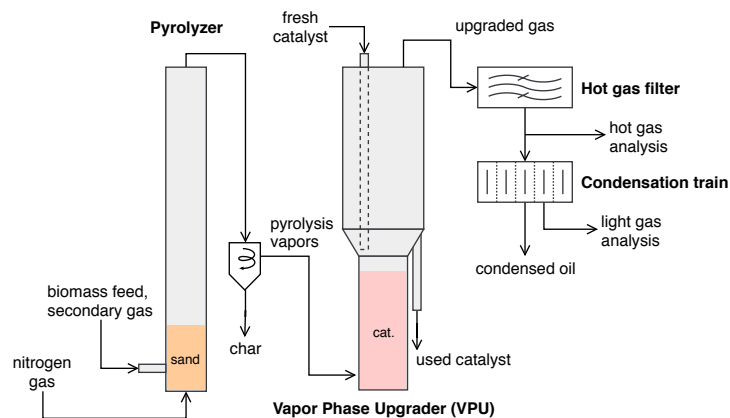


Figure 1: Overview of the NREL 2FBR system. Biomass fast pyrolysis occurs in the pyrolyzer (left) and gaseous products are catalytically upgraded in the vapor phase upgrader (right).

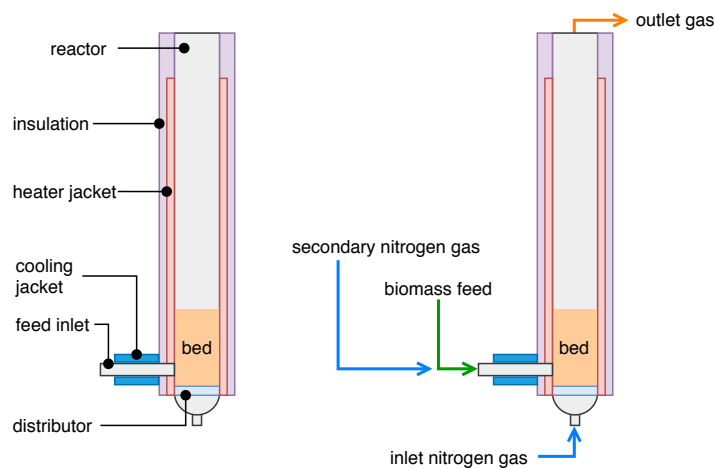


Figure 2: Components of the BFB biomass pyrolysis reactor referred to as the “pyrolyzer” in the NREL 2FBR system.



Figure 3: Dimensions and typical fast pyrolysis operating conditions for the BFB biomass pyrolysis reactor in the NREL 2FBR system.

### 3 Modeling approach

Engineering correlations, reduced-order models, and CFD modeling techniques were used to investigate the effects of recycled gas on the operation of a fluidized-bed biomass pyrolysis reactor. The following sections discuss approaches implemented in this work for calculating gas properties and the associated effects on fluidization conditions and pyrolysis yields.

#### 3.1 Gas properties

Density of the gas is calculated from the ideal gas law as shown in Equation 1 where  $\rho_{gas}$  is density ( $\text{kg/m}^3$ ),  $P$  is pressure (Pa),  $MW$  is molecular weight ( $\text{g/mol}$ ),  $R$  is the gas constant [ $(\text{m}^3 \text{ Pa}) / (\text{K mol})$ ], and  $T$  is temperature (K).

$$\rho_{gas} = \frac{P MW}{RT} \quad (1)$$

Gas viscosity ( $\mu_{gas}$  as  $\mu\text{P}$ ) is determined from Equation 2, thermal conductivity ( $k_{gas}$  as  $\text{W/mK}$ ) is estimated from Equation 3, and heat capacity ( $C_{p,gas}$  as  $\text{J/molK}$ ) is calculated from Equation 4. Temperature of the gas in Kelvin is represented by  $T$  while the regression coefficients  $A$ ,  $B$ ,  $C$ ,  $D$ ,  $E$ ,  $F$ , and  $G$  for each gas are obtained from Yaws' Handbook [25].

$$\mu_{gas} = A + BT + CT^2 + DT^3 \quad (2)$$

$$k_{gas} = A + BT + CT^2 + DT^3 \quad (3)$$

$$C_{p,gas} = A + BT + CT^2 + DT^3 + ET^4 + FT^5 + GT^6 \quad (4)$$

Several methods are available to calculate the viscosity of a gas mixture. Equation 5 calculates the mixture viscosity from the sum of the mole fraction and viscosity product of each gas component in the mixture [9] while Equation 6 accounts for the molecular weight of each gas component [10].

$$\mu_{mix} = \sum (x_i \cdot \mu_i) \quad (5)$$

$$\mu_{mix} = \frac{\sum (\mu_i \cdot x_i \cdot \sqrt{MW_i})}{\sum (x_i \cdot \sqrt{MW_i})} \quad (6)$$

The Prandtl number is a dimensionless number representing the ratio of momentum diffusivity to thermal diffusivity. It is calculated from the equation shown below where  $C_p$  is heat capacity (J/kg·K),  $\mu$  is dynamic gas viscosity (kg/m·s), and  $k$  is thermal conductivity (W/m·K).

$$Pr = \frac{C_p \mu}{k} \quad (7)$$

### 3.2 Fluidization correlations

For a bed of particles, the minimum fluidization velocity  $U_{mf}$  is the gas velocity at which the drag force of the upward moving gas equals the weight of the particles. Kunii and Levenspiel [12] provide the following equation for calculating minimum fluidization velocity

$$U_{mf} = \frac{Re_{p,mf} \mu}{d_p \rho_g} \quad (8)$$

where  $\mu$  is gas viscosity (kg/m·s),  $d_p$  is particle diameter (m),  $\rho_g$  is gas density (kg/m<sup>3</sup>), and  $Re_{p,mf}$  is the particle Reynolds number (-) at minimum fluidization conditions. The Reynolds number is calculated from the Archimedes number ( $Ar$ ) and two dimensionless constants ( $a, b$ ) which represent experimental coefficients. Different  $U_{mf}$  correlations were evaluated based on experimental data from Wen and Yu where  $(a, b) = (33.7, 0.0408)$ , from Richardson where  $(a, b) = (25.7, 0.0365)$ , and from Grace where  $(a, b) = (27.2, 0.0408)$  [12].

$$Re_{p,mf} = (a^2 + bAr)^{1/2} - a \quad (9)$$

$$Ar = \frac{d_p^3 \rho_g (\rho_s - \rho_g) g}{\mu^2} \quad (10)$$

According to Kunii and Levenspiel [12], the constants ( $a, b$ ) can be derived from the Ergun pressure drop equation based on the constants  $K_1$  and  $K_2$  where  $\epsilon_{mf}$  is the bed void fraction (-) at minimum fluidization and  $\phi$  is sphericity (-) of the bed particles. For this paper,  $U_{mf}$  is estimated based on the Ergun, Grace, Richardson, and Wen and Yu correlations.

$$a = \frac{K_2}{2K_1} \quad b = \frac{1}{K_1} \quad (11)$$

$$K_1 = \frac{1.75}{\epsilon_{mf}^3 \phi} \quad K_2 = \frac{150(1 - \epsilon_{mf})}{\epsilon_{mf}^3 \phi^2} \quad (12)$$

### 3.3 Pyrolysis kinetics

A pyrolysis kinetics scheme based on the work of Di Blasi was implemented to predict the conversion of biomass into gas, tar, and char products [2, 3]. Figure 4 gives an overview of the scheme and its reaction mechanisms. Reactions 1–3 represent the primary conversion of biomass while reactions 4–5 are secondary reactions that reduce tar yield at long residence times.

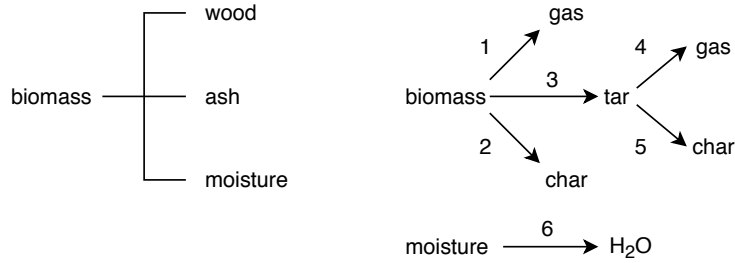


Figure 4: Diagram of the Di Blasi pyrolysis kinetics scheme for conversion of biomass to gas, tar, and char products.

The pyrolysis reactions were modeled as first-order Arrhenius type equations where the reaction rate is given as

$$r_i = C_i A_i e^{-E_i/RT} \quad (13)$$

where  $r_i$  is the rate of reaction  $i$  such that  $C_i$  is a mass based concentration,  $A_i$  is the pre-factor (1/s),  $E_i$  is the activation energy (kJ/mol),  $R$  is the gas constant, and  $T$  is the reaction temperature (K). Kinetic parameters for each reaction are listed in Table 1 where  $\Delta H$  is the heat of reaction (kJ/kg).

Table 1: Kinetic parameters for the Di Blasi biomass pyrolysis scheme.

Reaction	A (1/s)	E (kJ/mol)	$\Delta H$ (kJ/kg)	Reference
1	$4.38 \times 10^9$	152.7	-20	[3]
2	$3.27 \times 10^6$	111.7	-20	[3]
3	$1.08 \times 10^{10}$	148.0	255	[3]
4	$4.28 \times 10^6$	108.0	-42	[2]
5	$1.00 \times 10^6$	108.0	-42	[2]
6	$5.13 \times 10^6$	87.6	2700	?

### 3.4 CFD-DEM simulation

A coarse-grained CFD-DEM model was implemented for biomass pyrolysis in MFiX, an open-source, Fortran-based code [X]. The implemented coarse-grained CFD-DEM model in this research is an extension of the standard MFiX release. Gas phase transport was described using conservation equations of mass, momentum, energy, and chemical species in the Eulerian framework (Equations 14–17, respectively).

$$\frac{d(\epsilon_g \rho_g)}{dt} + \nabla(\epsilon_g \rho_g u_g) = S_\rho \quad (14)$$

$$\frac{d(\epsilon_g \rho_g u_g)}{dt} + \nabla(\epsilon_g \rho_g u_g u_g) = -\epsilon_g \nabla p + \nabla(\epsilon_g \tau) + \epsilon_g \rho_g g + S_u \quad (15)$$

$$\frac{d(\epsilon_g \rho_g E)}{dt} + \nabla(\epsilon_g \rho_g u_g E) = -\nabla Q + S_E \quad (16)$$

$$\frac{d(\epsilon_g \rho_g Y_i)}{dt} + \nabla(\epsilon_g \rho_g u_g Y_i) = -\nabla(D_i \nabla Y_i) + S_{Y_i} \quad (17)$$

where  $\epsilon_g$ ,  $\rho_g$ ,  $u_g$ ,  $p$ ,  $\tau$ ,  $Q$ , and  $Y_i$  are gas phase volume fraction, density, velocity, pressure, stress tensor, conductive heat flux, and  $i$ th chemical species, respectively,  $t$  is time,  $g$  is acceleration due to gravity,  $D_i$  is mass diffusion coefficient for species,  $S_\rho$ ,  $S_u$ ,  $S_E$ , and  $S_{Y_i}$  are mass, momentum, energy, and chemical species source terms, respectively. Fixed quantities of discrete particles with identical initial conditions were lumped into a computational coarse-grained parcel (CGP), whose motion was governed by Newton's second law of motion. All particle forces and contact dynamics were calculated on the parcel scale, whereas heat and mass transfers were calculated on particle scale and projected to the entire parcel. Accordingly, all particles in same coarse-grained parcel possess identical temperature, chemical species concentration, and momentum. The mass and diameter of each coarse-grained parcel was such that:

$$m_{CGP} = m_p W \quad (18)$$

$$d_{CGP} = d_p W^{1/3} \quad (19)$$

where  $m_{CGP}$  is CGP mass,  $m_p$  is distinct particle mass,  $W$  parcel statistical weight,  $d_{CGP}$  is CGP diameter, and  $d_p$  is distinct particle diameter. Instantaneous accelerations (translational and rotational) for each coarse-grained parcel were calculated as:

$$\frac{du_{CGP}}{dt} = g - \frac{F_p}{m_{CGP}} + \frac{F_c}{m_{CGP}} + \frac{F_d}{m_{CGP}} \quad (20)$$

$$\frac{d\omega_{CGP}}{dt} = \frac{T_{CGP}}{I_{CGP}} \quad (21)$$

where  $u_{CGP}$  and  $\omega_{CGP}$  are the CGP translational and rotational velocities,  $g$  is acceleration due to gravity,  $m_{CGP}$  is CGP mass,  $T_{CGP}$  is net torque on the

CGP, and  $I_{CGP}$  is CGP moment of inertia. The term  $F_p$  represents pressure gradient force and is calculated as product of the CGP volume and pressure gradient. The CGP collision forces  $F_C$  (parcel-parcel and parcel-wall collisions) was modeled according linear spring-dashpot model [18]. Since the number of CGP collisions is significantly lower than the number of collisions expected in system with distinct particles, the CGP coefficient of restitution was modified as a correction for energy dissipations during collisions. The proposed modification to the CGP coefficient of restitution is calculated following kinetic theory of granular flow [13] as:

$$e_{CGP} = \sqrt{1 + (e_p^2 - 1)W^{1/3}} \quad (22)$$

where  $e_{CGP}$  is CGP coefficient of restitution and  $e_p$  is distinct particle coefficient of restitution. Two different drag models were used to estimate CGP drag force  $F_d$  based on well-documented difference in the fluidization behavior of sand and biomass in the literature [19]. Drag force was estimated following Ganser-corrected Gidaspow drag model for sand (bed material) particles and a filtered drag model for biomass particles. Th Ganser correction [6] was coupled to the Gidaspow model [8] to account for non-sphericity of the sand particles as expressed below:

$$\beta_{Ganser} = \begin{cases} \beta_{Ergun} & \text{if } \epsilon_g \leq 0.8 \\ \beta_{WenYu} & \text{if } \epsilon_g > 0.8 \end{cases} \quad (23)$$

$$\beta_{Ergun} = 150 \frac{(1 - \epsilon_g)^2 \mu_g}{\epsilon_g d_{CGP}^2 \phi^2} + 1.75 \frac{(1 - \epsilon_g) \rho_g}{\epsilon_g d_{CGP} \phi} |u_g - u_{CGP}| \quad (24)$$

$$\beta_{WenYu} = \frac{3}{4} C_d \frac{(1 - \epsilon_g) \rho_g}{d_{CGP} \phi} |u_g - u_{CGP}| \epsilon_g^{-2.65} \quad (25)$$

$$C_d = \begin{cases} \frac{24}{Re K_1} (1 + 0.1118 (Re K_1 K_2)^{0.6567}) + \frac{0.4305 K_2}{1 + \frac{3305}{Re K_1 K_2}} & \text{if } Re < 1,000 \\ 0.44 & \text{if } Re \geq 1,000 \\ 0.0 & \text{if } Re = 0.0 \end{cases} \quad (26)$$

$$K_1 = \left( \frac{1}{3} + \frac{2}{3} \phi^{-0.5} \right)^{-1} - 2.25 \frac{d_{CGP}}{D} \quad (27)$$

$$K_2 = 10^{1.8148(-\log \phi)^{0.5743}} \quad (28)$$

The filtered drag model (modified Sarkar drag model) used in this research for biomass particles was proposed by Gao et al. [7] and was found by the authors to have relatively high prediction strength across multiple flow regimes in fluidized bed. The modified Sarkar drag model is derived fine-grid simulation with Wen-Yu drag model and can be computed as:



$$\beta_{Sarkar} = \beta_{WenYu}(1 - H_{Sakar}) \quad (29)$$

$$H_{Sakar} = \begin{cases} 0.95 \left(1 - e^{-\alpha_1 \alpha_2 (u_{slip}^* - u_0)^p}\right) & u_{slip}^* > u_0 \\ 0.0 & u_{slip}^* \leq u_0 \end{cases} \quad (30)$$

$$u_{slip}^* = \frac{|u_g - u_{CGP}|}{u_t} \quad (31)$$

$$\alpha_1 = \frac{(a_1 + a_2(1 - \epsilon_g) + a_3(1 - \epsilon_g)^2 + a_4(1 - \epsilon_g)^3 + a_5(1 - \epsilon_g)^4)}{1 + e^{100((1 - \epsilon_g) - 0.55)}} \quad (32)$$

$$\alpha_2 = \left(1 + \frac{a_6}{\Delta_{filter}^*} + \frac{a_7}{(\Delta_{filter}^*)^2}\right) \left(1 + \frac{a_8}{(u_{slip}^*)^2}\right) \quad (33)$$

$$u_0 = \frac{a_9 + a_{10}(1 - \epsilon_g)}{0.01 + (1 - \epsilon_g)^{a_{11}}} \left(1 + \frac{a_{12}}{\Delta_{filter}^*} + \frac{a_{13}}{(\Delta_{filter}^*)^2}\right) \quad (34)$$

$$p = (a_{14} + a_{15}(1 - \epsilon_g) + a_{16}(1 - \epsilon_g)^2) \left(1 + \frac{a_{17}}{\Delta_{filter}^*} + \frac{a_{18}}{(\Delta_{filter}^*)^2}\right) \quad (35)$$

$$\Delta_{filter}^* = \max\left(\frac{g\Delta_{filter}}{u_t^2}, \frac{1}{2}\right) \quad (36)$$

$$\Delta_{filter} = 2(\Delta_x \times \Delta_y \times \Delta_z)^{1/3} \quad (37)$$

$$u_t = \frac{gd_{CGP}^2(\rho_{CGP} - \rho_g)}{18\mu_g} \quad (38)$$

$a_1$	$a_2$	$a_3$		0.75597773	2.73931487	-5.60196497	
$a_4$	$a_5$	$a_6$		-1.65853820	16.70299223	-0.44145335	
$a_7$	$a_8$	$a_9$		0.18195034	-0.01827347	0.28441799	
$a_{10}$	$a_{11}$	$a_{12}$	=	-1.943573770	0.22177961	0.31175890	(39)
$a_{13}$	$a_{14}$	$a_{15}$		-0.15971960	0.47750002	0.062794180	
$a_{16}$	$a_{17}$	$a_{18}$		5.13011673	0.67680355	-0.54535726	

## 4 Model parameters

Parameters for the reduced-order model and CFD simulations are provided in Tables 2 and 3. Biomass particle characteristics and properties are representative of loblolly pine. Bed particle characteristics are for typical sand material. Operating conditions and reactor dimensions are based on the previously discussed NREL 2FBR fluidized bed pyrolysis unit.

Table 2: Particle size distribution for biomass feedstock.

Sauter mean diameter ( $\mu\text{m}$ )	Mass fraction (%)	Mass flow rate (kg/hr)
278	12.1	0.018
344	51.0	0.076
426	34.2	0.051
543	2.7	0.004

Table 3: Parameters for the biomass, sand (bed material), and reactor operation. Biomass  $C_p$  calculated from particle composition.

Parameter	Value	Description
biomass particle		
$e_p$	0.2	particle-particle coefficient of restitution
$e_w$	0.2	particle-wall coefficient of restitution
$e_s$	0.2	particle-sand coefficient of restitution
$\mu_p$	0.1	particle-particle coefficient of friction
$\mu_w$	0.2	particle-wall coefficient of friction
$\mu_s$	0.1	particle-sand coefficient of friction
$k_n$	100 N/m	particle spring constant
sand particle		
$d_p$	453 $\mu\text{m}$	particle diameter
$\rho_p$	2500 $\text{kg/m}^3$	particle density
$C_p$	830 $\text{kJ}/(\text{kg K})$	particle heat capacity
$\phi$	0.94	particle sphericity
$e_p$	0.61	particle-particle coefficient of restitution
$e_w$	0.61	particle-wall coefficient of restitution
$\mu_p$	0.1	particle-particle coefficient of friction
$\mu_w$	0.2	particle-wall coefficient of friction
$k_n$	100 N/m	particle spring constant
reactor operation		
$d_{\text{inner}}$	5.25 cm	inner reactor diameter
$H_{\text{reactor}}$	43.18 cm	reactor height
$H_{\text{static}}$	10.16 cm	static bed height
$p_{\text{gas}}$	101.325 kPa	gas pressure
$T_{\text{gas}}$	773.15 K	gas temperature
$Q_{\text{gas}}$	14 SLM	inlet gas flowrate

Table 4 represents the CFD simulations conducted for this paper. Each row is for a different simulation case which is performed for a particular gas

composition.

Table 4: Simulation cases for different gas mixtures where columns denote gas percentage.

Case	N <sub>2</sub>	H <sub>2</sub>	H <sub>2</sub> O	CO	CO <sub>2</sub>	CH <sub>4</sub>
1	100	0	0	0	0	0
2	0	100	0	0	0	0
3	0	0	100	0	0	0
4	0	0	0	100	0	0
5	0	0	0	0	100	0
6	0	0	0	0	0	100
7	20	20	0	20	20	20
8	50	0	0	0	50	0
9	50	0	0	50	0	0
10	0	0	50	50	0	0
11	100	0	0	0	0	0
12	80	20	0	0	0	0
13	60	40	0	0	0	0
14	50	50	0	0	0	0
15	40	60	0	0	0	0
16	30	70	0	0	0	0
17	20	80	0	0	0	0
18	15	85	0	0	0	0
19	10	90	0	0	0	0
20	5	95	0	0	0	0
21	0	100	0	0	0	0

Table 5: Chemical species composition of biomass feedstock.

Species	Mass fraction (%)	Density (kg/m <sup>3</sup> )
moisture	4.0	1,000
wood	95.9	500
ash	0.1	2,000
char	0.0	300

Table 6: Simulation parameters settings.

Parameter	Value
CFD cell size, $\Delta_x \times \Delta_y \times \Delta_z$ (mm)	$4.3 \times 4.4 \times 4.3$
time step, $\Delta_x$ (s)	varies
biomass parcel statistical weight	10
sand parcel statistical weight	20
gas phase equation of state	ideal

## 5 Results and discussion

This section provides results and related discussions for the effects of different fluidization gases on the operation and conversion of a bubbling fluidized bed reactor.

### 5.1 Comparison of gas properties

Molecular weight, viscosity, density, thermal conductivity, heat capacity, and Prandtl number of the individual gases investigated in this paper are shown in Figure 5. The gas properties were calculated at a pressure of 101,325 Pa and a temperature of 773.15 K (500°C). The lightest gas in terms of molecular weight and density is hydrogen while the heaviest gas is carbon dioxide. The highest viscosity is noted for the nitrogen gas while hydrogen has the lowest viscosity. The largest thermal conductivity is for hydrogen at approximately 0.36 W/(mK) while the other gases remain below 0.12 W/(mK). The highest heat capacity is obtained for methane at 62 J/(molK) while the lowest is for hydrogen at 29 J/(molK). The Prandtl number is similar for all the gases except for water vapor.

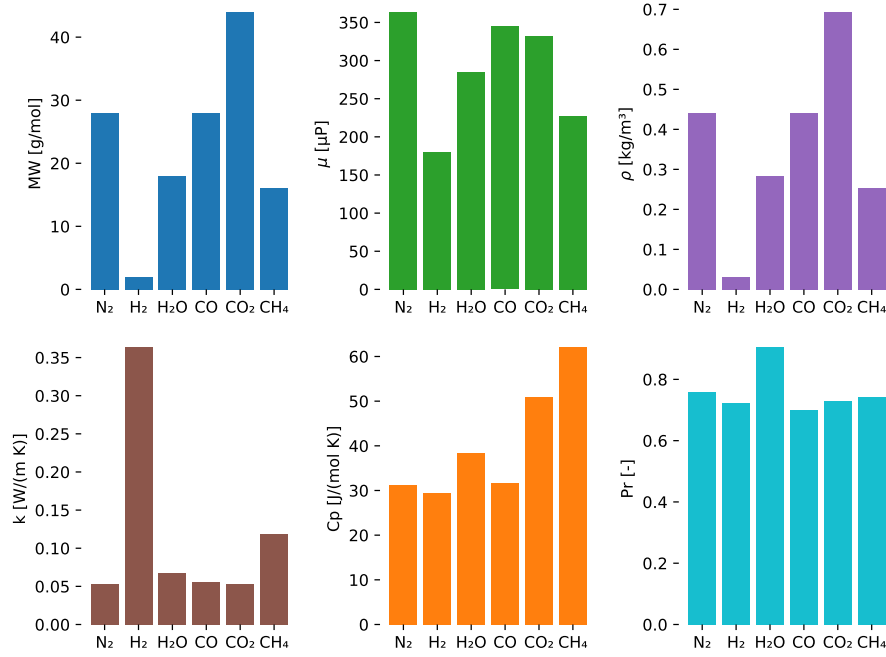


Figure 5: Comparison of molecular weight (MW), viscosity ( $\mu$ ), density ( $\rho$ ), thermal conductivity ( $k$ ), heat capacity ( $C_p$ ), and Prandtl number ( $Pr$ ) for each gas at 101,325 Pa and 773.15 K (500°C).

Properties for molecular weight, viscosity, and density for the gas mixtures investigated in this paper are shown in Figure 6. Similar to the individual gas properties, the mixture properties were calculated at 101,325 Pa and 773.15 K (500°C). The fraction of each gas in the mixture is given by the values shown at the top of each column in the figure. For example, the hydrogen and nitrogen mixture is comprised of 80% hydrogen and 20% nitrogen which is labeled as  $0.8 + 0.2$ . As expected, the carbon dioxide mixture is the heaviest in terms of molecular weight and density.

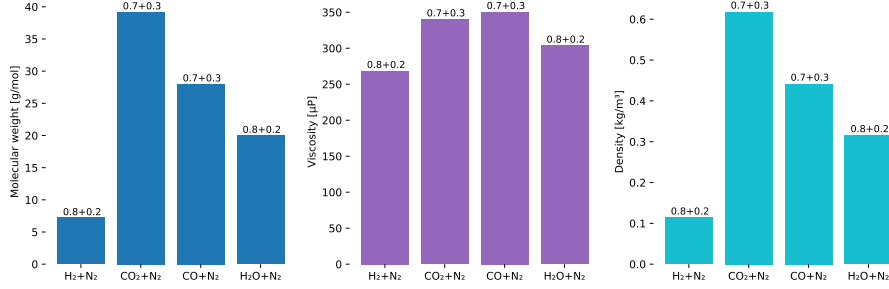


Figure 6: Comparison of gas mixture properties for molecular weight, viscosity, and density at 101,325 Pa and 773.15 K. Fraction of each gas component is shown at the top of each column.

## 5.2 Fluidization effects

Minimum fluidization velocity ( $U_{mf}$ ) of the bed material for the different fluidization gases is presented in Table 7. Hydrogen requires about twice the gas velocity to fluidize the bed of sand compared to the nitrogen, carbon monoxide, and carbon dioxide gases. This is due to hydrogen's lower viscosity and much lower density compared to the other gases. Water vapor and methane require moderately higher fluidization velocities compared to the nitrogen gas. A comparison of  $U_{mf}$  for the various fluidization gases is displayed in Figure 7.

Table 7: Minimum fluidization velocity (m/s) of the bed material calculated from various correlations for different fluidization gases. Last row represents the average  $U_{mf}$  value for each gas.

Correlation	N <sub>2</sub>	H <sub>2</sub>	H <sub>2</sub> O	CO	CO <sub>2</sub>	CH <sub>4</sub>
Ergun	0.14	0.30	0.18	0.15	0.16	0.23
Grace	0.10	0.21	0.13	0.11	0.11	0.16
Richardson	0.10	0.20	0.12	0.10	0.11	0.15
Wen and Yu	0.08	0.17	0.11	0.09	0.09	0.13
average	0.11	0.22	0.14	0.11	0.12	0.17

The superficial gas velocity ( $U_s$ ) of the nitrogen is calculated as 0.3072 m/s which is based on the 14 SLM gas flow through the distributor plate. Using this value, the ratio of  $U_s$  to  $U_{mf}$  is shown in Table 8 for different fluidization gases. The BFB pyrolysis reactor at NREL typically operates at a  $U_s/U_{mf}$  of 3 with nitrogen gas. For gases such as H<sub>2</sub>, H<sub>2</sub>O, CO, CO<sub>2</sub>, and CH<sub>4</sub>, the gas flow into the reactor must be increased to have similar fluidized bed characteristics as the nitrogen case. A comparison of the increased  $U_s$  for each gas along with the associated  $U_s/U_{mf}$  is shown in Figure 8. As expected, the hydrogen gas flow must be approximately doubled compared to the nitrogen case to achieve

similar fluidization of the bed material.

Table 8: Ratio of  $U_s$  to  $U_{mf}$  for different fluidization gases. Last row represents the average  $U_s/U_{mf}$  value for each gas.

Correlation	N <sub>2</sub>	H <sub>2</sub>	H <sub>2</sub> O	CO	CO <sub>2</sub>	CH <sub>4</sub>
Ergun	2.13	1.04	1.67	2.02	1.97	1.34
Grace	2.99	1.47	2.35	2.84	2.76	1.88
Richardson	3.16	1.55	2.48	3.00	2.91	1.98
Wen and Yu	3.69	1.82	2.90	3.50	3.39	2.32
average	2.99	1.47	2.35	2.84	2.76	1.88

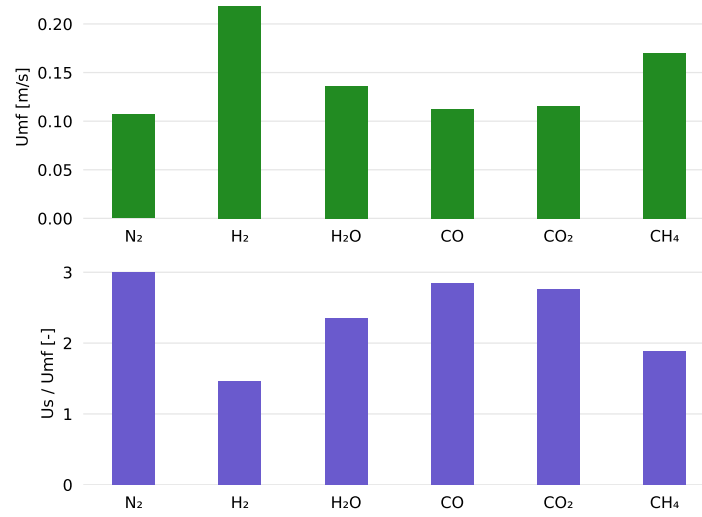


Figure 7: Comparison of the minimum fluidization velocity ( $U_{mf}$ ) and the ratio of  $U_s/U_{mf}$  for different fluidization gases. Superficial gas velocity is  $U_s$ .

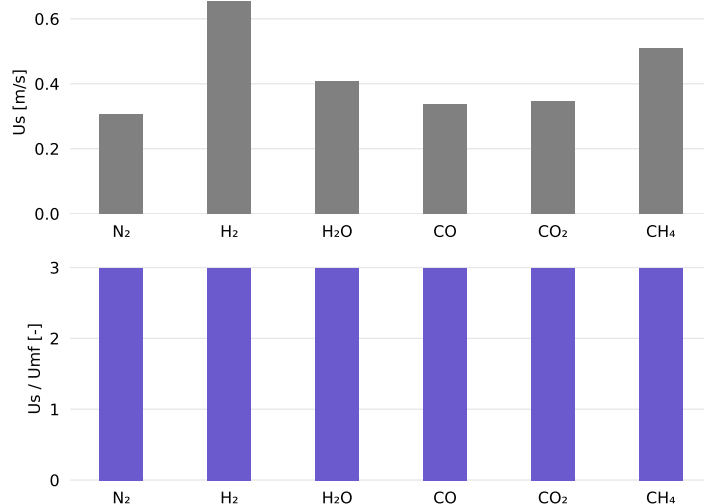


Figure 8: Comparison of the superficial gas velocity ( $U_s$ ) and the associated  $U_s/U_{mf}$  for different fluidization gases. Minimum fluidization velocity is  $U_{mf}$ .

### 5.3 Evaluation of the kinetic scheme

The Di Blasi kinetics were put to use in a batch reactor model to investigate the time scales associated with the reaction mechanisms. Figure 9 is an overview of the biomass conversion and product yields using the Di Blasi kinetics in a batch reactor at 773.15 K (500°C). At this temperature, without the effects of secondary reactions, the kinetics offer a maximum achievable tar yield of 78% within 5 seconds. However, if secondary reactions occur during the entire pyrolysis process then a maximum tar yield of only 53% is possible. The Di Blasi kinetics suggest that minimizing the extent of secondary reactions is critical to producing the maximum possible tar yield.

A range of reaction temperatures were applied to the Di Blasi kinetics in the batch reactor model as shown in Figure 10. The kinetics suggest that temperature has a negligible effect on primary tar yield but effects of secondary reactions are more pronounced. When secondary reactions occur during the entire pyrolysis process, maximum tar yields are realized at higher temperatures but with shorter residence times. These results suggest that if secondary reactions are minimized then temperature should not have a drastic effect on tar yield.



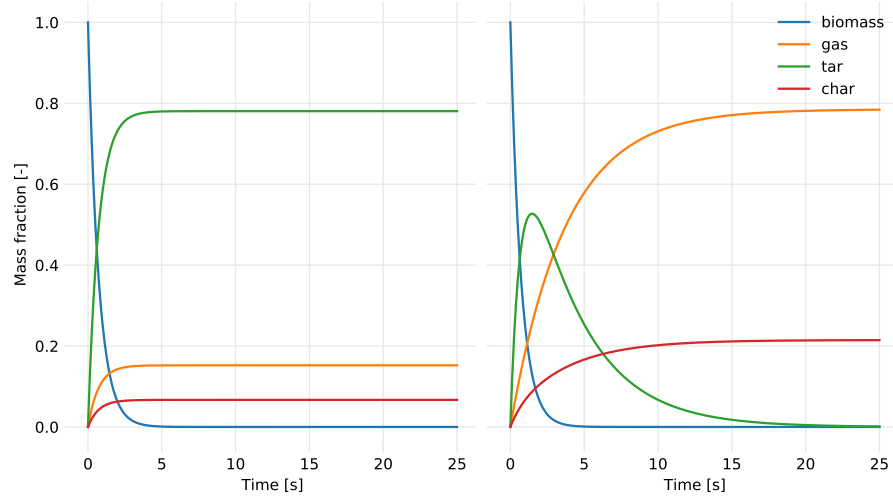


Figure 9: Biomass conversion and product yields in a batch reactor model at 773.15 K (500°C) according to the Di Blasi kinetic reactions. Results shown for primary reactions only (left) along with primary and secondary reactions (right).

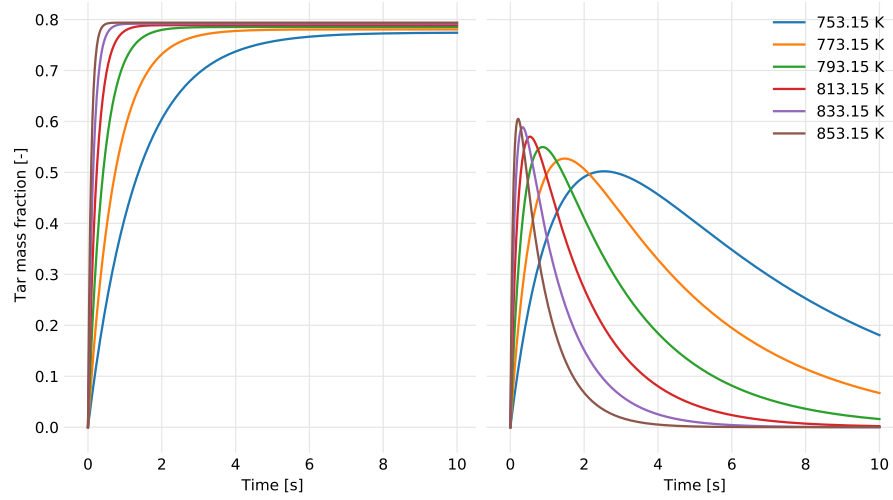


Figure 10: Tar yields for reaction temperatures of 753.15–853.15 K (480–580°C) using the Di Blasi kinetics in a batch reactor model. Results shown for primary tar (left) along with primary and secondary tar (right).

## 5.4 CFD-DEM validation

The predicted yield of pyrolysis products (bio-oil, light gas, and biochar) was validated against experimental data reported by [XXX]. In their experimental work, [XXX] carried out biomass pyrolysis in the same NREL 2FBR fast pyrolysis system that is modeled and simulated in this research. Additionally, the process variables used in the experimental work are consistent with those implemented for the  $N_2$  and  $H_2$  cases in this research. Figure 11 shows that the predicted yields of pyrolysis products closely follow the experimental data with absolute deviation ranging between 1% and 6%. The largest observed deviations occur in the prediction of bio-oil and are attributed to the non-closure of mass balance for the experimental data. The reported mass closure for the experimental data was about 94%. A mass-proportional adjustment of the experimental data to enforce 100% mass closure decreases the absolute deviation of bio-oil prediction to about 2% or less.

From a qualitative point of view, the implemented CFD-DEM simulation in this research was able to acceptably predict the increase in light gas yield and decrease in biochar yield when fluidizing gas was changed from  $N_2$  to  $H_2$ , as seen in the experimental data. Predicted bio-oil yield slightly increased when fluidizing gas was changed from  $N_2$  to  $H_2$ , contrary to experimental data showing a slight decrease. The relative change in bio-oil yield between  $N_2$  to  $H_2$  was however quite small for both experimental data (2%) and CFD-DEM prediction (-1%).

These results demonstrate that the CFD-DEM model implemented in this research is capable of realistically simulating the characteristic effects of fluidizing gas on the performance of lignocellulosic biomass pyrolysis.

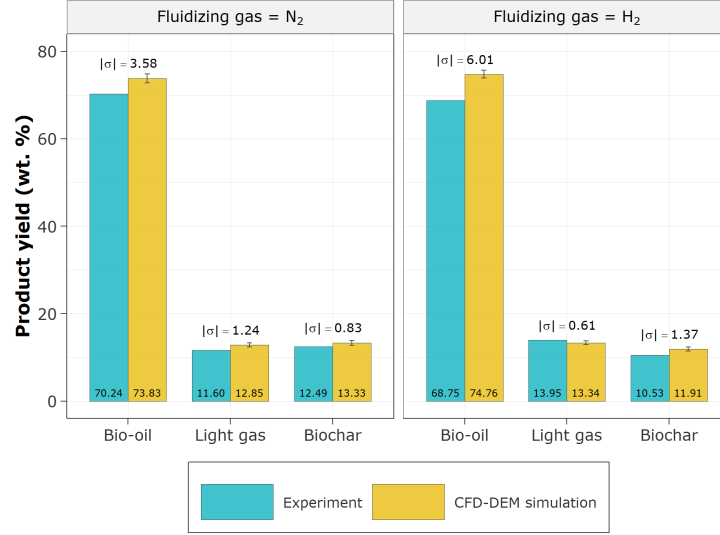


Figure 11: CFD-DEM simulation validation against experimental data. Product yields are calculated on a biomass basis. Deviation between experiment and simulation given by  $\sigma$ .

## 5.5 Fluidizing gas effect on pyrolysis performance

Figure 12 presents the volume-time averaged pressure drop and temperature along the height of the fluidized bed reactor. The different fluidizing gases considered in this research demonstrated similar effects on the pressure drop profile along the reactor height. Overall, the averaged bed height – as evidenced by the inflection point on the pressure drop curve – was about 0.14 m, regardless of fluidized gas. Similarly, the total pressure drop across the reactor was consistently about 1440 Pa for all fluidizing gases and mixtures. The volume-time averaged gas temperature ranged between 495°C and 500°C, depending on position along the reactor height and fluidizing gas. Gas temperature generally dipped around the biomass inlet and at the dense-bed/dilute-phase interface. The most noticeable trend in gas temperature occurs in the dilute-phase, with increasing gas temperature along the height of the reactor. Also noteworthy is the fact that gas temperature in the dilute-phase was highest when  $H_2$  was used as fluidizing gas. This observation is attributable to the large difference in the thermal conductivity of  $H_2$  and the other fluidizing gases (Figure XXX). The impact of the difference in the thermal conductivity of fluidizing gases is also evident in the average particle temperature and mass loss profile (Figure 13). When  $H_2$  was used as fluidizing gas, biomass particles experienced significantly higher heating rate, and consequently higher mass loss rate, compared to when other fluidizing gases were used. Biomass heating and mass loss rate follow the order:  $H_2 > CH_4 > H_2O > CO_2 > N_2 + CO_2 > N_2 > N_2 + CO > CO$ ,

irrespective of the initial size of the biomass particle.

Furthermore, it was observed that tar conversion reactions (Reactions 4 and 5) slightly changed among fluidizing gas used, with the lowest being  $N_2$  and the highest being  $H_2$  (Figure 14). This observation explains the reason why despite  $H_2$  yielded the highest particle heating and mass loss rate (Figure 13), and one of the longest residence times (Figure 15), its bio-oil yield relative to biomass flow rate is negligibly different from the bio-oil yield with other fluidizing gases, especially  $N_2$ . Nevertheless, the fact that we found that fluidizing gas can notably increase biomass heating and mass loss rate (pyrolysis conversion rate) suggest potential process intensification implication because increased heating and pyrolysis rate represents a system where pyrolysis can be completed at an increased rate and consequently offering increased system throughput. Our finding suggests that, at the least, fluidizing gas with produced light gases can be recirculated as fluidizing gas without detrimental consequences on pyrolysis performance.

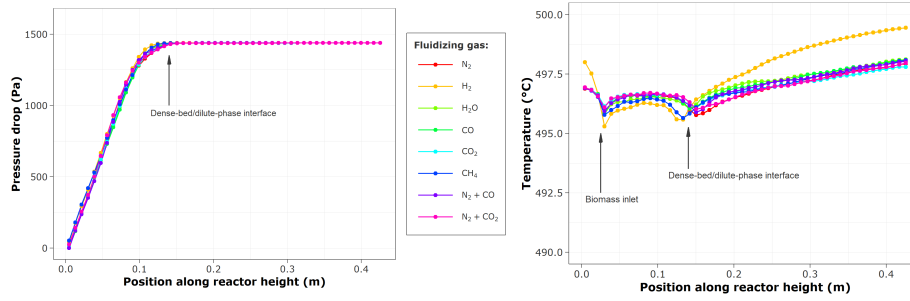


Figure 12: Time-averaged distribution of gas phase pressure drop (left) and temperature (right) along the reactor height.

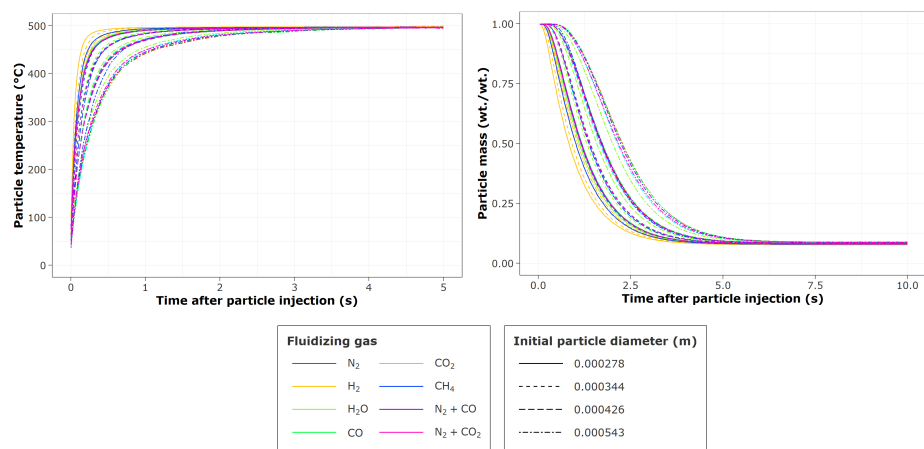


Figure 13: Average particle temperature (left) and mass loss (right) profile during pyrolysis. Line color discriminates among fluidizing gas, whereas line type discriminates among initial particle diameter.

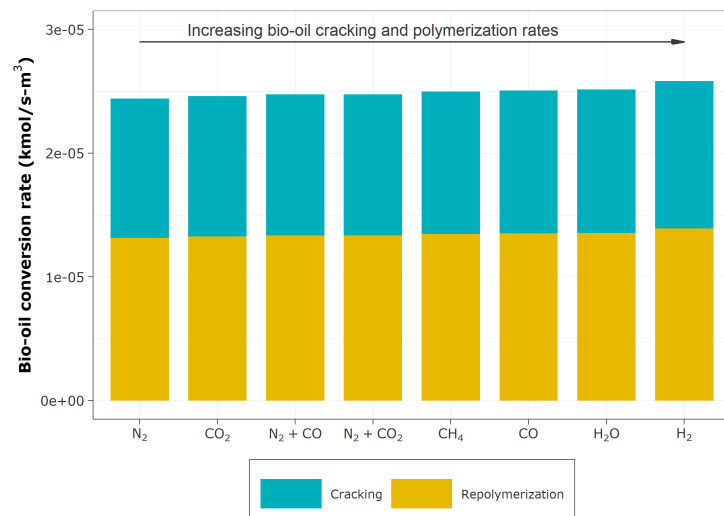


Figure 14: Time-average bio-oil cracking and polymerization rates.

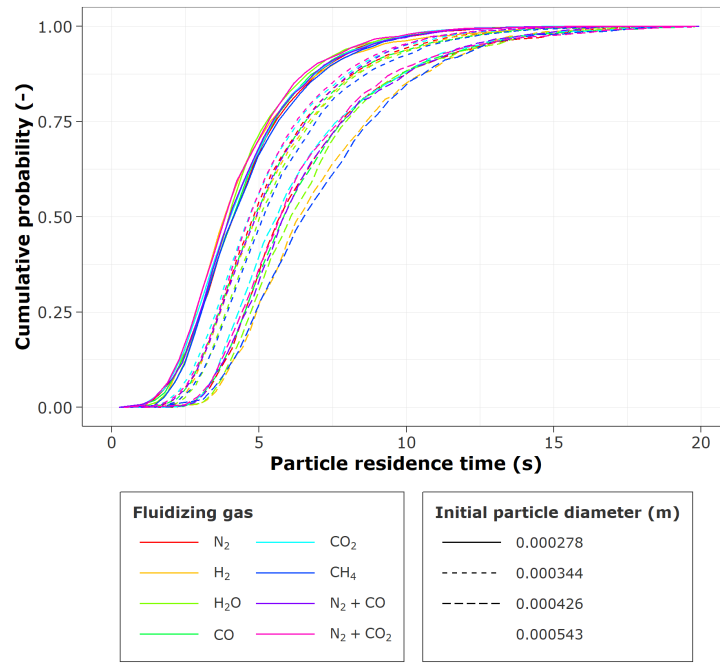


Figure 15: Cumulative particle residence time distribution as affected by fluidizing gas and initial particle diameter.

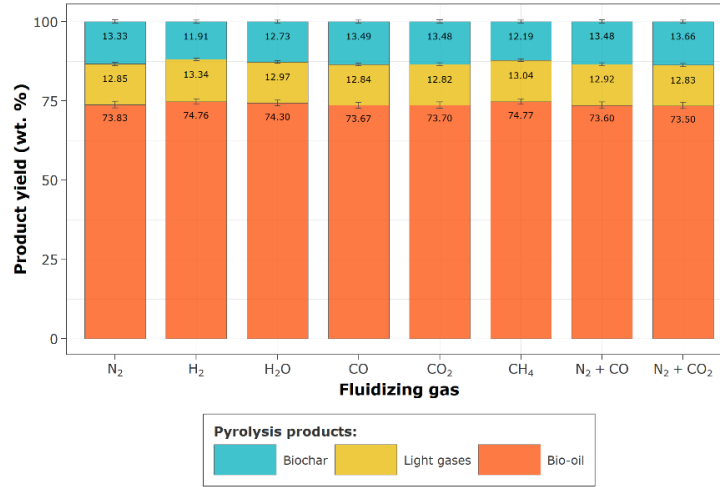


Figure 16: Pyrolysis product distribution as affected by fluidizing gas. Product yields are calculated on a biomass basis.

## 6 Conclusion

Here.

## 7 Source code

Python models used to generate results for this article are available on the CCPC GitHub at <https://github.com/ccpcode> in the X repository. Functionality provided by the Chemics package was used for gas properties and various fluidization calculations. See the Chemics documentation at <https://chemics.github.io> for more information.

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