

1 Development and validation of CFD-DEM coupling
2 interface for Heat & Mass Transfer using partitioned
3 coupling approach

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8 **Abstract**

9 This work demonstrates the rapid development of a simulation environment
10 to achieve Heat and Mass Transfer (HMT) between Discrete Element Method
11 (DEM) and Computational Fluid Dynamics (CFD). This coupling holds po-
12 tential for simulating various processes like drying, pyrolysis, combustion,
13 melting, and solid-fluid reactions, finding applications in biomass furnaces,
14 boilers, heat exchangers, and flow through packed beds among others. To
15 accurately model these applications, diverse CFD features and solvers must
16 integrate with DEM to capture intricate physics.

17 The proposed method employs the preCICE coupling library on volumet-
18 ric meshes, uniting CFD-DEM through an Eulerian-Lagrangian approach
19 for HMT. The prototype uses eXtended Discrete Element Method (XDEM)
20 for DEM calculations and OpenFOAM for CFD. XDEM receives key CFD
21 data fields through preCICE, setting particle boundary conditions based on
22 fluid domain properties and flow conditions. Heat and mass source terms
23 computed by XDEM fed into the CFD solver, representing the particle con-
24 tributions.

25 This coupling framework, comprising preCICE, XDEM, and its adapter,
26 accommodates a wide array of applications involving convective heat transfer
27 between particles and fluids. Validation includes comparisons with experi-
28 ments and a specialized solver, affirming the accuracy of predicted numerical
29 results across heat transfer, drying, and pyrolysis cases. Additionally, the
30 study delves into the computational costs associated with different coupling
31 approaches, offering valuable performance insights.

32 *Keywords:* Multi-physics, Coupled Simulations, CFD-DEM, Heat & Mass
33 Transfer, Partitioned Coupling

34 **1. Introduction**

35 The field of engineering faces problems related to multiphase media, which
36 may include a continuous phase such as fluids, and a discrete phase such as
37 powders, granular media, etc. Furthermore, these phases can behave and
38 interact on multiple scales. The engineering applications involving such com-
39 plexities are very difficult to study through experimentation. Therefore, such
40 complex multi-physics, multi-scale problems are usually studied via numeri-
41 cal simulations.

42 The problems involving such mixed media, cannot be resolved well by
43 only a continuous or a discrete phase alone. Such problems need to account
44 for both continuous and discrete media along with their interactions with
45 each other [1]. Such an approach is known as the Combined Continuum and
46 Discrete Model (CCDM) [2]. In the present work, we will deal with the Heat
47 and Mass Transfer (HMT) between the continuous fluid phase and discrete
48 particles phase. HMT between fluid and particles can be used to describe
49 processes such as drying, pyrolysis, combustion, gasification, and melting.
50 These processes have a wide variety of applications in industrial sectors such
51 as mining, energy production, waste management, pharmaceuticals, manu-
52 facturing & production, and process industries. Due to the challenges in per-
53 forming experiments, it is desirable to have an HMT multi-physics simulation
54 environment between particles and fluids to better capture these phenomena.

55 Such novel and rapidly evolving applications demand a rapid develop-
56 ment of a simulation environment. In the literature, to achieve CFD-DEM
57 coupling either commercial CFD software such as ANSYS Fluent®[3–8] is
58 coupled with commercial, open-source or in-house DEM software such as
59 Rock-DEM® [9]. Open-source software such as CFDEM (OpenFOAM +
60 LIGGGHTS) [10] are also utilized extensively for HMT applications [11, 12].

61 The CFD-DEM couplings mentioned above are achieved by solving dif-
62 ferent sets of segregated equations iteratively. This is ordinarily achieved by
63 a single code coupling, where all physics models are implemented in one code
64 also known as the monolithic coupling approach. Or they are coupled using a
65 partitioned coupling approach, that couples existing single-physics software
66 on a high level [13].

67 The monolithic coupling approach can be more robust when applied to
68 specific applications. Additionally, years of extensive research and devel-
69 opment are required to achieve such a simulation environment for a specific
70 application. An extensive review [14] of developments in CFD-DEM coupling
71 approaches for different applications demonstrates the same. Even with these
72 developments, a lot more research remains to be done. However, due to the
73 intrinsic nature of the monolithic coupling approach, it is rigid in its imple-
74 mentations. Moreover, such approaches rarely allow easy modifications or
75 adaptations to make the model closer to reality, or adapt for different applica-
76 tions. However, the partitioned coupling approach allows such modifications
77 or exchange of physical components.

78 In addition to the limitations mentioned above, due to the nature of
79 CFD-DEM coupling, the over-lapping domain (often the entire computa-

80 tional domain) is affected, and it is important to exchange the information
81 for coupling [15]. Furthermore, these simulations are very computationally
82 expensive. Hence, it is very important to parallelize such pieces of software.
83 To tackle these problems a co-located partitioning strategy is proposed [15,
84 16]. Although this strategy solves the problems mentioned above, it also has
85 limitations when dealing with non-conforming meshes due to the mesh/grid
86 alignments. The unresolved CFD-DEM coupling further adds restrictions on
87 the smallest CFD cell size, based on the largest particle size.

88 To circumvent the constraints of the monolithic coupling approach and
89 offer more flexibility, we employ the preCICE coupling library [17] to de-
90 velop a partitioned multi-physics simulation environment. In the partitioned
91 coupling approach, a multi-physics problem is decomposed into multiple
92 single physics parts and solved separately. The preCICE coupling library
93 can be then used to couple these new or existing (highly specialized, op-
94 timized, purpose-built) single physics solvers/software to achieve the said
95 multi-physics problem [18].

96 The preCICE coupling library treats these solvers/software as black-box
97 and enables communication, and data mapping strategies. This type of cou-
98 pling approach only needs nodal information from the black box. Subse-
99 quently, only standard Dirichlet and Neumann boundary conditions are ap-
100 plied [13].

101 Hence, there is no need to have access to source code, furthermore, no
102 need to have expert knowledge of the source code of each of the solvers/software
103 used in the partitioned multi-physics simulation. This also allows us to cou-
104 ple of solvers/software implemented in different languages (where currently
105 supported languages are C++, Python, MatLAB, Fortran, and Julia). Al-
106 though, to enable this communication and data mapping, a "preCICE cou-
107 pling adapter" needs to be developed. Such a development requires a basic
108 understanding of the solver/software along with its API. Hence the develop-
109 ment of a preCICE adapter is a fairly accessible and achievable task.

110 The preCICE coupling library [17] and its adapters [19, 20] have been
111 used to model Conjugate Heat Transfer (CHT) [21] between fluid and solid.
112 Volume coupling has been utilized to simulate fracturing in a poro-elastic
113 medium due to fluid flow [22]. Although the physical nature of coupling in
114 this work is volumetric, the coupled system uses surface terms for equilibrium.
115 The state-of-art on the CFD-DEM coupling is quite vast[14]. It is more
116 useful to bring attention to the previous works using OpenFOAM and XDEM
117 monolithic coupling [15, 23–27].

118 To rapidly establish an HMT simulations environment, our prototype cou-
119 ples OpenFOAM [28] with eXtended Discrete Element Method (XDEM) [1]
120 to achieve Heat & Mass Transfer between CFD and DEM. Although either
121 of the software mentioned can be replaced with an alternative due to the
122 modular nature of the coupling.

123 Our contributions, which are novel or related to the CFD-DEM cou-
124 pling strategies, are (1) a flexible partitioned CFD-DEM coupling approach
125 achieved by (a) developing an original preCICE adapter for XDEM (first
126 DEM preCICE adapter), (b) extending the OpenFOAM preCICE adapter
127 [29] to enable coupling over volumetric meshes, and mass transfer; (2) the
128 verification against monolithic coupling and validation against experimental
129 observations of the proposed partitioned CFD-DEM coupling approach; (3)
130 preliminary performance analysis of monolithic versus partitioned coupling
131 approach.

132 The paper is organized as follows: in section 2 we present the mathe-
133 matical modeling of CFD and DEM. In section 3, the partitioned coupling
134 strategies and software development are described. In section 4, we present
135 and compare numerical simulation results with experimental observations,
136 these cases include heating up, drying, and pyrolysis. In section 5, we com-
137 pare and discuss the performance of the partitioned coupling approach with
138 the monolithic coupling approach. Finally in section 6 we discuss the devel-
139 opment followed by conclusions.

140 2. Model Description

141 In the following section, the governing equations for continuum fluids
142 and discrete particles are presented. In the partitioned coupling, we couple
143 two single-physics software, namely CFD and DEM to achieve the multi-
144 physics CFD-DEM environment. These are presented in section 2.2 and 2.1
145 respectively. The partitioned coupling approach is described in the section
146 3. The partitioned coupling approach is compared with a legacy monolithic
147 coupling approach. Consequently, the reader is referred to the literature for
148 a detailed description of the XDEM + OpenFOAM legacy coupling [26, 27,
149 30].

150 2.1. Governing equations for discrete particles

151 XDEM predicts both the thermodynamics and dynamics of the particu-
152 late system. In the current work, the main focus will be the thermodynamics

153 of such particulate systems. The particle position, velocity, and acceleration
 154 are computed with the *dynamics* module of the XDEM, whereas the tem-
 155 perature and chemical processes are computed with the *conversion* module.

156 2.1.1. Conversion module

157 The conversion module of XDEM handles the heat and mass transfer
 158 within the particles and between the particles. It also accounts for various
 159 processes such as drying, gasification, combustion, etc. describing the inflow
 160 and outflow of the gas mixture. The detailed model description of the con-
 161 version module can be found in [1, 31], a summary of the governing equations
 162 for the fluid present in the porous regions within particles is given below.

163 Mass conservation equation for fluids in particle pores:

$$\frac{\partial}{\partial t} (\epsilon_f \rho_f) + \vec{\nabla} \cdot (\epsilon_f \rho_f \vec{v}_f) = m'_{s,f} \quad (2.1)$$

164 where $m'_{s,f}$ is the sum of all individual species' mass production or consump-
 165 tion rates due to chemical reactions, ϵ_f denotes the porosity within individual
 166 particles occupied by fluid(s). The fluid species transport within this porous
 167 space of the particle obeys Darcy's law:

$$-\frac{\partial p}{\partial r} = \frac{\mu_f \epsilon_f}{K} (\vec{v}_f) \quad (2.2)$$

168 One-dimensional transient energy conservation equations for particles:

$$\frac{\partial \rho c_p T}{\partial t} = \frac{1}{r_n} \frac{\partial}{\partial r} \left(r^n \lambda_{eff} \frac{\partial T}{\partial r} \right) - r^n (\vec{v}_f \rho_f c_{p_f} T) + \epsilon_f \sum_{k=1}^l \dot{\omega}_k H_k \quad (2.3)$$

169 The mass balance and transport equation of individual fluid species within
 170 the particle pores:

$$\frac{\partial}{\partial t} (\epsilon_f \rho_{f,i}) + \nabla \cdot (\epsilon_f \rho_{f,i} \vec{v}_f) = \frac{1}{r_n} \frac{\partial}{\partial r} \left(r^n \epsilon_f D \frac{\partial \rho_{f,i}}{\partial t} \right) + m'_{s,f,i} \quad (2.4)$$

171 The following boundary conditions apply to the governing equations men-
 172 tioned above:

$$-\lambda_{eff} \frac{\partial T}{\partial r} \Big|_{r=0} = 0 \quad (2.5)$$

$$-\lambda_{eff} \frac{\partial T}{\partial r} \Big|_{r=R} = \alpha(T_R - T_\infty) + q''_{rad} + q''_{cond} \quad (2.6)$$

174

$$-D_{i,eff} \frac{\partial \rho_i}{\partial r} \Big|_{r=R} = \beta_i (\rho_{i,R} - \rho_{i,\infty}) \quad (2.7)$$

175 In the Eq 2.6, q''_{cond} and q''_{rad} are conduction and radiation heat sources
 176 respectively from the neighboring particles. A detailed description of the
 177 conduction and radiation between particles is given by B. Peters in [31].

178 To solve for heat & mass transfer within the particle, the particle radius
 179 is discretized. This radial discretization can be uniform or non-uniform, as
 180 shown in fig 1. In the present work, uniform radial discretization is used. The
 181 non-uniform radial discretization allows for having a smaller cell length near
 182 the particle surface that allows the model to capture the sharp temperature
 183 and mass flow gradients.

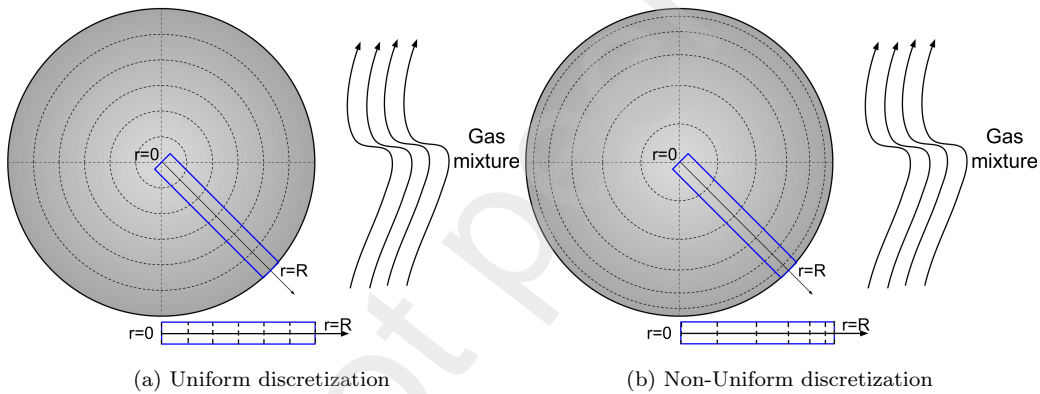


Figure 1: Radial discretization for heat & mass transfer calculations within a particle

184 *2.1.2. Dynamics module*

185 The discrete element method used in the dynamics module of XDEM
 186 is based on the soft sphere model. In this method, it is assumed that the
 187 particles are deformable and can overlap each other, where the magnitude of
 188 overlap is decided by the contact force using the force-displacement law. The
 189 hardness of the particle is expressed via Young's Modulus, while the particle
 190 energy dissipation is described with a dampener and/or dashpot. The trans-
 191 lational and rotational movements of individual particles are tracked using
 192 classical mechanics equations. A detailed description can be found in [23].

193 The scope of the present work is heat and mass transfer, hence the cases
 194 chosen for this study do not have particle(s) in motion. However, if the fluid

195 velocities were to be increased, particles could move due to the momentum
 196 transfer. The fluid velocity is modeled as one of the external forces F_i^{ext} . An
 197 example of such a case can be found in the modeling of a raceway zone in a
 198 blast furnace [32].

199 A summary of the translational and rotational motion equations is given
 200 below:

201 Equations of particle motion, where $F_i^{\vec{ext}}$ is the sum of all the external forces
 202 acting on the particle, such as buoyancy forces \vec{F}_B and drag forces \vec{F}_D :

$$m_i \frac{d\vec{v}_i}{dt} = m_i \frac{d^2 \vec{X}_i}{dt^2} = \vec{F}_i^c + \vec{F}_i^g + \vec{F}_i^{\vec{ext}} \quad (2.8)$$

203

$$I_i \frac{d\vec{\omega}_i}{dt} = \sum_{j=1}^n \vec{M}_{i,j} \quad (2.9)$$

204 2.2. Governing equations for fluid

205 In the Eulerian volumetric average method, the conservation equation of
 206 mass (Eq 2.10), momentum (Eq 2.11) and energy (Eq 2.12) are written over
 207 a representative volume.

208 Conservation of mass:

$$\frac{\partial}{\partial t} (\rho_f) + \nabla \cdot (\rho_f \vec{v}_f) = m' \quad (2.10)$$

209 Conservation of momentum:

$$\frac{\partial}{\partial t} (\rho_f \vec{v}_f) + \nabla \cdot (\rho_f \vec{v}_f \vec{v}_f) = -\nabla p + \rho_f \vec{g} + \mu \nabla^2 \vec{v}_f + \vec{S} \quad (2.11)$$

210 Conservation of energy:

$$\frac{\partial}{\partial t} (\rho_f h_f) + \nabla \cdot (\rho_f \vec{v}_f h_f) = \frac{\partial p}{\partial t} + \vec{v}_f \cdot \nabla p + q' \quad (2.12)$$

211 Mass conservation equation for chemical species i in CFD is given as
 212 follows in Eq 2.13

$$\frac{\partial}{\partial t} \rho_{f,i} + \nabla \cdot (\rho_{f,i} \cdot \vec{v}_f) = m'_i \quad (2.13)$$

213 In the XDEM + OpenFOAM legacy coupling [26, 27, 30], the governing equa-
 214 tions of the fluid contain a term ϵ (porosity), where porosity (ϵ Eq 2.14) refers

215 to the interstitial space between the solid particles. The porosity calculation
 216 in brief is as follows:

$$\epsilon = 1 - \frac{1}{V_c} \sum_i^n \eta_i V_i \quad (2.14)$$

217 where V_c is the volume of the cell in consideration, V_i is the volume of each
 218 particle multiplied by η_i denoting the amount of particle volume present in
 219 the current volume.

220 This porosity term is not directly included in the current CFD model, as
 221 the fluid solver needs to be modified and tested thoroughly. This process is
 222 highly intrusive and defeats the rapid development of the HMT simulation
 223 environment.

224 Thus when computing the heat and mass source terms in the XDEM, the
 225 porosity is taken into account. Furthermore, this porosity is exchanged as a
 226 field. Subsequently, it is used to model the drag offered by the particles to
 227 the fluid as follows [33]:

$$\kappa = \frac{d_{pmean}^2 \epsilon^3}{150(1 - \epsilon^2)} \quad (2.15)$$

228 where d_{pmean} is the mean particle diameter.

$$C = \frac{1.75(1 - \epsilon)}{d_{pmean} \epsilon^3} \quad (2.16)$$

229

$$drag = \frac{\mu}{\kappa} + \rho C \vec{v}_f \epsilon \quad (2.17)$$

230 3. Partitioned Coupling Implementation with preCICE

231 A flexible multi-physics simulation environment is achieved through the
 232 preCICE coupling library due to its minimal invasion of the solvers through
 233 the usage of high-level API (Application Programming Interface). This in-
 234 tegration of preCICE into the solver is known as an "adapter" [34], seen in
 235 a schematic in figure 2. For a well-developed in-house, open-source, or any
 236 other kind of solver, an API for the solver is generally available. Alterna-
 237 tively, solvers developed in-house are well understood and can be developed
 238 to facilitate data field communication through preCICE. By utilizing the API
 239 from the solver and preCICE, the solver code remains unchanged, and the
 240 adapter can be easily implemented and compiled as a separate library called
 241 by the solver during runtime. During coupled simulation, the solver passes

242 the required data via its adapter to preCICE, which in turn communicates it
 243 to the other coupled solver(s) using MPI messages or TCP/IP sockets. A list
 244 of data fields exchanged for the HMT CFD-DEM coupling for the current
 245 work is presented in table 1.

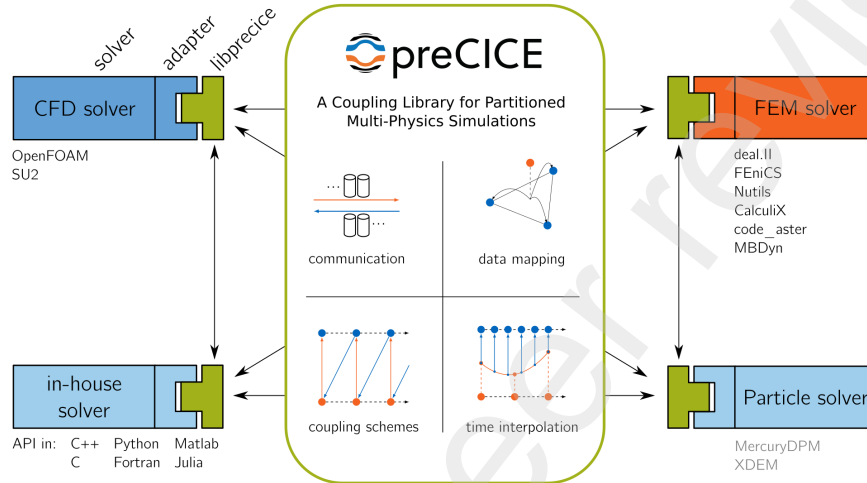


Figure 2: A schematic outlining the coupling procedure [18] (reproduced with permission)

246 3.1. OpenFOAM Adapter for preCICE coupling

247 The OpenFOAM adapter [29] is already available for surface coupling.
 248 It is used in different examples and applications such as Conjugate Heat
 249 Transfer (CHT) and Fluid-Structure Interactions (FSI) that can be achieved
 250 when coupled with other software. Although the default adapter contains
 251 all the fluid fields required to achieve HMT between CFD-DEM, they are
 252 described on surfaces.

253 To enable CFD-DEM volume coupling, a new coupling interface is imple-
 254 mented in the OpenFOAM adapter. Four different modules are implemented.
 255 These modules enable an exchange of different data fields related to Fluid
 256 Properties, Momentum Transfer, Heat Transfer (HT), and Mass Transfer
 257 (MT). Depending on the type of simulation these modules can be switched
 258 on or off (similar to the pre-existing modules).

259 Data fields such as fluid density, viscosity, thermal conductivity, and spe-
 260 cific heat are added to the Fluid properties module. The fluid temperature
 261 and heat source fields are added to the HT module, whereas chemical species
 262 mass fractions, mass source, and species mass source are added to the MT

263 module. Fluid velocity and pressure fields are exchanged via the Momentum
264 transfer module. The user is free to select which data fields they want to ex-
265 change, and what modules to use. Depending on the simulation type, these
266 data fields are communicated via preCICE and the adapters to the other
267 solver(s) and make them available for calculation.

268 In addition, OpenFOAM adapter receives various source fields such as
269 heat (q'), mass (m'), and species (m'_i) through their respective modules.
270 These source fields are then injected into the respective governing equations
271 through the finite volume plugin `fvOptions` of OpenFOAM. As the presence
272 of particles in the fluid is only represented by source terms in the finite
273 volume options, there is no need to modify any of the OpenFOAM solvers to
274 accommodate this CFD-DEM coupling. In the present work, `rhoPimpleFoam`
275 (HT) and `reactingFoam` (HT & MT) solvers provided by OpenFOAM are
276 used.

277 In practice, when using OpenFOAM, one only needs to change the CFD
278 solver name in `controlDict`. If this solver has the fields required as men-
279 tioned in table 1, no more work is required to run a CFD-DEM multi-physics
280 simulation. In broad scope, it is also possible to switch between differ-
281 ent OpenFOAM versions seamlessly to avail of different functionalities and
282 solvers. With some more effort, one can also implement an adapter for an
283 in-house CFD solver, and couple it with the needed solver (XDEM in this
284 case).

285 3.2. XDEM Adapter for preCICE coupling

286 Similar to preCICE, XDEM is also implemented in C++, thus when
287 implementing the XDEM adapter for preCICE, we utilize C++ API of pre-
288 CICE. An XDEM coupling interface class is implemented and then utilized
289 to access data fields from the XDEM adapter.

290 XDEM adapter is developed to be flexible for diverse types of simula-
291 tion. Similar to some other preCICE adapters provided by preCICE, the
292 XDEM adapter is developed so that one can choose what fields are to be
293 exchanged. If required fluid fields are not exchanged, default values are used
294 for required calculations. XDEM adapter provides a summary of the data
295 fields exchanged and possible types of simulation being run based on the data
296 fields used. XDEM adapter receives several fields describing fluid properties
297 and flow conditions. These values are then used as boundary conditions on
298 the particles. In the context of current work, XDEM offers several HT laws
299 and MT laws [35–38]. These are set through the XDEM input file. These

Data Fields	CFD →DEM	DEM →CFD
Fluid Temperature	✓	
Fluid Viscosity	✓	
Fluid Conductivity	✓	
Fluid Specific Heat	✓	
H ₂ O	✓	
O ₂	✓	
N ₂	✓	
⋮	✓	
species n	✓	
Heat Source (q')		✓
Mass Source (m')		✓
Heat Transfer Coefficient		✓
Volume Porosity (ϵ)		✓
source H ₂ O (m'_{H_2O})		✓
source O ₂ (m'_{O_2})		✓
source N ₂ (m'_{N_2})		✓
⋮		✓
source species n (m'_n)		✓

Table 1: The data fields that are exchanged for the heat and mass transfer coupling

300 HT and MT laws are then utilized to compute the heat, mass, and chemi-
301 cal species source terms. Depending on the species mass concentrations and
302 fluid flow conditions. XDEM also performs species transport and Solid-Fluid
303 reactions.

304 These source terms are then transferred to the CFD solver through pre-
305 CICE. The XDEM coupling interface class and the XDEM adapter are de-
306 signed in such a way that ideally we can switch between any desired CFD
307 solver/software. It does not make an assumption the kind of CFD solver
308 used, rather it just assumes it receives and sends some specific fields that can
309 be configured. Thus providing flexibility in choosing a CFD solver based on
310 the application.

311 3.3. Mapping methods over Volumetric mesh

312 In the current work, we achieve the HMT coupling over volumetric meshes.
313 Normally, we already have a volumetric mesh for the CFD. In contrast, DEM

314 is a mesh-less method. In the XDEM suite, the DEM simulation domain is
 315 defined by a simple box, and individual particles are tracked within this box.
 316 An example of such a domain can be seen in figure 15 (b). However, due to
 317 DEM methods being costly, they require some parallelization. To this end,
 318 the simple box in the XDEM suite can be discretized over the three axes.
 319 Figure 16 (b) shows the DEM domain discretized. The domain is sliced to
 320 reveal the cells and cell-centres. Consequently, the CFD cell size does not
 321 depend on the largest particle diameter for the unresolved coupling presented
 322 in this work.

323 The numerical experiments presented to use the default mapping of-
 324 ferred by the preCICE coupling library. The **nearest projection** map-
 325 ping method is applied when mapping data from CFD to DEM. In con-
 326 trast, **nearest neighbor** mapping method is applied when mapping data
 327 from DEM to CFD. The **nearest-projection** mapping method is mostly
 328 a second-order method. This method first projects the data onto the mesh
 329 and uses linear interpolation within each element [39]. An illustration of this
 330 method can be seen in the top half of figure 3. This method requires the
 331 mesh connectivity information. The **nearest neighbor** mapping method
 332 is a first-order method as presented in the lower half of the figure 3. The
 333 cases under consideration do not warrant complex mapping methods such as
 334 **nearest neighbor gradient** or **radial basis function**. Although these
 335 mapping methods are available in the preCICE coupling library.

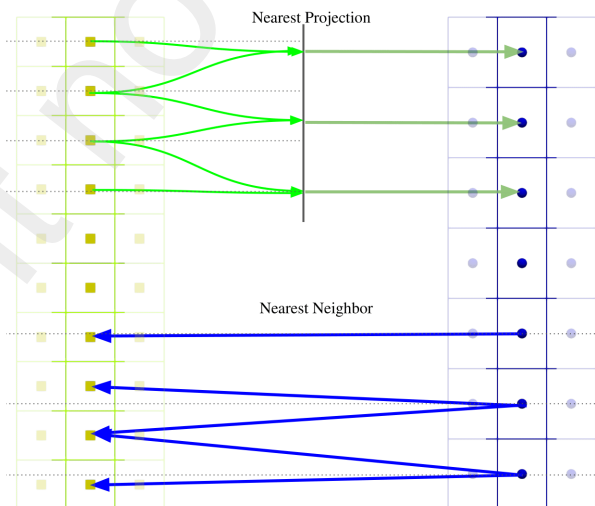


Figure 3: A schematic showing the two data mapping strategies used

336 Further, the mappings have two different types of constraints so as to ac-
337 count for mapping between non-conforming meshes. These are **consistent**
338 and **conservative** [39]. The **conservative** mapping constraint aggregates
339 the data to be mapped such that the total amount of data coupled is the
340 same on the two meshes. In the present work, we apply the **conservative**
341 mapping constraint when mapping data from DEM to CFD. This ensures
342 that the exactly same amount of heat, mass, and/or species source generated
343 by the particles is injected into the CFD domain. Although, one has to be
344 careful with non-conforming meshes. More discussions on how this mapping
345 constraint affects the simulation results are in sections 4.4.1 and 6. The
346 **consistent** mapping constraint is applied when mapping data from CFD
347 to DEM. This constraint is applied for physical quantities such as tempera-
348 ture or velocities. This mapping constraint will apply the exact value seen
349 on the originating mesh. In the applications under consideration, this map-
350 ping constraint can be very useful as the CFD-DEM coupling is unresolved.
351 Depending on the mesh size differences, it can be enough to know the fluid
352 conditions corresponding to the CFD cell center closest to the particle center.

353 This constraint can be quite limiting if the mesh size difference between
354 CFD and DEM is large. This mapping constraint can also be limited if we
355 see large gradients in physical values over a distance shorter than particle
356 diameter. These limitations can be counteracted by employing the **radial**
357 **basis function** mapping instead of either methods mentioned above. Al-
358 though this method will give more accurate mapping and, thus more accurate
359 simulation results, this method is more costly. In the current work, the CFD
360 and DEM cell size disparity is almost non-existent (for single particle cases)
361 or very minor (for packed bed case).

362 *3.4. Coupling Strategies*

363 It is very important to consider the type of coupling needed for a given
364 problem. A further restriction of the monolithic coupling approaches is that
365 the application hence the type of coupling strategies are predefined. This can
366 lead to two issues, either the coupling strategy used is okay but might lead
367 to additional costs or the coupling strategy is ill-suited for a new problem.

368 In this section, we briefly discuss the coupling strategies available for
369 the presented partitioned coupling. There are two main distinct coupling
370 strategies available for partitioned coupling approaches, broadly known as
371 explicit and implicit. The explicit coupling strategy executes and calls the
372 coupled solvers for a set number of coupling time steps, whereas the implicit

373 coupling strategy is used when either the numerical solution is unstable or
374 there is a need to completely capture the coupled solution. The reader is
375 referred to the literature for in-depth reading [13, 39]. In the numerical
376 experiments presented in the next sections, we only use the explicit coupling
377 strategy. However, using the implicit coupling strategies could be used just
378 as easily without any additional work.

379 3.5. Execution Strategies

380 Although parallelization and scalability are not the primary focus of this
381 work, due to the computational costs involved in discrete methods, this issue
382 is bound to come up. The current partitioned coupling approaches allow for
383 the rapid development of heat and mass transfer simulation environments.
384 Although this might allow to establish a multi-physics simulation environ-
385 ment required for a certain application, it is also important that it is scalable
386 for the intended application. These simulations can be quite large when
387 considering industrial scale applications.

388 In the classical monolithic coupling approach, each set of equations de-
389 scribing one of the physics involved is executed consecutively or serially.
390 These equations might be parallelized, but their execution is serial. This
391 leads to the computing resources being idle. The different execution strate-
392 gies are illustrated in figure 4. Figure 4 (a) and (b) illustrates respectively the
393 execution of the monolithic coupling and the serial execution of partitioned
394 solvers. In terms of solving the equations, they tend to behave similar to
395 each other. However, in our monolithic approach, the two physics are cou-
396 pled into a single executable, they share the data on the memory, that is used
397 to exchange the information between the solvers. On the contrary, for the
398 partitioned execution, the exchange of data is handled through the preCICE
399 coupling library leading to extra memory copies and communications.

400 Figure 4 (a) and (b), in terms of the execution, there is not much difference
401 apart from the way the information is shared. The big difference comes in
402 the parallelization of individual solvers. In the monolithic approach, as the
403 equations are intermingled, the domain needs to be divided exactly the same
404 way for both the solvers, hence the co-located partitioning approach [16]. In
405 contrast, for the partitioned coupling approach, the domains for each solver
406 can be divided as the need arises. This feature plays an important role when
407 there are non-conforming domains/meshes involved in the coupling, which is
408 the case for most real-world applications.

409 Furthermore, figure 4 (a) and (b) demonstrates another fatal flaw of this
410 type of execution, i.e. wasting computing resources by idling. As the solvers
411 execute one after the other, one solver always has to wait for the other solver
412 to finish. Consequently, a parallel execution strategy can be utilized to avoid
413 this problem, as illustrated in figure 4 (c). Here, both solvers are executed
414 simultaneously. In this example, we see that the DEM solver requires more
415 time, hence the CFD solver computing resources stay idle, but the overall
416 idle time as compared to the serial execution is less. This problem of idling
417 can also be solved by further load balancing.

418 4. Results

419 In the present work, we use simple fundamental test cases to demonstrate
420 the robustness of the partitioned HMT coupling between CFD and DEM.
421 Along with the simple cases, we also study the drying process of a packed
422 bed [40] to demonstrate the coupling with a large number of particles. To
423 validate and verify the coupling, we compare the simulation results from
424 the current coupling methodology with experimental results and simulation
425 results from legacy CFD-DEM (XDEM). The current work only focuses on
426 convective heat transfer between particles and fluid, as inter-particle heat
427 transfer has been extensively studied in previous work [30, 33]. Conduction
428 and radiation between particles and fluid can also be modeled similarly, but
429 for the current cases their contribution is insignificant, hence we ignore these
430 heat sources.

431 The XDEM + OpenFOAM preCICE coupling uses OpenFOAM v7 [28].
432 However, XDEM + OpenFOAM legacy coupling uses FOAM-Extend v3.2 [41],
433 which is a fork of OpenFOAM. This different implementation might lead to
434 minor numerical differences in the results. The software used for legacy
435 coupling are a modified version of foam-extend 3.2 (git hash *3912d19b*) and
436 XDEM (git hash *fd06b8a0*). The preCICE coupling uses OpenFOAM 7,
437 XDEM (git hash *a6f0b7f9*) and preCICE 2.5.0. The simulations are carried
438 out on the *Aion* cluster at the University of Luxembourg that offers 354 com-
439 puting nodes, consisting of two AMD Epyc ROME 7H12 2.6Ghz processors
440 accounting for 128 cores per computing node, each equipped with 256 GB of
441 memory.

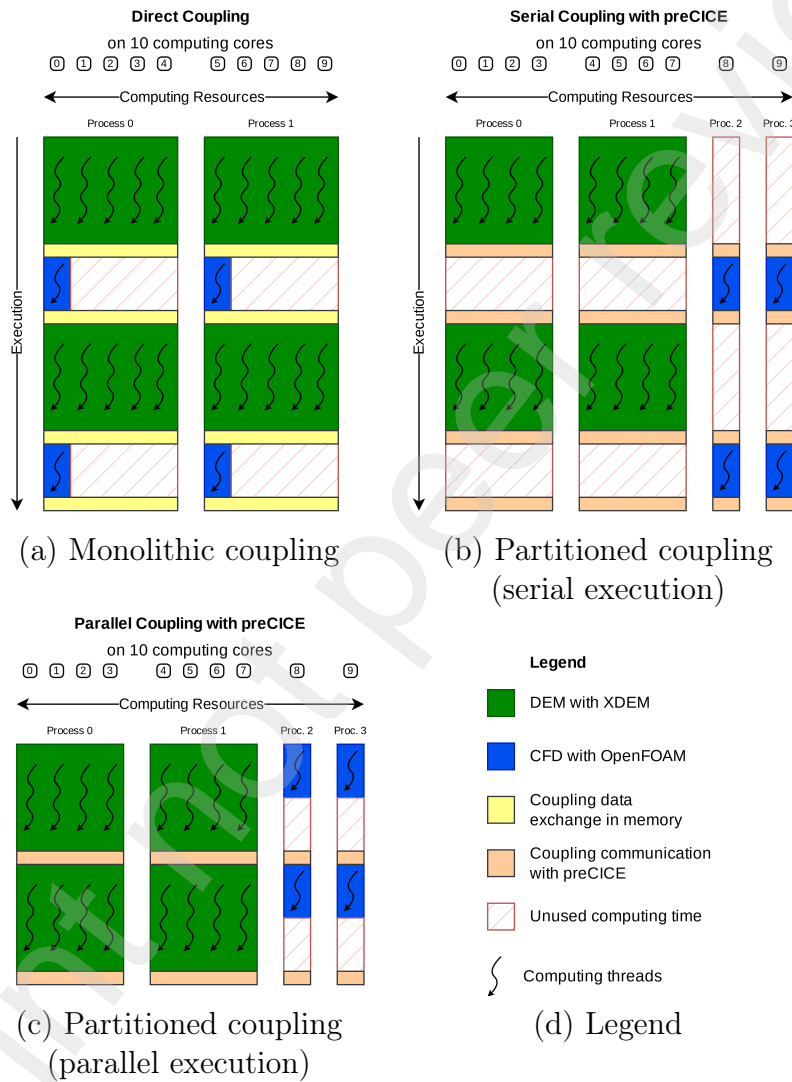


Figure 4: Comparison of the coupling and execution strategies

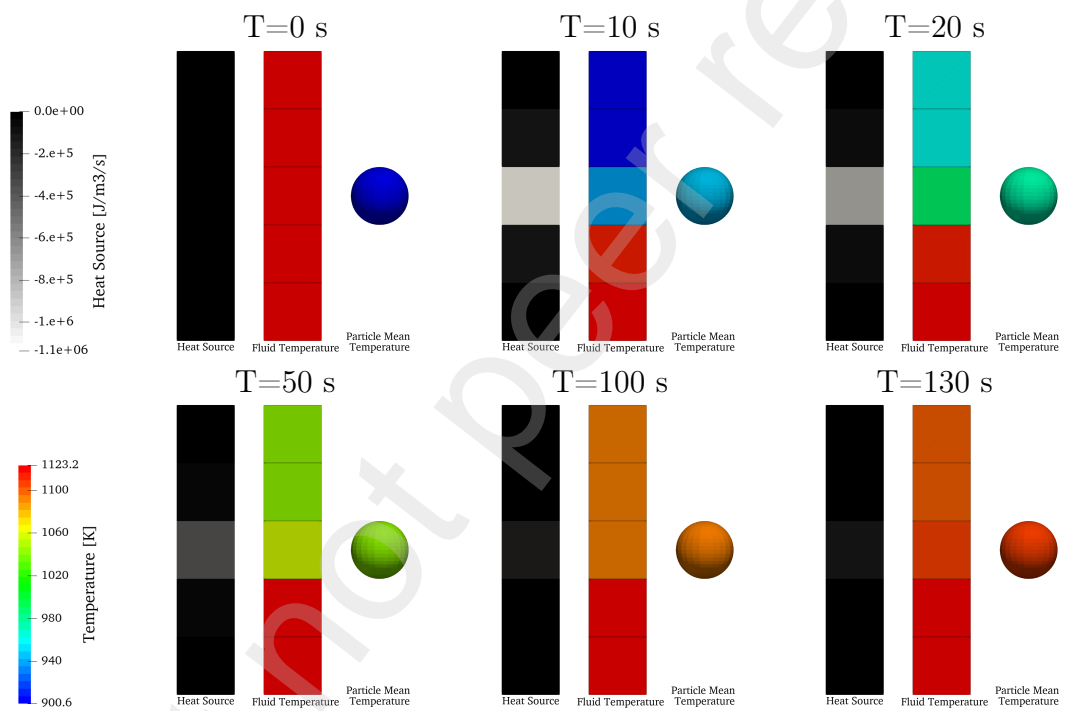


Figure 5: Fluid fields demonstrate the effect of the presence of cold particle heating-up

442 *4.1. Heat Transfer only: Single particle heat-up*

443 In the heat transfer only case, we consider one particle at room temper-
444 ature heating up due to the hot air surrounding it. The CFD domain is
445 $0.02 \times 0.02 \times 0.1$ m in size, discretized into 5 cells in the vertical direction
446 only (uniform 3D Grid $1 \times 1 \times 5$). The air inside the fluid domain is at 1123
447 K and atmospheric pressure is $1e + 05$ Pa. The air enters from the bottom
448 of the CFD domain with 0.38 m/s and a temperature of 1123 K, mimicking
449 the experimental setup in [42]. The air exits the CFD domain from the top.

450 The DEM domain contains a dry spherical Beech wood particle of di-
451 ameter 0.02 m, with wood properties found in Table 2 [25]. The particle
452 is discretized radially into 30 uniform segments for 1D HMT computations
453 within the particle. The particle is at 300 K at the beginning of the simula-
454 tion. The particle is located at (0.01, 0.01, 0.05) m and it remains stationary
455 throughout the simulation.

Properties	Beech wood [40]	Fir wood [43]
Density ρ (kg/m ³)	750	330
Porosity ϵ (-)	0.64	0.6
Pore diameter	50×10^{-6}	50×10^{-6}
Specific Heat c_p (J/kg K)	2551.3	1733
Conductivity λ (W/m K)	0.47	0.2

Table 2: Physical properties of the wood particles

456 *4.1.1. Heat transfer case results*

457 Figure 5, we see the temporal evolution of the CFD domain and particle
458 surface temperature. Figure 5 and 6 show that the particle uses thermal
459 energy from fluid to heat up. This drain of thermal energy leads to a drop
460 in the air temperature downwind. As the particle heats up, the rate of heat
461 transfer drops, and we see that the air temperature downwind gradually
462 increases, although it remains somewhere between the CFD inlet temperature
463 and the particle surface temperature. From figure 6, we see that the particle
464 surface temperature comes close to the fluid temperature. These results
465 demonstrate how two-way HMT coupling works, as we see the effects of
466 fluid conditions on the particle and the effect of the particle on the fluid

467 temperature field. Figure 6 also shows the drop in fluid outlet temperature.
468 The sudden initial drop-off in fluid outlet temperature is because the results
469 for fluid are recorded every 10 s, starting at 10 s.

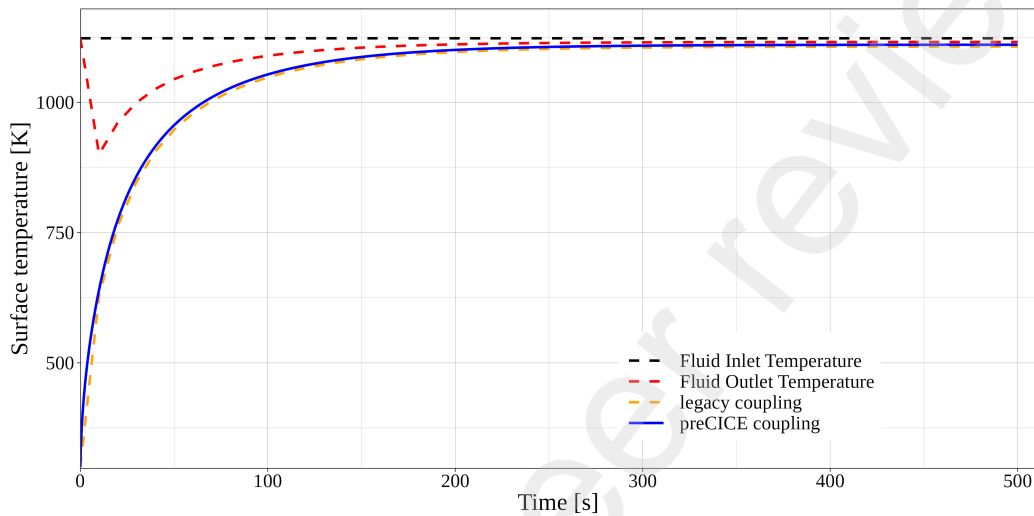


Figure 6: Influence on fluid temperature due to the presence of particle plotted along with the surface temperature of the particle

470 Figure 6 and 7 show the particle surface and center temperature evolution
471 over time. The current results are compared with the XDEM-OpenFOAM
472 legacy coupling which has been thoroughly verified and validated against ex-
473 perimental results. Figure 6 and 7 shows that the temperature evolution of
474 the particle for XDEM + OpenFOAM preCICE coupling is in good agree-
475 ment with XDEM + OpenFOAM legacy coupling.

476 We can see that the temperature profile is in very good agreement, but
477 we see a minor difference in the numerical results. This is because we use
478 different OpenFOAM versions.

479 Figure 6 and 7 shows that the XDEM + OpenFOAM preCICE coupling
480 simulation results, specifically the particle surface and particle center tem-
481 peratures are in good agreement.

482 As there is no experimental data for the heating up of a single wood par-
483 ticle, we use the experimental results by Petek [42]. The simulation setup
484 mimics the experimental setup, the only difference being the current case
485 does not simulate any chemical reactions (pyrolysis). In figure 8, the sim-
486 ulated surface temperature of the particle closely follows the experimental

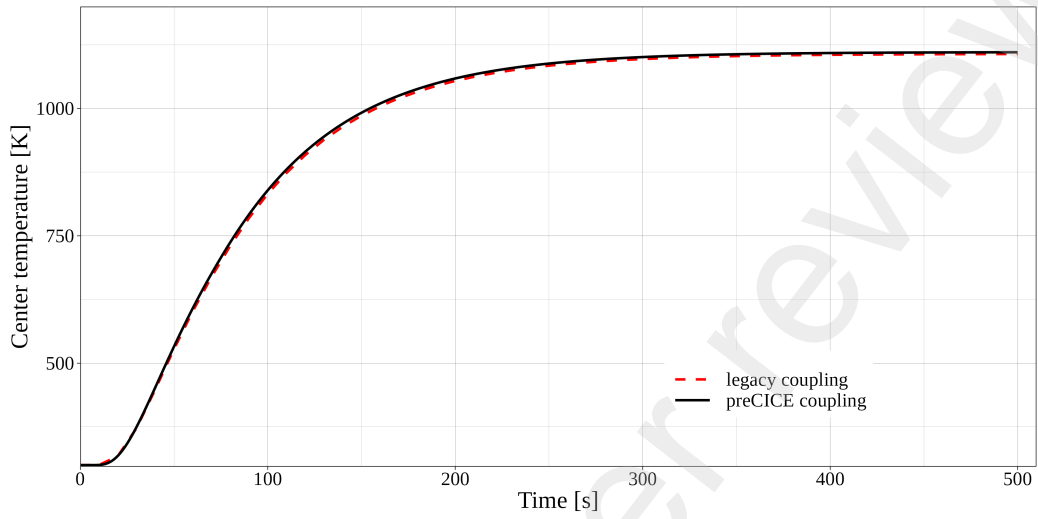


Figure 7: Temperature at the center of the particle compared for two different couplings

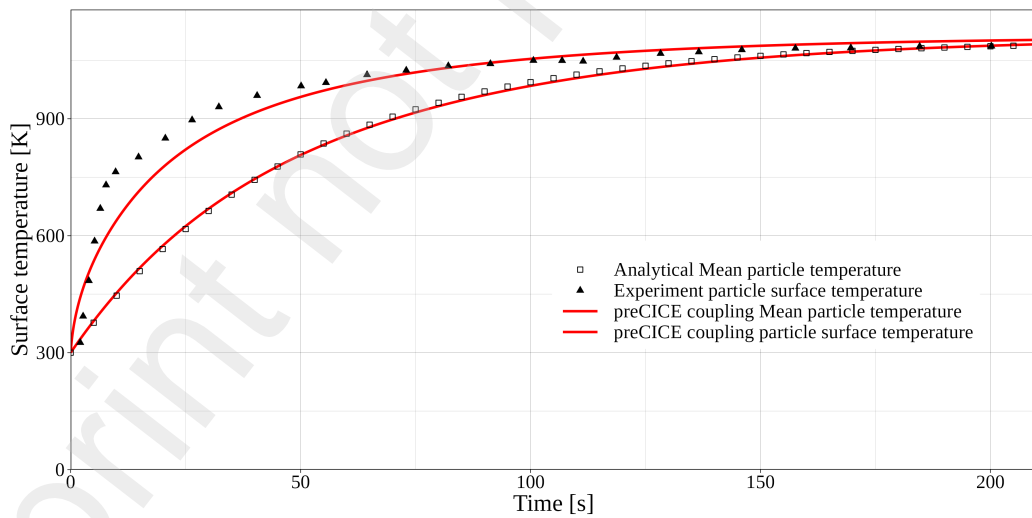


Figure 8: Particle surface and mean temperatures compared against experimental observations [42] and analytical solution resp.

487 observations. This is expected as the majority of the chemical reactions are
488 taking place within the particle as compared to on the particle surface. In
489 the experiment, the particle undergoes pyrolysis (an endothermic reaction),
490 thus we see lower temperatures in the numerical simulations as compared to
491 experimental observations for the time range 0 s & 75 s.

492 Furthermore, as the presented case is simple, we utilize the same initial
493 and boundary conditions and get an analytical solution for the heat-up of
494 the particle. As this is an analytical solution, and the particle diameter is
495 not discretized as shown in figure 1, thus we only have the analytical solution
496 for the overall particle temperature. The numerical result of the mean par-
497 ticle temperature is compared with the analytical solution in figure 8. The
498 numerical results are in good agreement with the analytical solution.

499 *4.2. Heat & Mass Transfer: Drying of Fir wood particle*

500 In the previous section, we establish that the 2-way heat transfer between
501 the CFD and DEM is working well. In the current section, we want to see
502 the effects of this heat transfer on the composition of the particle. Particle
503 drying is selected to validate the mass transfer as the moisture content in the
504 particle and water vapor after evaporation stays stable, i.e. does not react
505 with the surrounding fluid. Thus it is easy to track, in experiments as well
506 as in simulations.

507 In the heat and mass transfer case, we consider drying a spherical Fir
508 wood particle with properties given in Table 2 [43] with some moisture con-
509 tent. The simulation set-up conditions mimic the experimental setup by B.
510 Peters [43]. The experiments were performed with Fir wood particles with
511 33% and 66% moisture content. In the current work, we perform two simu-
512 lations with these two different initial moisture content. The CFD domain
513 is $0.15 \times 0.15 \times 0.5$ m in size, discretized uniformly as $3 \times 3 \times 10$. The air
514 inside the fluid domain is at 743 K and atmospheric pressure is $1e+05$ Pa, air
515 enters from the bottom of the CFD domain with 0.28 m/s and a temperature
516 of 743 K. The air exits the CFD domain from the top.

517 The DEM domain contains a Fir wood particle of diameter 0.008 m,
518 located at (0.075, 0.075, 0.125) m. The particle is discretized radially into
519 21 uniform segments for 1D HMT computations within the particle. The
520 particle is at 297 K at the beginning of the simulation.

521 In the current study, the heat sink model (constant evaporation model)
522 is applied for the calculation of drying rate [40, 43]. The model is described

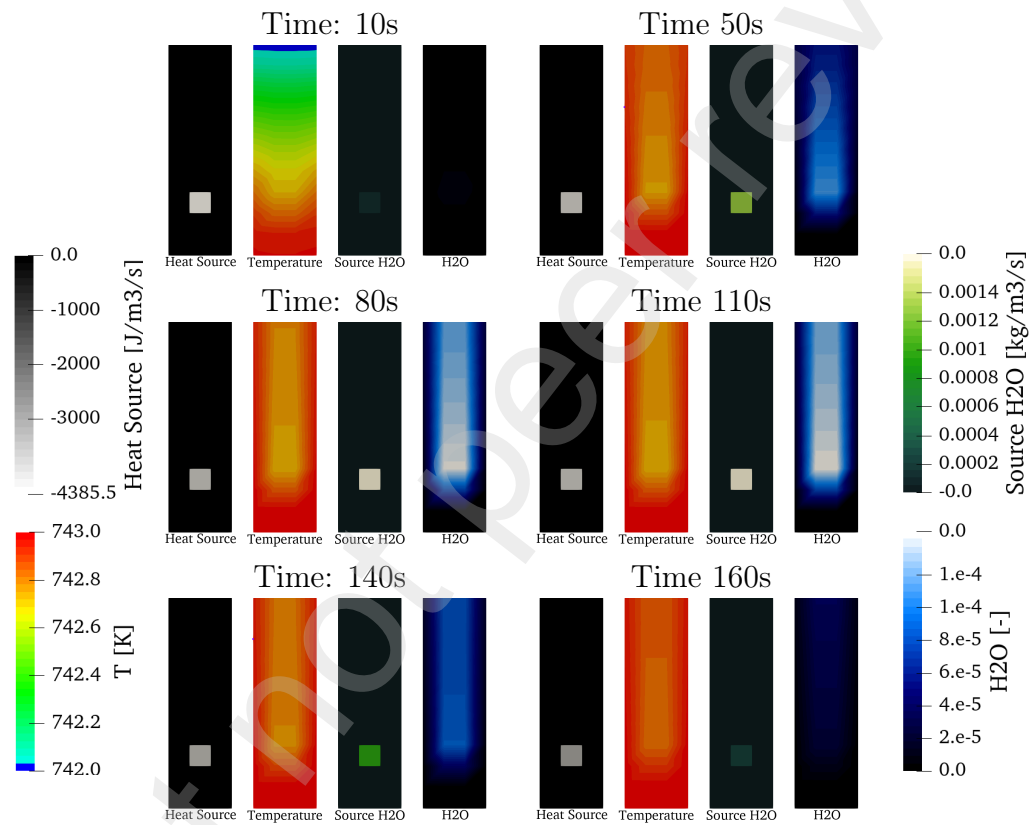


Figure 9: Evolution of heat source, fluid temperature, water vapor source, and water vapor mass fraction over time in the CFD domain showing drying process of wet particle.

523 as follows:

$$\dot{w}_{H_2O} = \begin{cases} \frac{(T-T_{evap})\rho c_p}{H_{evap}\delta t} & \text{if } T \geq T_{evap} \\ 0 & \text{if } T \leq T_{evap} \end{cases} \quad (4.1)$$

524 where ρ and c_p are the density, and thermal capacity of the dry wood, H_{evap}
525 is the evaporation enthalpy. In this drying model, the evaporation tempera-
526 ture T_{evap} is utilized for evaporation without distinguishing between free and
527 bound water.

528 4.2.1. Heat and mass transfer: Drying case results

529 Figure 10 shows a comparison of the moisture content of numerical sim-
530 ulation with experimental observations for the Fir wood particle. In figure
531 10, the triangles and the circles represent the experimental results [43]. The
532 solid and small dashed lines represent the moisture content of the particle for
533 the XDEM + OpenFOAM preCICE coupling over time. Drying is described
534 as evaporation due to energy balance in conjunction with a given evapora-
535 tion temperature for the current work. We see that these simulated moisture
536 contents of the wood particle is in good agreement with the experimental
537 results. Whereas for the case with 66% initial moisture, as seen in the figure
538 10, the mean particle temperature goes beyond the water evaporation tem-
539 perature (373K at atmospheric pressure) and remains their from ~ 75 s to
540 ~ 175 s. Because of this we initially see comparatively lower evaporation in
541 the simulated results as compared to the experimental observations. Finally,
542 at around 175 s, the evaporation of water matches the experimental observa-
543 tions, but we see comparatively accelerated evaporation due to higher mean
544 particle temperature. Although the residual moisture mass fraction does not
545 exactly match the simulated results for the 66% initial moisture content, the
546 results are in good agreement as the overall evaporation time and profile are
547 similar.

548 We also compare the XDEM + OpenFOAM preCICE coupling to XDEM
549 + OpenFOAM legacy coupling and we see that the results are almost iden-
550 tical. In Figure 11, the mean temperature of the particle is compared for
551 the different coupling approaches. We can see that the temperature profile
552 for 33% moisture content in the particle is almost identical for the different
553 coupling methods, and the temperature profile for 66% moisture content in
554 the particle is in very good agreement for $2/3^{rd}$ of the simulated time, with
555 minor differences towards the end.

556 In Figure 9, we see various fluid fields at different stages of time. A
 557 negative heat source is seen on the fluid side, which denotes that thermal
 558 energy from the fluid is siphoned off to heat the particle. This is confirmed
 559 by the drop in air temperature downstream of the particle location. As the
 560 particle heats up, the thermal energy is used to evaporate the water in the
 561 wood particle. This water vapor is being injected into the fluid domain. We
 562 confirm the injection of water vapor from the particle into the fluid domain
 563 by observing the transport and diffusion of the water vapor downstream.

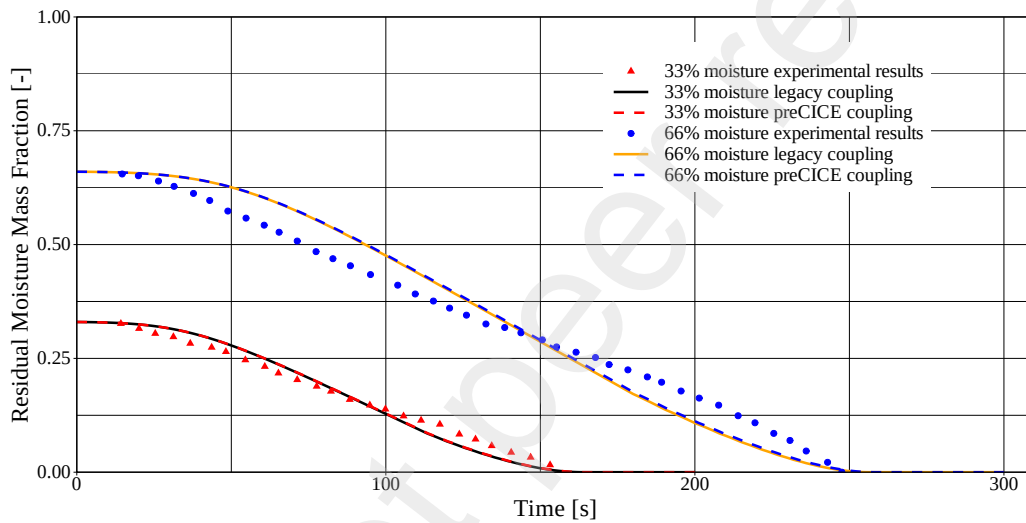


Figure 10: Particle drying simulations compared with experimental drying observations for different initial moisture content in the particle

564 4.3. Heat & Mass Transfer: Pyrolysis of Beechwood particle

565 In the two previous sections, it is thoroughly established that the 2-way
 566 Heat & Mass Transfer coupling between CFD and DEM works well. In the
 567 previous section, where we simulate the drying process, although the particle
 568 loses mass, there are no changes in the chemical composition of the particle.
 569 In the current case, this is exactly what is achieved. The CFD domain is
 570 $0.02 \times 0.02 \times 0.1$ m in size, discretized into 5 cells in the vertical direction
 571 only (uniform 3D Grid $1 \times 1 \times 5$). The air inside the fluid domain is at
 572 1123 K and atmospheric pressure is $1e + 05$ Pa. The air enters from the
 573 bottom of the CFD domain with 0.38 m/s and a temperature of 1123 K,
 574 mimicking the experimental setup by Petek [42]. The particle undergoes

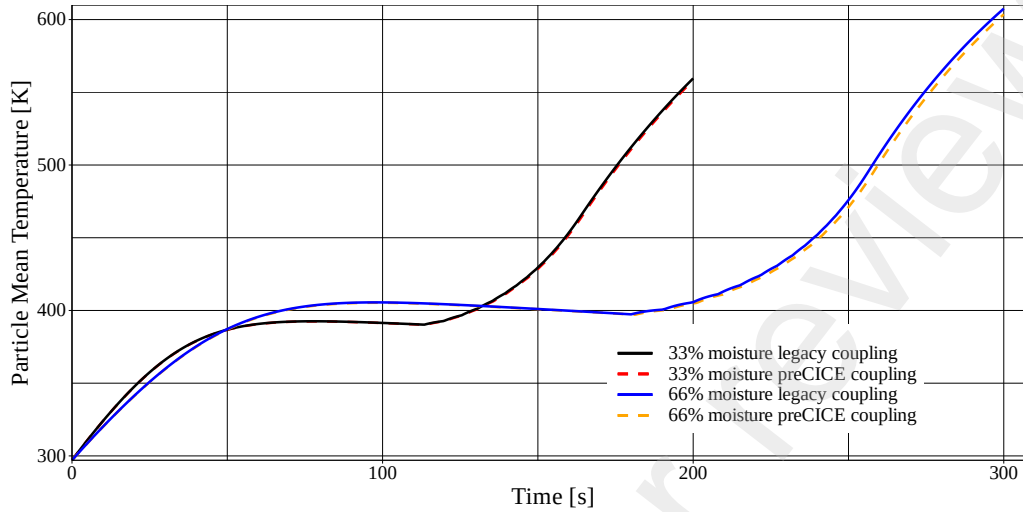


Figure 11: Comparison of the evolution of particle mean temperature over time for different coupling and different initial particle moisture content

575 chemical conversion described in the chemical reactions 4.2, 4.3 and 4.4. The
 576 air exits the CFD domain from the top.

577 4.3.1. Chemical reactions

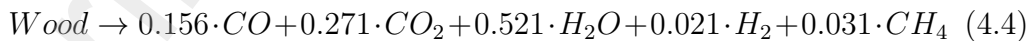
578 In the present work, pyrolysis of a wood particle is simulated and validated
 579 against the experiments performed by Petek [42]. Pyrolysis is described with
 580 three independent reactions 4.2, 4.3 and 4.4 expressing the decomposition of
 581 wood into its main products char, tar and gases [30].



582



583



584 4.3.2. Heat and mass transfer: Pyrolysis case results

585 The simulation results for particle mass loss due to pyrolysis are pre-
 586 sented in figure 12 and validated against the experimental observations. The
 587 predicted particle mass loss is in good agreement with the experimental obser-
 588 vations. The particle surface temperature and centre temperature simulation

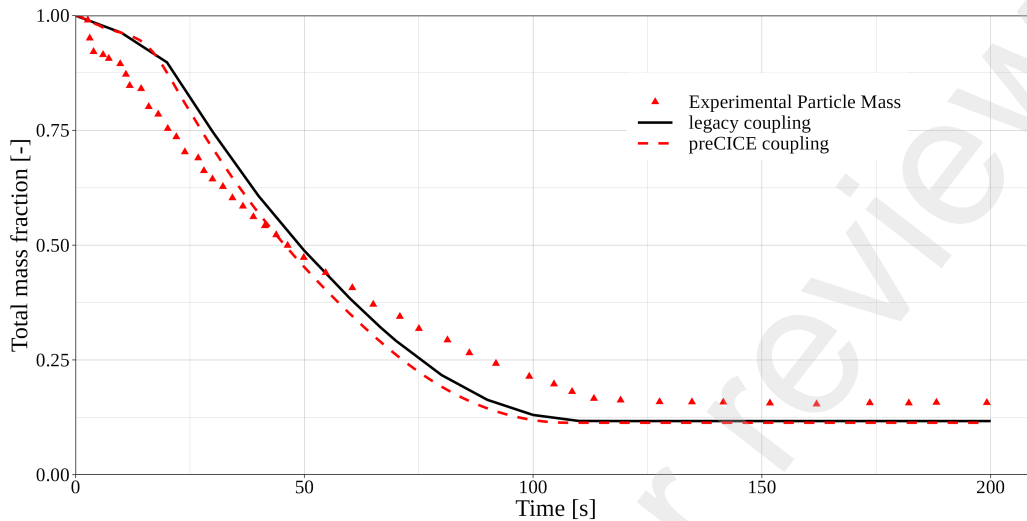


Figure 12: Comparison of the evolution of particle mass overtime for different coupling validated against experimental results of Petek [42]

589 results are compared with the experimental observations in figure 13 and fig-
 590 ure 14 respectively. The predicted particle surface and center temperatures
 591 are in good agreement with the experiments.

592 From figure 13, it can be seen that the particle surface temperature for
 593 numerical simulation rises slowly as compared to the experimental obser-
 594 vations, from 0 s to ~ 80 s. The particle in the experiments experiences
 595 comparatively higher fluid velocities than the numerical simulations, as the
 596 particle in the numerical simulation, as it occupies a comparatively higher
 597 area in the fluid flow. This is because the particle in the numerical simula-
 598 tions is not fully resolved in the fluid domain. The two different simulations
 599 presented use two different strategies. For the XDEM+OpenFOAM legacy
 600 coupling, the CFD-DEM coupled solver is developed specifically for such ap-
 601 plications. In this specialized solver, the particles are represented as porosity
 602 (eq 2.14). Whereas for the XDEM+OpenFOAM preCICE coupling particles
 603 are represented as source term, in this case, momentum source term. As can
 604 be seen from the figure 13, XDEM+OpenFOAM legacy coupling has a compar-
 605 atively higher temperature than XDEM+OpenFOAM preCICE coupling
 606 due to the same reason mentioned above. Due to the presence of the porosity
 607 term in the fluid governing equations, legacy coupling particles experience a
 608 comparatively higher velocity than preCICE coupling.

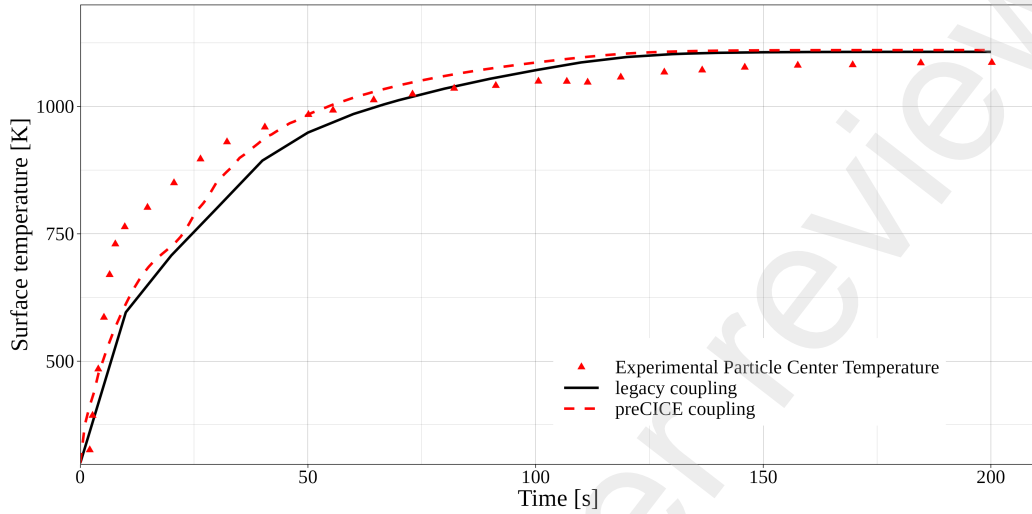


Figure 13: Comparison of the evolution of particle surface temperature over time for different coupling validated against experimental results of Petek [42]

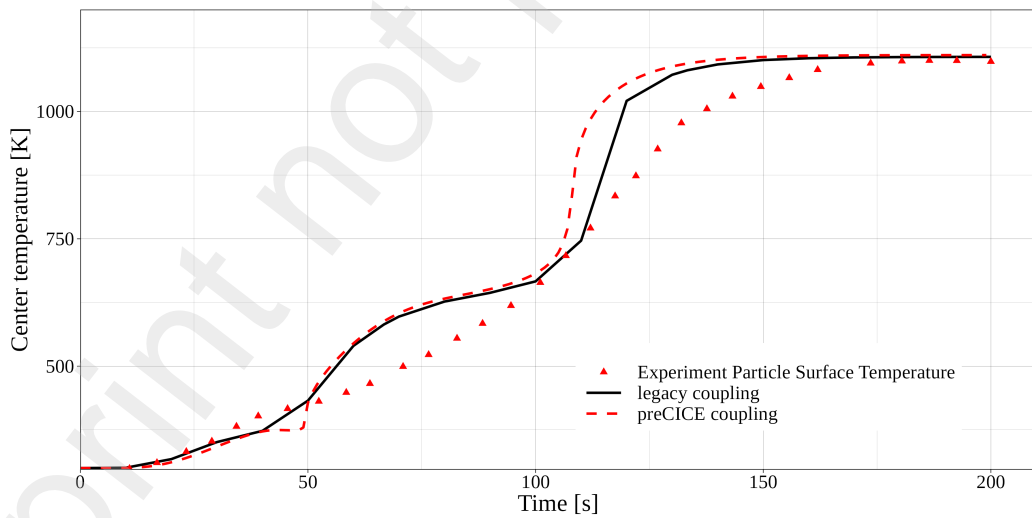


Figure 14: Comparison of the evolution of particle center temperature over time for different coupling validated against experimental results of Petek [42]

609 As the particle surface temperatures are higher for the legacy coupling,
610 we also see a similar phenomenon for the particle center temperature in figure
611 14.

612 *4.4. Heat & Mass Transfer: Drying of packed bed*

613 The experimental data used for validation in the current case was ob-
614 tained by Peters [40] on the test reactor Pantha. The reactor was set up
615 to investigate heating-up, drying, and pyrolysis of packed beds. The ex-
616 periments were carried out on around 2kg of air-dried $10 \times 10 \times 10 \text{ mm}^3$
617 cubical Beech wood containing about 10% moisture by mass. The Beech-
618 wood is placed in a cylindrical bed of 250 mm diameter and 190 mm height.
619 The simulation model is based on the experimental setup, a detailed descrip-
620 tion of the experimental setup is available in reference [40]. The experiments
621 were performed using cubical particles, which are modeled as spheres of equal
622 volume (particle radius = 6.2 mm). Thus the bed is filled with 2667 par-
623 ticles. The drying model described in equation 4.1 is utilized for the drying
624 of the packed bed. In addition to the convective heat transfer, the particles
625 also experience heat transfer through conduction.

626 The CFD simulation mesh can be seen in figure 15 (a) and the DEM
627 simulation domain with particles within it can be seen in the figure 15 (b). An
628 additional height of 80 mm and 60 mm on the top and bottom respectively.

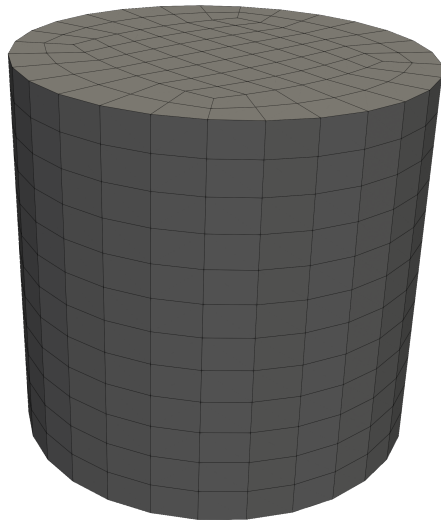
629 The dimensions for the CFD domain are same as those mentioned above
630 for the DEM model. The air enters the CFD domain from the top of the
631 cylindrical, with a temperature of 423 K and a velocity of 0.113 m/s. The
632 air exits the CFD domain at the bottom of the cylinder.

633 In the figure 16, the CFD domain and DEM domain are presented. These
634 meshes are sliced to expose the cell centers that are used to exchange data.

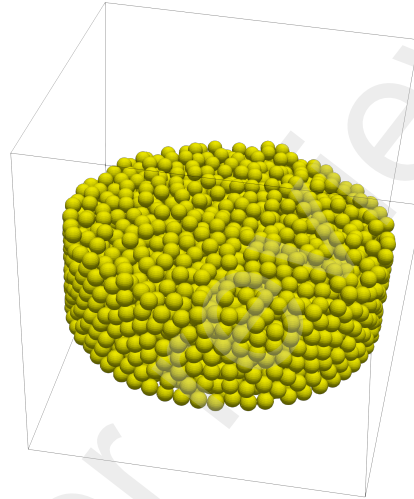
635 *4.4.1. Heat and mass transfer: Drying of packed bed case results*

636 In the experiments [40], as the beech wood particles were heated up and
637 dried, they were measured at certain time intervals to measure the mass
638 loss. The evaporated moisture was also collected in a cold tube and weighed.
639 These measurements were used for the mass balance in the experiments, are
640 are to be used for validation of the current numerical simulation results.

641 The particle weights in the numerical simulation are integrated for each
642 time step, so we have one value for the mass (or mass loss) of the entire
643 packed bed. This is possible as we track the information for all the parti-
644 cles. Finally, in figure 17 we compare the dimensionless moisture loss from

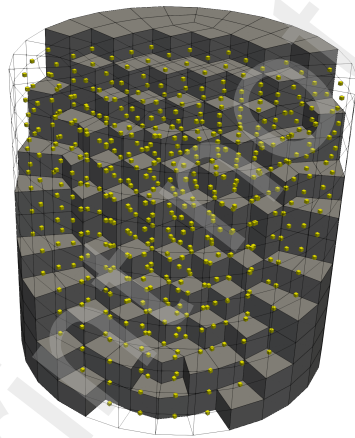


(a) CFD Mesh

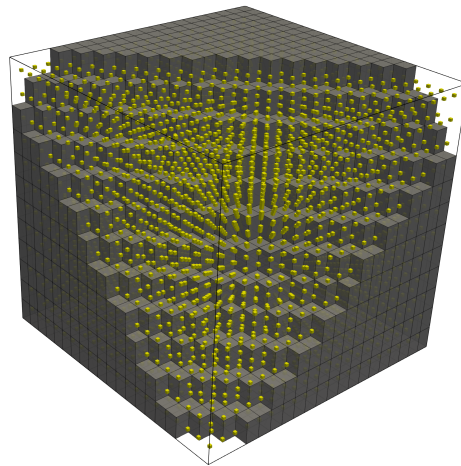


(b) DEM domain with wood particles

Figure 15: Simulation model for drying of packed bed



(a) CFD Mesh



(b) DEM domain

Figure 16: CFD and DEM mesh sliced to show the cells and cell centers used for volume coupling

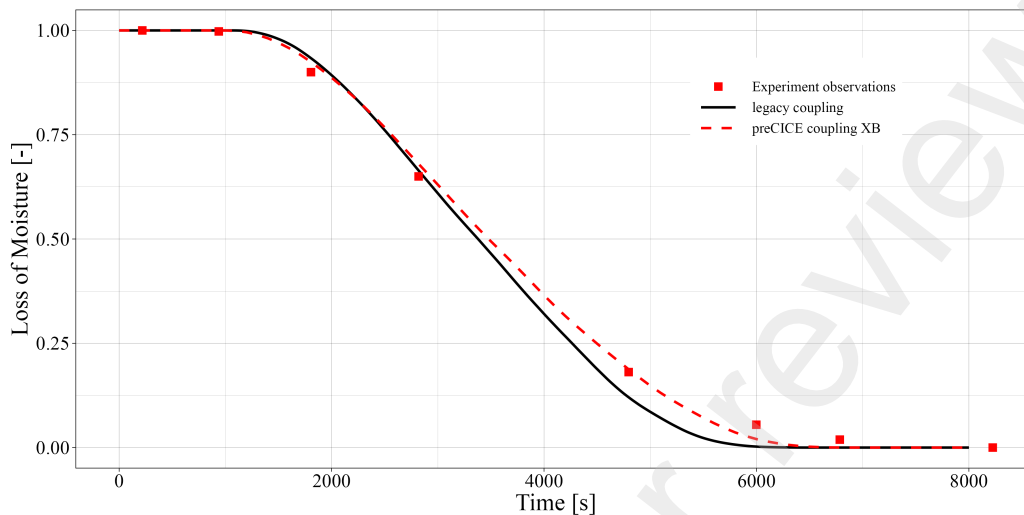


Figure 17: Comparison of the numerical simulation moisture content with the experimental observations

645 the current partitioned coupling strategy to the legacy CFD-DEM coupling
 646 as well as the experimental results. There is good agreement between the
 647 partitioned coupling approach numerical results with the experimental obser-
 648 vations. We also see that the numerical simulation results from the current
 649 work agree with the experimental observations better as compared to the
 650 legacy coupling.

651 Additionally, the evolution of the moisture content (left column) and the
 652 mean temperature (right) of the packed bed is presented in figure 18 and
 653 19 over 8000 s (same as experiment time). It is observed that the particles
 654 at the very edge start heating up more as compared to the particles in the
 655 center. Consequently, we see the drying of these particles first, as the particle
 656 temperature goes over the evaporation temperature. These initial pockets
 657 of concentrated heat in figure 18 are observed due to the **conservative**
 658 mapping constraint. This mapping constraint aggregates the heat source at
 659 the edges into the near wall, thus heating the particles at the edges faster
 660 than the centrally located particles. But as time progresses, and the particles
 661 in the pockets on the edge reach a temperature similar to fluid temperature,
 662 the rest of the bed starts heating up. This behavior is similar to that we see
 663 in the numerical results presented in the literature [26]. As the particles

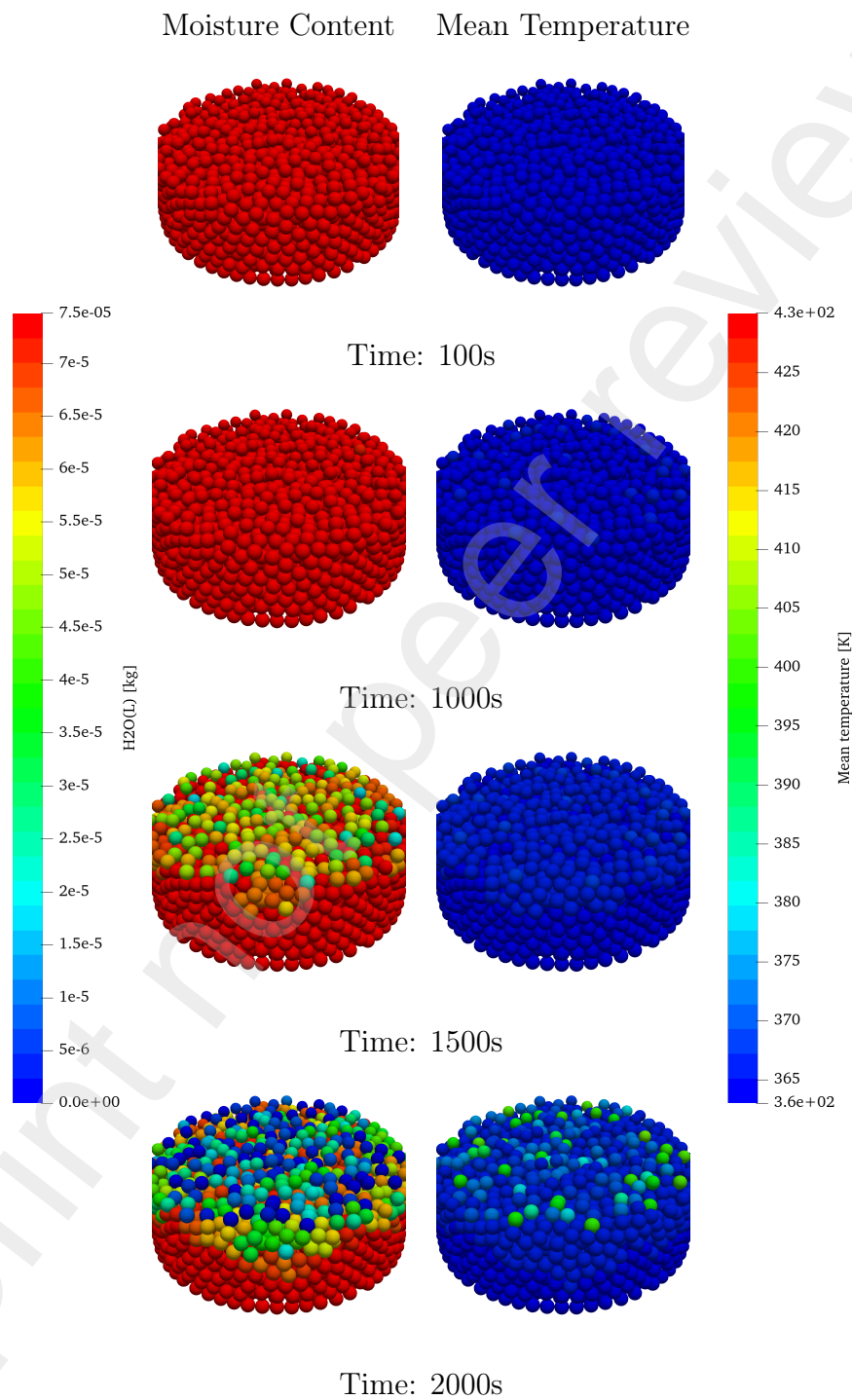


Figure 18: Evolution of particle moisture content and particle mean temperature from 100 s to 2000 s

Moisture Content Mean Temperature

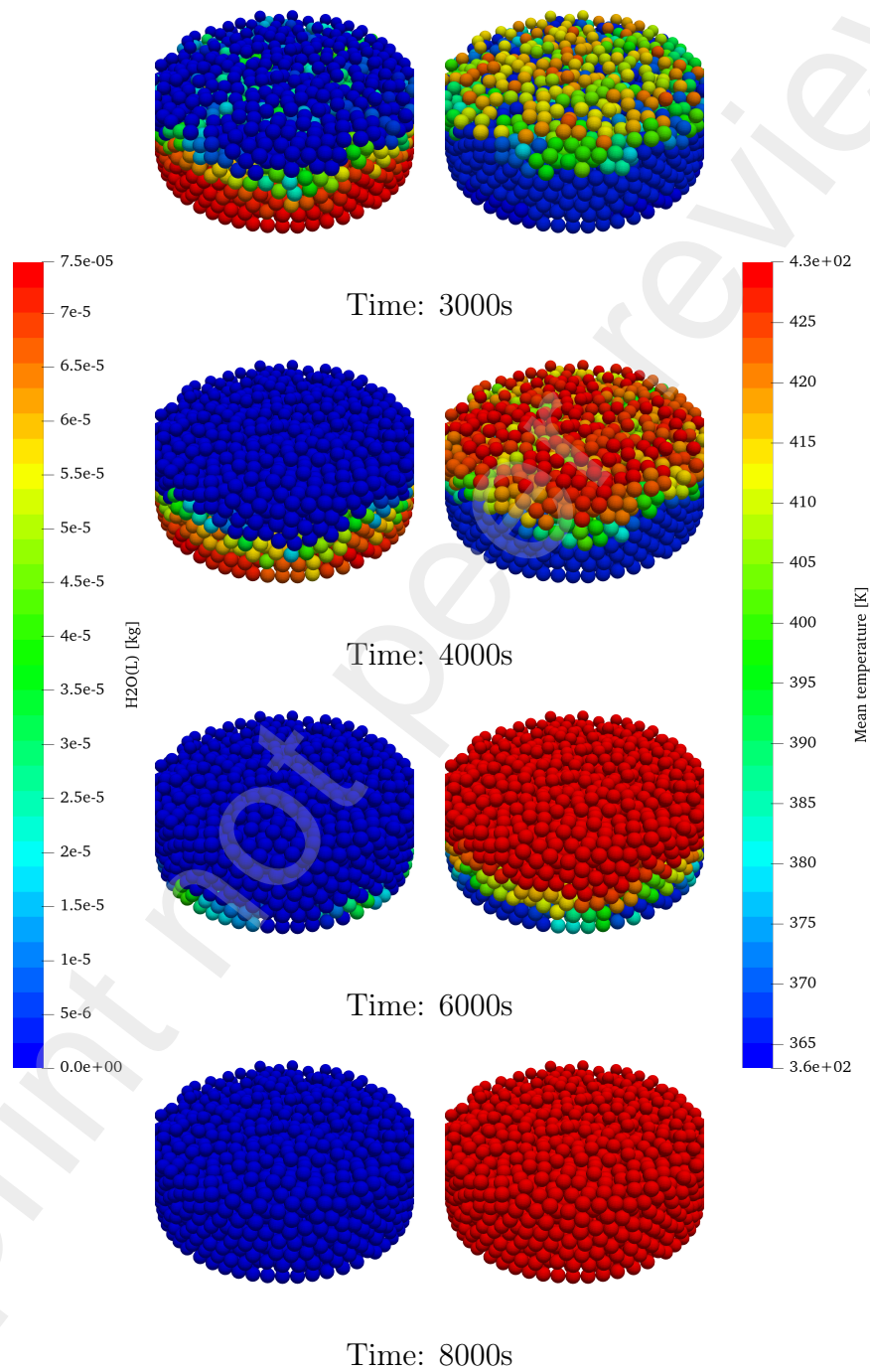


Figure 19: Evolution of particle moisture content and particle mean temperature from 2000 s to 8000 s

664 5. Performance study

665 The previous sections of the article sufficiently demonstrate that the CFD-
666 DEM HMT coupling works well and the results agree with the experimental
667 observations. Although the scope of this study is only the development and
668 validation of the proposed partitioned coupling approach, the authors be-
669 lieve a brief performance study will round out the completeness and inform
670 the reader well when choosing between monolithic and partitioned coupling
671 approaches.

672 The XDEM suite allows two kinds of parallelization: coarse-grain par-
673 allelism with MPI and fine-grain parallelism with OpenMP. On their side,
674 foam-extend and OpenFOAM only support MPI parallelization. The pre-
675 CICE coupling library allows different coupling strategies: `serial` vs `parallel`
676 and `explicit` vs `implicit`, as explained in the section 3.5 and illustrated
677 in the figure 4. The `serial` type of couplings refers to staggered execution
678 of the coupled solvers. On the contrary, the `parallel` type coupling al-
679 lows the coupled solvers to execute simultaneously, allowing functional par-
680 allelism. The `explicit` type of coupling only executes once per coupling
681 time step whereas `implicit` refers to the type of coupling where the coupled
682 solvers execute until convergence. In the present work, `serial-explicit` and
683 `parallel-explicit` coupling schemes are utilized. To summarise, the legacy
684 coupling implementation behaves in the same way as `serial-explicit` pre-
685 CICE coupling, where each coupled solved is executed in a staggered way.

686 Furthermore, it is to be noted that the load balancing is dependent on the
687 case and its configuration. To illustrate this point, we present a performance
688 study for the Pantha case in its original form and the Pantha case where the
689 number of particles and CFD cells are increased. For the performance study,
690 we only simulate 100 s, as this is enough to get performance behavior. This
691 is because we have a constant number of particles throughout the simulation,
692 and they do not move.

693 5.1. Performance evaluation of packed bed

694 In the Pantha simulation case presented in section 4.4.1, the packed bed
695 contains only 2667 particles, thus the use of a single computing node with 128
696 cores for XDEM is sufficient. On the CFD side, the CFD mesh is composed
697 of 1260 cells, hence it does not warrant using parallel execution and it is
698 executed sequentially.

699 In the figure 20, we compare the legacy coupling with the preCICE cou-
700 pling. The blue column in all the plots signifies the execution time needed
701 for XDEM, whereas the green column represents the execution time for the
702 OpenFOAM. As the legacy coupling is implemented as a monolithic solver,
703 the CFD and DEM solvers are executed one after the other. The data ex-
704 change or the coupling is done over a shared memory. Hence, we do not
705 record a separate coupling time, it is included in the XDEM execution time.

706 On the contrary, for the preCICE coupling, the execution time for XDEM,
707 OpenFOAM, and total time are recorded. The red column representing
708 preCICE contains all the time not spent on XDEM and/or OpenFOAM
709 execution. Consequently the red column representing preCICE cost, in-
710 cludes data exchange, data communication through sockets/network, inter-
711 polation of data between meshes, mapping data, and synchronization be-
712 tween the solvers/processes. The time required for the mapping for the
713 `serial-explicit` and `parallel-explicit` are the same. However, it should
714 be noted that the synchronization time, hence the preCICE time also includes
715 the time a solver is waiting for the other solver to finish and proceed.

716 In the figure 20 (a), (b) and (c), the XDEM is executed using 16 OMP
717 threads, whereas in the figure 20 (d), (e) and (f), XDEM is executed using
718 64 OMP threads. It is apparent from these figures that for the case under
719 consideration, preCICE coupling costs are quite significant. For this case,
720 preCICE coupling takes almost twice as much time as required by the legacy
721 coupling. It is to be noted that although the original case set-up remains
722 identical for the two couplings, the legacy coupling uses the CFD mesh for
723 coupling (containing 1260), whereas preCICE considers both CFD mesh and
724 DEM domain discretized as seen in figure 16, which has 4800 cells. This issue
725 is addressed in the next section.

726 In figure 20 (c) and (f), the OpenFOAM and XDEM are executed simul-
727 taneously using the `parallel` coupling, hence they are plotted side-by-side.
728 The light colors for each solver signify the idle/waiting time for the respective
729 solver. It can be seen in the figure 20 (c), XDEM takes more time overall
730 than OpenFOAM, hence the OpenFOAM ends up waiting for XDEM. In con-
731 trast, when we use more computing cores for XDEM, as seen in figure 20 (f),
732 XDEM ends up waiting for OpenFOAM. Due to the nature of the computa-
733 tional load, we see a minor performance gain when using `preCICE-parallel`
734 as opposed to `preCICE-serial`.

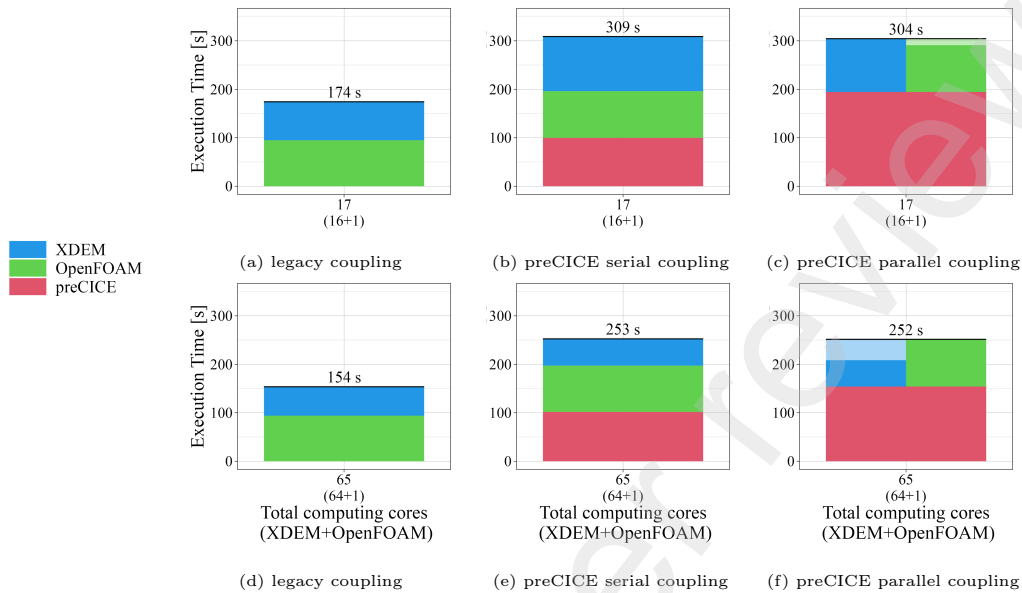


Figure 20: Performance comparison of the Pantha case for the legacy coupling vs the preCICE serial and parallel coupling

735 5.2. Performance evaluation of large packed bed

736 Through the literature [44], it is known that load balancing for the multi-
 737 physics coupled simulations is challenging and dynamic depending on various
 738 factors. In the previous section 5.1, it seems like preCICE is performing
 739 poorly as opposed to the legacy coupling. Hence, we extend the Pantha
 740 drying case to have more particles and CFD cells for a numerical experiment.
 741 This still keeps the underlying physics the same while allowing the study
 742 of performance for a computationally heavier case. The particle sizes are
 743 reduced, and we pack 23,999 particles in the domain. The CFD mesh is
 744 further discretized to have 4,800 cells.

745 Figure 21 shows the performance for the different coupling. The first
 746 notable observation when comparing figure 21 (a) versus (b) and (d) versus
 747 (e), is that with the scaled-up cases, the legacy coupling and preCICE serial
 748 coupling are closely matched. Furthermore, the increased load shows the
 749 disparity between staggered and simultaneous execution. The figure 21 (c)
 750 and (f), shows that the simultaneous execution of the solver gives a substan-
 751 tial performance advantage. As the CFD load is still quite small compared
 752 to the DEM load, it can be seen that the CFD solver spends a lot of time

753 idling, especially in figure 21 (c). As the number of computing resources is
 754 increased, this idling time is reduced in figure 21 (f), but still present.

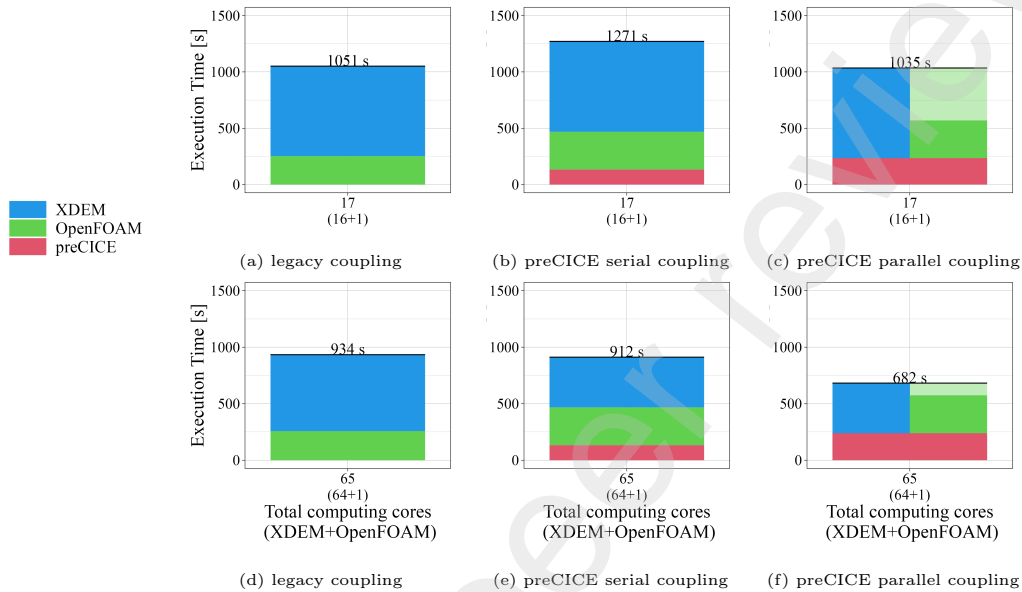


Figure 21: Performance comparison of the large case for the legacy coupling vs the preCICE serial and parallel coupling

755 6. Discussion

756 The results presented in the sections 4, illustrate beyond any doubt that
 757 the partitioned coupling approach presented in this work can capture the
 758 multi-physics behavior. The minor differences in the numerical results be-
 759 tween the legacy coupling and the proposed partitioned coupling are to be
 760 expected due to the different implementations of the OpenFOAM used along
 761 with the different data mapping strategies. In our study, we compared the
 762 results of the partitioned coupling approach to experimental data, as well as
 763 with the numerical simulation results obtained using the monolithic legacy
 764 coupling. The results are in good agreement with the experimental observa-
 765 tions as well as the legacy coupling. This is the case for simple single-particle
 766 cases as well as packed beds.

767 The volumetric coupling employed uses a simple grid on the DEM solver
 768 to exchange data to and from to the CFD solver. Hence, even when employing

769 unresolved CFD-DEM coupling, the CFD mesh size is no longer dependent
770 on and limited by the largest particle. This opens up avenues to explore
771 applications in need of refined CFD mesh smaller than the particles without
772 the need to use resolved CFD-DEM coupling.

773 Along with the verified and validated numerical results, this type of cou-
774 pling gives us the advantage of modularity and flexibility. As the solvers
775 are not intermingled, one solver can be easily swapped out for the other.
776 Furthermore, we would like to point out that now the software language is
777 also no more a restriction. Highly optimized solvers are usually developed in
778 C++, whereas experimental and research implementations are done in Mat-
779 LAB or Python, or similar high-level language. Using the preCICE coupling
780 library these solvers implemented in different programming languages can be
781 coupled without any intrusion. Although it is possible to implement mono-
782 lithic solvers with some functionalities presented in the coupling section 3,
783 the amount of work needed to do so is substantial. This work demonstrates
784 that single-physics numerical solvers can now be coupled with another single-
785 physics software to achieve a coupled multi-physics simulation environment.

786 Additionally, the performance study for a packed bed is presented. With
787 the two cases presented, we show that the performance is dependent on case
788 to case, and the computational resources allocated. It is also shown that the
789 partitioned coupling approach scales well, and performs just as well as the
790 legacy coupling or even better when executing the solvers simultaneously.
791 There are also restrictions on partitioning when using monolithic coupling.
792 Apart from the issue of non-conforming meshes, as presented in section 5
793 there might be cases where the partitioning and resource allocation needs of
794 the two involved solvers are different. When employing partitioned coupling,
795 the parallelization capabilities of the individual solvers can be utilised to the
796 fullest.

797 The monolithic coupling approach solves the set of equations in the same
798 solver. This means that both the solvers involved have to use the same
799 time-step. This time-step is usually dictated by the unstable solver, where
800 reducing the time-step size leads to stability. However, due to the nature
801 of the monolithic coupling, one solver is executed at lower time steps and
802 penalized in computational time due to the unstable solver. When using
803 the partitioned coupling approach, the individual solver time steps are in-
804 dependent of each other, however, they cannot be more than the coupling
805 time-step. This way, the stable solver can retain its time-step, while the
806 unstable solver can utilize a lower time-step for stability. Furthermore, more

807 computational resources can be allocated to the solver with a lower time-step,
808 to balance the computational load.

809 The current approach solves several issues faced in the state of the art
810 for the development of coupled simulation environment. With this work, the
811 authors intend to demonstrate the capabilities and flexibility of using the
812 partitioned coupling approach. We believe that our findings will be useful
813 for researchers and practitioners working in the field of particulate matter
814 processes, particularly those interested in modeling CFD-DEM multi-physics
815 simulations.

816 As a part of future work, this HMT coupling is to be applied to large-scale
817 applications such as blast furnaces and biomass furnaces. A thorough study
818 of the performance and the load balancing challenges using this partitioned
819 coupling approach is to be done [44]. These cases also involve the motion of
820 the particles along with the heat and mass transfer processes.

821 **7. Conclusion**

822 In this work, we present the rapid development of the simulation environ-
823 nment for HMT coupling between CFD and DEM. With the flexibility from
824 preCICE, a user can switch the CFD solver for a preferred one or they can
825 modify OpenFOAM solvers for preferred functionality. In any scenario, this
826 kind of coupling allows the user to test out HMT coupling between particles
827 and fluids. The user may use their own tested, proven, validated CFD or
828 DEM solver to replace the software used in this work to achieve CFD-DEM
829 coupling and simulate desired HMT processes. With the presented results, it
830 is seen that the flexible CFD-DEM black box coupling has similar if not the
831 same results as a specialized CFD-DEM solver. The numerical results are in
832 good agreement with the experimental observations.

833 Heat and mass transfer modules are added to the XDEM and OpenFOAM
834 adapters with relevant required data fields to be exchanged. This enables the
835 rapid development of a multi-physics environment for HMT between particles
836 (DEM) and fluids (CFD). Simple cases are employed to prove that the HMT
837 coupling is working properly. The numerical simulation results are validated
838 against the experimental results, and they are in good agreement. Thus prov-
839 ing HMT coupling using preCICE works i.e. two-way HMT coupling between
840 CFD (OpenFOAM) and DEM (XDEM). This opens up opportunities for the
841 simulation of HMT processes such as drying, gasification, combustion, and
842 pyrolysis.

843 In the HMT validation case, we use chemical species H₂O, O₂, N₂, but
844 our adapter also supports other species such as CH₄, CO₂, CO, H₂, Tar
845 commonly used in the biomass combustion process, or iron making processes.
846 Although these species cover a wide range of applications, one might still need
847 to use many different chemical species. We are working on automating the
848 exchanged species based on fields defined in the preCICE configuration.

849 This work was limited to HMT applications involving gaseous fluid mix-
850 tures. To simulate processes such as melting, and phase change, the Open-
851 FOAM adapter, XDEM adapter, and XDEM have to be updated to handle
852 multiphase Euler-type CFD solvers.

853 The proposed partitioned coupling approach performs just as well or bet-
854 ter than the legacy coupling for large-scale simulations. Thus this type of
855 coupling is scalable and applicable to large-scale applications.

856 In future work, we validate individual processes such as gasification, and
857 combustion similar to the drying and pyrolysis process in the current work.
858 Complex cases such as biomass furnaces and blast furnaces are being inves-
859 tigated using the current implementation and are being validated.

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1009 **Nomenclature**

Scalars

c_p	Specific Heat ($J/(Kg.K)$)
d	Particle diameter (m)
I_i	Moment of inertia ($kg.m^2$)
m	Mass (kg)
m'	Mass source ($kg/m^3.s$)
p	Pressure (Pa)
q'	Heat source (W/m^2)
q''	Heat flux (W/m^2)
r, R	Radius (m)
t	Time (s)
T	Temperature (K)
T_{final}	Length of simulation (s)

1010

Greek symbols

α	Heat transfer coefficient ($W/(m.K)$)
β	Momentum exchange ($kg/(m^3.s)$)
∂	Differential operator (-)
ϵ	Volume Fraction/Porosity (-)
μ	Kinematic viscosity ($Pa.s$)
∇	Nabla operator (-)
ρ	Density (kg/m^3)

First order tensor (vectors)

\vec{g}	Acceleration due to gravity (m/s)
\vec{F}^c	Contact Forces (N)
\vec{F}^g	Gravitational Force (N)
\vec{F}^{ext}	External Forces (N)
\vec{F}_B	Buoyancy Force (N)
\vec{F}_D	Drag Force (N)
$\vec{M}_{i,j}$	torque generated by inter-particle forces ($N.m$)
\vec{S}	Momentum source due particles
\vec{v}_f	Fluid velocity field
\vec{X}_i	Positional vector (m)
$\vec{\omega}$	Rotational velocity (rad/s)

Subscripts

c	Cell
$cond$	Conduction
eff	Effective values
f	Fluid
i, j	Particle
n	Normal direction
p, P	Particle
rad	Radiation
t	Tangential direction