

# CFD-based erosion and corrosion modeling in pipelines using a High-order discontinuous Galerkin multiphase solver

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## Abstract

We present a Computational Fluid Dynamics (CFD)-based methodology for the modeling of erosion and corrosion in hydrocarbon pipes. The novelty of this work is the use of a high-order Discontinuous Galerkin Spectral Element Method (DGSEM) approximation of the incompressible Navier-Stokes/Cahn-Hilliard model for the CFD simulation. This technique permits a very detailed three dimensional representation of the flow regime, phases distribution and contact surfaces that conform the pipe, which results in accurate computations of erosion and corrosion rates and distribution over the pipeline surface.

The developed methodology is validated with experiments relevant for oil and gas industry. In particular, we simulate the erosion in a one-phase ascending pipe with two elbows and the corrosion in a two-phase pipe under several flow regimes.

*Keywords:* Computational fluid dynamics, Phase-field, High-order Discontinuous Galerkin, Erosion, Corrosion, Pipelines, Multiphase flow.

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## 1. Introduction

The transportation of hydrocarbons is characterized by the presence of multiphase flow, usually containing a mixture of liquid crude oil, natural gas, water and sand particles. This transport may result in the wear of the pipe due to phenomena such as erosion and corrosion. On the one hand, erosion usually occurs as a result of the transport of the fluids with entrained solid particles. The erosion is caused when these particles, dragged by the fluid, hit violently the wall of the pipe, begin able to remove part of its material. On the other hand, multiphase flows usually contain acidic chemical components, such as  $\text{CO}_2$  or  $\text{H}_2\text{S}$ , dissolved in the aqueous phase. Corrosion appears if those components get in contact with the walls at certain pressure, temperature and shearing conditions. Both erosion and corrosion can be responsible for failure of the equipment, transport pipelines and production tubing. The prediction of these phenomena permits to estimate the operating life and to identify the locations where severe wear is likely to occur. For example, elbows and plugged tees, which are typically used in piping systems to redirect fluids, are particularly vulnerable to erosion when sand particles are present.

Erosion is a highly complicated phenomenon with a wide range of factors contributing to its severity (e.g. fluid flow and particles rates, properties of the fluid, properties of the particles, wall material or particle impingement angle). A variety of erosion models have been developed to predict this phenomenon. Most methods are based on a limited amount of experimental data, and therefore are only applicable to specific conditions. The use of general guidelines (e.g. American Petroleum Institute Recommended Practice 14E (API RP 14E) [1]) conduces to too conservative designs for most of the situations. Although the accuracy of these general guidelines can be improved (see for example [2, 3, 4, 5]), it is required to have detailed data on the velocity of the particles and their impact angle to really refine the prediction of the erosion rate and its location. Following this idea, McLaury [6] proposed a generalized erosion prediction procedure that involves flow simulation, particle tracking, and erosion modelling, which is now generally known as Computational Fluid Dynamics (CFD)–based erosion prediction [7, 8, 9]. Erosion models are key to get accurate results using CFD–based erosion prediction. Finnie’s erosion model [10] has been employed by many authors in

35 the past to compute the erosion ratio, but it is necessary to adjust correctly  
36 the empirical parameters to obtain accurate results [11, 12, 13]. In contrast,  
37 there are more advanced models (see for example Oka’s model [14]) that can  
38 provide precise results avoiding many empirical parameters [15, 16, 17]. The  
39 use of CFD–based erosion prediction techniques on industrial systems allow  
40 to minimize erosion damages and to optimize the pipe geometry and the  
41 flow conditions. For example, see [18, 19] where an optimization of the pipe  
42 geometry was conducted to reduce the erosive wear.

43 Corrosion in steel pipes mainly depends on the presence of an aqueous  
44 phase in contact with the pipe. Most of the published models for pipeline  
45 internal corrosion apply for sweet corrosion only [20, 21]. However, the water  
46 and the oil phases can be either mixed (e.g. emulsions) or separated (e.g.,  
47 stratified, slug or annular regimes). The fluid flow regime, which depends on  
48 the properties of the fluids, water cut, or the pipe geometry and inclination,  
49 will determine if the pipe is wetted by either oil or the corroding aqueous  
50 phase [22]. Besides, the pipe corrosion can be modeled by a combined effect  
51 of electro–chemical reactions, mass transfer and wall shear stress. As a result,  
52 sweet corrosion of steels should include the influence of fluid mechanics [23,  
53 24, 25, 26].

54 In this work, we propose a methodology to predict the erosion and corro-  
55 sion in pipes using a multiphase CFD technique. We solve the flow field with  
56 the method described by Manzanero et al. [27]. This method uses a three-  
57 phase phase–field approach coupled with the incompressible Navier–Stokes  
58 equations to describe the flow in pipes with (up to) three immiscible phases.  
59 Compared to other interface capturing methods such as Volume of Fluid  
60 [28] or Level Set [29], phase field methods (also known as diffuse interface  
61 methods) [30, 31, 32] provide a useful alternative that overcomes the most  
62 commonly found problems as mass conservation or the accurate computation  
63 of the surface tension. In diffuse interface methods, a phase–field function  
64 that describes the N–phase system is defined. The sharp fluid interface is  
65 replaced by a smooth transition layer that connects the two immiscible fluids.  
66 The evolution of the phase–field function in [27] is modeled by means of the  
67 convective Cahn–Hilliard equation [33], in particular with the model of Boyer  
68 et al. [34].

69 The three-phase model in [27] is numerically approximated in space with  
70 a high-order Discontinuous Galerkin Spectral Element Method (DGSEM)  
71 [35] that uses the symmetric interior penalty method [36, 37, 38, 39, 40]. The  
72 DGSEM has been used in the past to discretize multiphase (two phase) flows  
73 [41, 42, 43, 44, 45], and it is popular because of its arbitrary order of accuracy  
74 [46, 35], its low dissipative and dispersive errors [47, 48, 49, 50], the ability to  
75 represent arbitrarily three-dimensional complex geometries through the use  
76 of unstructured meshes with curvilinear elements [51], efficient mesh adap-  
77 tation techniques [52, 53, 54], and the design of provably stable schemes  
78 [55, 56, 57, 58, 59, 60, 44].

79 Previous studies of the erosion and corrosion have used CFD techniques  
80 to predict the flow behaviour. However, they use low order methods imple-  
81 mented in commercial solvers such as ANSYS-Fluent, or CFX (e.g. [7, 8, 9]  
82 for erosion prediction or [61, 26] for corrosion prediction). To the best of  
83 the authors' knowledge, this is the first work that uses a phase-field model  
84 discretized with a high-order CFD method for the prediction of the erosion  
85 and corrosion. The phase-field model has one advantage compared to their  
86 counterparts, as it permits to model phase separation, even in the absence  
87 of gravity effects. This is important as no special inlet boundary condition  
88 must be imposed, the phases may enter in the pipeline mixed, and the sep-  
89 aration will occur as the flow develops in the simulation domain. From the  
90 advantages of high order methods previously enumerated, we highlight the  
91 enhanced representation of the geometries. The latter results is an improved  
92 prediction of the particles impact angles and the wall shear stress, with a  
93 subsequent more accurate estimation of the erosion and corrosion. It should  
94 be noticed that the CFD method chosen [27] permits to solve up to three  
95 phases flow simulations. This is interesting for hydrocarbon transport pipes,  
96 where water, liquid oil and natural gas can be found. However in this work  
97 we restrict ourselves to two-phase simulations.

98 The content of this text is organized as follows. First, Section 2 introduces  
99 the numerical methodology to predict the flow regime, and based on the flow  
100 variables obtained, compute corrosion and erosion. Then, Section 3 includes  
101 numerical experiments to validate our erosion and corrosion toolchain com-  
102 putation. Finally, Section 4 summarizes the main conclusions of the work.

## 103 2. Methodology

### 104 2.1. Computation of the fluid flow field. Flow simulation

105 In this section we describe the equations that govern the evolution of the  
106 flow variables, and their numerical approximation. The numerical method  
107 used is fully described in previous works: see [62] for the discretization of  
108 the three component Cahn–Hilliard equation and [27] for the discretization  
109 of the three phase flow model. The methodology is implemented in a flexible  
110 solver that handles general unstructured high order meshes. In this work,  
111 for simplicity we only summarize those schemes, referring the reader to the  
112 previous references for details.

113

We use the incompressible Navier–Stokes/Cahn–Hilliard model of [27]. Although this model can solve up to three–phase flows, we restrict ourselves to its two–phase variant. The model is a diffuse interface phase field method, which introduces a scalar field that represents the volume fraction, or concentration, of each of the phases in a continuous fashion,  $c(\mathbf{x}, t)$ . The concentration typically ranges from 1 (Phase 1) to 0 (Phase 2). Across the interface, the thermodynamic variables (e.g. density and viscosity) face a smooth transition from their values in one phase to the other (assumed constant values),

$$\rho(c) = \rho_1 c + \rho_2 (1 - c). \quad (1)$$

The evolution of the concentration follows the convective Cahn–Hilliard equation

$$c_t + \vec{\nabla} \cdot (c\mathbf{u}) = M_0 \vec{\nabla} \cdot (\vec{\nabla} \mu), \quad (2)$$

where the chemical potential,

$$\mu = \frac{24\sigma}{\varepsilon} (c(1 - c)^2 - c^2(1 - c)) - \frac{3}{2}\sigma\varepsilon \vec{\nabla}^2 c, \quad (3)$$

is driven by a double–well chemical free–energy. In (3),  $\sigma$  represents the interface tension coefficient between the phases. The size of the interface is effectively controlled by the *interface width*,  $\varepsilon$ , a parameter of the model whose value is limited in practice by the mesh resolution. The Cahn–Hilliard equation includes antidiffusive and diffusive chemical effects. On the one

hand, the antidiffusion is introduced by the chemical free-energy, which favors phase separation and it is one of the features that makes this model attractive. This enables the possibility to initialize the solutions from a homogeneous mixture of the phases, and then let the phases to grow, evolve, and coalesce naturally. On the other hand, the diffusion is known as the interfacial energy, and it is responsible for the regularization of the antidiffusive effects. The strength of one term over the other depends on the interface width  $\varepsilon$  parameter,

$$\varepsilon \propto \frac{\text{diffusive}}{\text{antidiffusive}}, \quad (4)$$

114 such that increases of  $\varepsilon$  stabilize the solution at the expense of a wider inter-  
115 face.

A final remark on the Cahn–Hilliard equation, is that the concentration is driven by two effects: convection and the chemical diffusion. If the convection dominates over the chemical diffusion, the flow is not able to fully perform phase separation and the phases tend to remain dispersed. On the contrary, if the chemical diffusion dominates, the strength of the interface is too high and the velocity field cannot warp the bubbles formed. Therefore, we seek for a delicate balance between these two effects, controlled with a constant parameter called *mobility*,  $M_0$ ,

$$M_0 \propto \frac{\text{chemical diffusion}}{\text{convection}}. \quad (5)$$

116 For more details on the proper selection of the mobility parameter, see [63].

117

Unlike conventional methods which rely in the Reynolds–Averaged Navier–Stokes (RANS) equations to compute the velocity field, an implicit Large Eddy Simulation (LES) approach [64] is adopted in this work, where the numerical viscosity of the scheme acts as a subgrid filter. The LES methodology results in a more accurate treatment of turbulent effects, with the disadvantage of requiring more spatial resolution. For a comparison of RANS and LES approaches in one phase pipe simulations see [65]. The velocity field  $\mathbf{u}(\mathbf{x}, t)$  is obtained from the momentum equation, which incorporates a volume approximation of the capillary forces

$$(\rho\mathbf{u})_t + \vec{\nabla} \cdot (\rho\mathbf{u}\mathbf{u}) = -\vec{\nabla}p + \mu\vec{\nabla}c + \vec{\nabla} \cdot \left( \eta \left( \vec{\nabla}\mathbf{u} + \vec{\nabla}\mathbf{u}^T \right) \right) + \rho\mathbf{g}. \quad (6)$$

118 In (6),  $\eta$  is the viscosity, computed from the (constant) equilibrium phases  
 119 viscosities  $\eta_{1,2}$  in a similar fashion to the density (see (1)). The term  $\mu \vec{\nabla} c$   
 120 is the phase field approximation of the capillary pressure [66], and  $\mathbf{g}$  is the  
 121 gravity acceleration.

122

Finally, the pressure is computed from an artificial compressibility model [67, 68, 69] as:

$$p_t + \rho_0 c_0^2 \vec{\nabla} \cdot \mathbf{u} = 0, \quad (7)$$

123 with  $\rho_0 = \max(\rho_1, \rho_2)$  and  $c_0$  the artificial compressibility sound speed.

124

125 The equations are approximated with a high-order Discontinuous Galerkin  
 126 Spectral Element Method (DGSEM) [35]. This method is convenient since it  
 127 provides arbitrarily high order accurate schemes while maintaining geomet-  
 128 rical flexibility through the use of unstructured meshes. The conservation  
 129 laws are written in a conservative fashion similar to finite volume schemes.  
 130 However, inside every element the solution is approximated by a polynomial  
 131 function of degree  $N$ , which allows the solution to spatially vary within the  
 132 element in the DG method. One of the main advantages of high-order meth-  
 133 ods is the spectral convergence, which results in exponential convergence for  
 134 smooth solutions. This results in less degrees of freedom for the same accu-  
 135 racy, when compared to traditional low order methods [46, 35]. Additionally  
 136 the operation count of the DGSEM differential operators scales linearly with  
 137 the polynomial order [70], resulting in a linear growth of run times with the  
 138 polynomial order for explicit computations, such as the ones conducted in  
 139 this work. The increase of the polynomial order presents one disadvantage  
 140 though, which is a steeper limitation of the explicit time step compared to a  
 141 conventional mesh refinement (the time step for explicit computations scales  
 142 with  $h/N^2$ , where  $h$  is the size of the element and  $N$  the polynomial degree  
 143 used in the computation see [46]). Of course, this limitation could be by-  
 144 passed by using implicit time integrators with a dual time step strategy, which  
 145 is not considered in this work. Besides, high-order schemes present additional  
 146 advantages such as their low numerical dissipation, a higher-order approxi-  
 147 mation of the geometry (e.g. retaining the curvature see Appendix A), their  
 148 compact stencils, and all without comprising their robustness [71, 72, 59, 44],

149 as they can be constructed provably stable. The thorough details of the  
 150 incompressible Navier–Stokes/Cahn–Hilliard system approximation can be  
 151 found in [27].

## 152 2.2. Erosion computation

153 To predict the erosion, we firstly inject a number of particles in the flow,  
 154 whose trajectories are then calculated. Secondly, the erosion produced by  
 155 each individual particle impact onto the pipe wall is computed. Finally,  
 156 the erosion rate and particle impact statistical parameters are interpolated  
 157 onto a fine surface mesh. The tracking of the particles is performed over  
 158 an averaged flux using a one–way coupling, i.e. it assumes no interaction  
 159 between particles and no influence of the particles in the fluid flow. For  
 160 more advanced approaches, where inter-particle collisions are considered see  
 161 [73, 74]. However, this one–way coupling has proven accurate enough for  
 162 particle laden flows where the mass fraction of the disperse phase is low.

### 163 2.2.1. Particle tracking

164 Particles are assumed spherical and with no interactions among them.  
 165 Under these assumptions, the dynamics of the particles is described by the  
 166 momentum equation:

$$\frac{d\mathbf{u}_p}{dt} = \mathbf{g} \frac{\rho_p - \rho}{\rho_p} - \frac{1}{2} \frac{\rho C_d A_{p,c}}{m_p} (\mathbf{u}_p - \mathbf{u}) |\mathbf{u}_p - \mathbf{u}|, \quad (8)$$

167 where  $\rho_p$  is the particle density,  $\mathbf{u}_p$  is the particle velocity,  $\mathbf{g}$  is the gravity,  
 168  $C_d$  is the drag coefficient (a function of the Reynolds number,  $Re$ ),  $A_{p,c}$  the  
 169 particle cross–sectional area and  $m_p$  is the particle mass. Fluid velocity,  $\mathbf{u}$ ,  
 170 and density,  $\rho$ , are interpolated to the position of the particle from the fluid  
 171 field. Additionally, the momentum equation takes into account buoyancy  
 172 effects.

As the velocity of the particles is known, the particle position,  $\mathbf{x}_p$ , can be  
 determined from the equation,  $\frac{d\mathbf{x}_p}{dt} = \mathbf{u}_p$ . The drag coefficient is computed  
 using the Brown and Lawler formula for spherical particles [75],

$$C_d = \frac{24}{Re_D} (1 + 0.15 Re_D^{0.681}) + \frac{0.407}{1 + \frac{8710}{Re_D}}, \quad (9)$$



valid for the subcritical region  $Re_D < 10^5$ , with

$$Re_D = \frac{\rho |\mathbf{u}_p - \mathbf{u}| 2r_p}{\eta}, \quad (10)$$

173 where  $\eta$  is the dynamic viscosity of the mixture interpolated from the fluid  
174 field and  $r_p$  the particle radius.

175 Finally, for the bounce of the particles, the Sommerfeld virtual wall model  
176 with random orientation [76] is used to model the pipe wall roughness. We  
177 use a deterministic approach for the normal coefficient of restitution ( $e_n$ ) and  
178 tangential coefficient of restitution ( $e_t$ ). The impact velocity is split into the  
179 normal  $\mathbf{u}_{1,n}$  and tangential  $\mathbf{u}_{1,t}$  components with respect to the virtual wall.  
180 Then the components of the particle velocity after the collision  $\mathbf{u}_{2,n}$  and  $\mathbf{u}_{2,t}$   
181 are calculated with the expression

$$\begin{aligned} \mathbf{u}_{2,n} &= -e_n \mathbf{u}_{1,n} \\ \mathbf{u}_{2,t} &= e_t \mathbf{u}_{1,t}. \end{aligned} \quad (11)$$

For the coefficients of restitution a simple two parameters model is used.  
The coefficient,  $e$ , is 1 for zero impact angle and decreases linearly with the  
impact angle  $\theta$  up to a limit angle  $\theta_e$ . After that, it remains constant with  
value  $e_h$  for impact angles higher than  $\theta_e$ ,

$$e = \begin{cases} 1 - \frac{(1-e_h)\theta}{\theta_e} & \theta \leq \theta_e \\ e_h & \theta > \theta_e. \end{cases} \quad (12)$$

182 The parameters for the normal ( $\theta_{e,n}, e_{h,n}$ ), and tangential ( $\theta_{e,t}, e_{h,t}$ ) com-  
183 ponents are adjusted to fit experimental data. The numerical values of the  
184 parameters used in this work are specified in Sec. 3.1.1.

### 185 2.2.2. Computation of erosion and penetration rates

The erosion of a pipeline is estimated by calculating the erosion caused by  
a sufficient number of individual particles to be statistically representative.  
According to the Finnie's model of 9 parameters [11, 12, 13], the *erosion*  
*ratio*,  $ER$  (kilograms of eroded material per kilogram of impacting material,  
( $\text{kg kg}^{-1}$ )) is defined as

$$ER = k |\mathbf{u}_1|^n f(\theta), \quad (13)$$

where  $\mathbf{u}_1$  is the particle impact velocity,  $f(\theta)$  is a function of the impact angle,  $\theta$ ,

$$f(\theta) = \begin{cases} a\theta^2 + b\theta, & \theta \leq \theta_0 \\ x \cos^2 \theta \sin(\omega\theta) + y \sin^2 \theta + z, & \theta > \theta_0 \end{cases}, \quad (14)$$

186 and  $k, n, a, b, x, w, y, z, \theta_0$  are empirical parameters that depend on the eroded  
187 material. In this work, the aluminium material values are used for the model  
188 constants (more details are given in Section 3.1.1).

We define two parameters of interest in erosion computations: the *penetration rate* ( $\mu\text{m kg}^{-1}$  or  $\text{mil lb}^{-1}$ ) and the erosion per unit of time and surface or *erosion rate* ( $\text{kg m}^{-2} \text{s}^{-1}$ ). These parameters are defined for a certain computational cell  $i$  by the expressions:

$$\begin{aligned} \text{penetration rate}_i &= \frac{\sum_1^{n_p} ER_j m_j}{M_t a_i \rho}, \\ \text{erosion rate}_i &= \frac{\dot{m} \sum_1^{n_p} ER_j m_j}{M_t a_i}, \end{aligned} \quad (15)$$

189 where  $ER_j$  and  $m_j$  are the erosion ratio and the particle mass for the impact  
190  $j$ ,  $n_p$  is the number of impacts in cell  $i$ ,  $M_t$  is the total mass of particles  
191 injected in the simulation,  $\dot{m}$  is the particles mass flow to be simulated,  $a_i$  is  
192 the cell area and  $\rho$  the density of the pipe material.

### 193 2.3. Computation of corrosion

The *corrosion rate*,  $CR$  ( $\text{mm year}^{-1}$ ), is defined following the Kanwar model [77] as:

$$CR = k P_{\text{CO}_2}^c \tau^b, \quad (16)$$

194 where  $\tau$  is the shear stress on the wall computed from the flow variables,  
195  $P_{\text{CO}_2}$  (MPa) is the partial pressure of  $\text{CO}_2$  and  $k, b$  and  $c$  are the non-dimensional  
196 parameters of the model. The parameters shown in Table 1 are valid for a  
197 carbon steel pipe. These values have been used in this work.

Corrosion is only active if there is free water phase in contact with the wall. Our CFD-based corrosion modeling approach predicts the amount of water ( $c_w$ ) in contact with the wall at each time step. The parameter  $c_w$  ( $c_w = c$  in (2)) varies continuously from 0 (oil) to 1 (water), and represents the volume of water in a cell divided by the volume of the cell. We propose to

Table 1: Parameters of the Kanwar model [78]

$P_{\text{CO}_2}$ , [MPa]	0.27
$k$	15.5
$b$	0.1
$c$	0.83

establish a threshold for the activation of the corrosion process. In particular, we define the following step function,

$$CR_{step} = \begin{cases} 0 & c_w \leq 0.25 \\ CR_k & c_w > 0.25 \end{cases}, \quad (17)$$

198 where  $CR_k$  is the corrosion obtained through expression (16). The value  
 199 of 0.25 for the concentration, or volumetric fraction, of water,  $c_w$ , was se-  
 200 lected because it showed good performance in preliminary tests, and remains  
 201 unchanged for all the test cases.

### 202 3. Numerical experiments

203 In this section we focus on the validation of the methodology using rel-  
 204 evant test cases for the oil and gas industry. In particular, we simulate the  
 205 erosion in a one-phase ascending pipe with two elbows and the corrosion in  
 206 a two-phase pipe under different flowing conditions.

#### 207 3.1. Erosion

208 The methodology to predict the erosion described in Section 2 is validated  
 209 with the test case presented by Chen et al. [11]. This validation consists of  
 210 three main parts: flow simulation (Section 2.1), particle tracking and erosion  
 211 calculation (Section 2.2).

##### 212 3.1.1. Setup

213 The validation case consists of a 1 inch diameter pipe ( $2.54 \cdot 10^{-2}$  m) with  
 214 an initial horizontal section of 2.13 m, an ascending vertical section of 1.22 m,  
 215 and a horizontal outlet section of 0.91 m. The sections are joined by elbows  
 216 of 90 deg and curvature ratio ( $r/D = 1.5$ ). The material used in the pipe is

217 pure aluminum. A scheme of the case geometry is shown in Figure 1. A flow  
 218 of air carrying sand particles is established between the inlet and the outlet  
 219 of the pipe. The simulation parameters used are shown in Table 2. They  
 220 replicate the experiment detailed in [11] with inflow velocity of 150 ft/s and  
 particle flow rate of 40 lb per day.

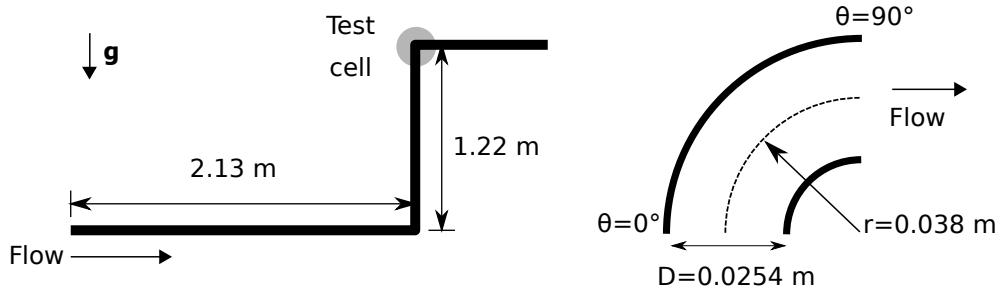


Figure 1: Scheme of the pipe geometry and location of the test cell [11]

221

Table 2: CFD simulation parameters

Fluid	Air
Velocity inlet, [ $\text{m s}^{-1}$ ]	45.72
Density, [ $\text{kg m}^{-3}$ ]	1.225
Kinematic viscosity, [ $\text{Pa}\cdot\text{s}$ ]	$1.85 \cdot 10^{-5}$
Particles diameter, [m]	$1.5 \cdot 10^{-4}$
Particles density, [ $\text{kg m}^{-3}$ ]	2650
Particle mass fraction, [-]	$7.44 \cdot 10^{-3}$
Pipe material	Pure aluminium
Density pipe material, [ $\text{kg m}^{-3}$ ]	2700
Gravity, [ $\text{m s}^{-2}$ ]	9.8

222 For the fluid flow field computation, a CFD simulation has been carried  
 223 out on a mesh of 3680 elements with polynomial order 4 (460000 degrees of  
 224 freedom). A convergence analysis, not included here, shows a good compro-  
 225 mise between accuracy and computational cost for this mesh. The particle  
 226 volumetric fraction in the experiment is  $3.4 \cdot 10^{-6}$ , such that the interaction  
 227 between particles and fluid flow turbulence is low [79] and the effect of the

228 particles on the mean flow is negligible. In this scenario, the use of the one-  
 229 way fluid–particle coupling model is justified. The particle tracking for the  
 230 erosion calculation has been performed with 100000 particles. According to  
 231 numerical experiments in [11] the variation in erosion prediction is negligible  
 232 for a number of particles above 50000. The particle tracking calculation is  
 233 computed on a steady solution. A time step of  $5 \cdot 10^{-3}$  s is used for the par-  
 234 ticle tracking integration. This time step permits an accurate computation  
 235 of the rebounds of the particles.

236 The parameters of the coefficient of restitution Eq. (12), shown in Fig-  
 237 ure 2, have been estimated using Reagle’s [80] and Sommerfeld’s experimental  
 238 data [76]. More advanced parameters such as the coefficient of friction [74]  
 239 were not considered for this model.

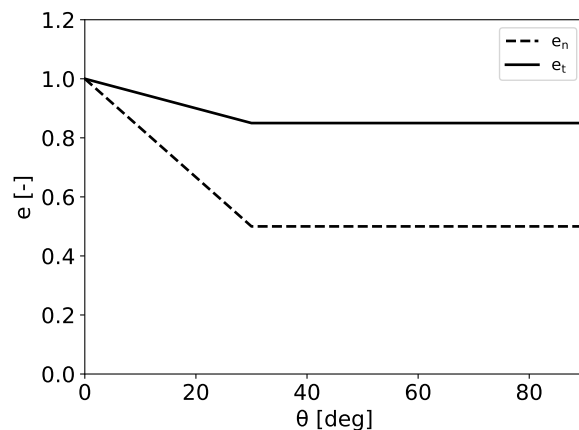


Figure 2: Normal and tangential coefficient of restitution in function of the impact angle used in the simulations [76]

240 Finally, the parameters of the erosion model used in the simulation are  
 241 shown in Table 3. References [12, 13] have been used to define two different  
 242 set of erosion model parameters ([12] for Simulation #1 and [13] for Sim-  
 243 ulation #2) for the same configuration under study. Following the original  
 244 works [12, 13], velocity,  $\mathbf{u}_1$ , in (13) enters in  $\text{ft s}^{-1}$  in Simulation #1 and in  
 245  $\text{m s}^{-1}$  in Simulation #2.

Table 3: Erosion model parameters

	Simulation #1 [12]	Simulation #2 [13]
$k$	$2.056 \cdot 10^{-8}$	$1.7 \cdot 10^{-8}$
$n$	1.73	2.3
$a$	-34.79	-7.0
$b$	12.3	5.45
$x$	0.147	0.4
$y$	-0.745	-0.9
$z$	1.0	1.556
$\omega$	5.205	-3.4
$\theta_0$ , [deg]	10	23

246 *3.1.2. Results*

247 The results obtained with the CFD-based erosion computation is split  
 248 into the three segregated steps used for the calculation: flow simulation, par-  
 249 ticle tracking and erosion calculation.

250

251 ***Flow simulation***

252 The CFD solution has been analyzed after 200000 iterations and approxi-  
 253 mately 3 times the residence time for the fluid in the pipe, enough to develop  
 254 a representative flow structure as shown in Figure 3. The simulation took  
 255 41h of computational time in 20 Intel Xeon Gold 6230 cores at 2.1 GHz in  
 256 the CESVIMA-UPM supercomputer center. The mesh used is valid for a  
 257 Large Eddy Simulation (LES) description of the flow, which is adequate for  
 258 this application. A time-average of the flow variables is computed using 10  
 259 snapshots over a period of 100000 iterations, which corresponds with 0.1 s  
 260 of physical time using a time step of  $1 \cdot 10^{-6}$  s. Particles are integrated over  
 261 this averaged flow.

262

263 ***Particle tracking***

264 Particles are injected in the initial section and integrated along the pipe.  
 265 The injected particles collide with the first bend and bounce multiple times  
 266 in the upstream section of the pipe. As a result, they reach the second bend

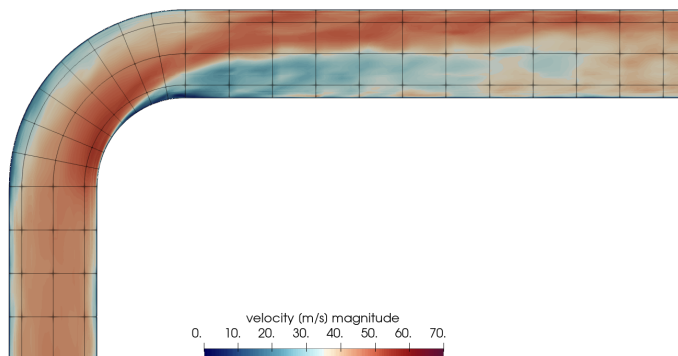


Figure 3: Velocity magnitude contours in the middle section of the pipe in the second elbow corresponding to the test section

267 practically in line with the main flow but with an appreciable loss of velocity,  
 268 due both to gravity and to the loss of energy in the successive bounces. In  
 269 Figure 4 the detail of the particle trajectories in the upper elbow is shown.  
 270 More than 2 million impacts have been analysed in each simulation.

271

### 272 *Erosion calculation*

273 In Figure 5 we detail the erosion results, which feature the characteristic  
 274 V-shape scar on the elbow. Let us mention that the shape of the erosion scar  
 275 is associated with the treatment of the wall, the V-shaped scar at the elbow  
 276 appears with a smooth wall, once the wall becomes rough this characteristic  
 277 shape changes [81]. Besides, such erosion pattern can be deteriorated if inter-  
 278 particle collisions are considered with the Discrete Phase Model for higher  
 279 mass loadings [74, 82]. Comparison with experimental results of Simulations  
 280 #1 and #2 are shown in Figure 6. The erosion rate in the external mean  
 281 section of the second elbow is represented together with the experimental  
 282 results with a good agreement. A similar simulation is presented, along with  
 283 the experimental results, in [11]. The simulation results shown in Figure 6  
 284 are better than those obtained in that work, where the predicted erosion was  
 285 10 times higher than the measured. We believe that the lack of agreement  
 286 between simulation and experiments in [11] might be due to an inappropriate

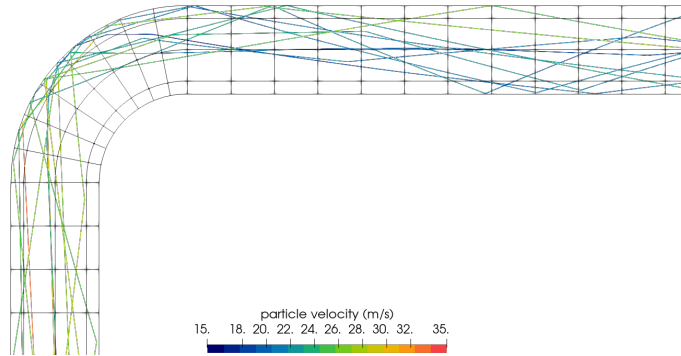


Figure 4: Detail of the particle trajectories in the upper elbow. The color of the trace represents the speed of the particle

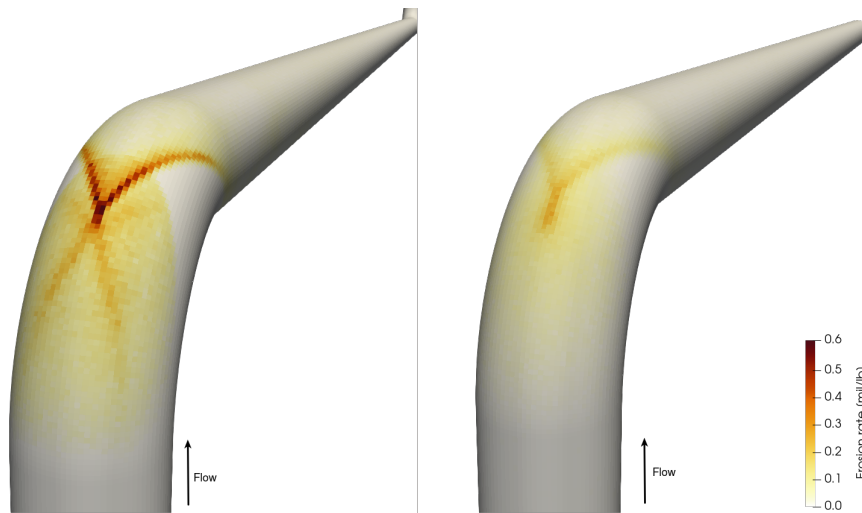


Figure 5: Estimated erosion rate in the first (left) and second (right) elbow for the Simulation #1

287 selection of a parameter in the model. In particular, the constant  $k$  in Finnie's  
 288 model is not specified in [11], so that might be the source of the discrepancy.

289 In summary, we confirm the capabilities of the developed method to esti-  
 290 mate erosion. This methodology comprises the flow simulation, the particle  
 291 tracking and the erosion computation through a model that takes into ac-



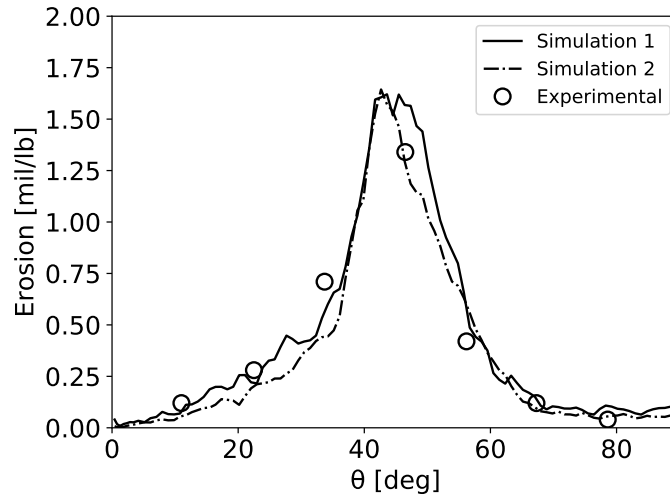


Figure 6: Penetration rate prediction ( $\text{mil lb}^{-1}$ ) in the second elbow for Simulations #1 and #2 compared with experimental results from [11]

292 count the impact of a high number of particles. The erosion rate results have  
 293 shown the importance of the model parameters (especially the parameter  $k$ ).  
 294 Finally a qualitative examination of the effect of the accurate representation  
 295 of the surface of the pipe, allowed by the high order method used, is included  
 296 in Appendix A.

297 *3.2. Corrosion*

298 In this section, the methodology introduced in Section 2.3 for the corro-  
 299 sion prediction is validated in carbon steel pipelines at different flow regimes  
 300 using the experimental data presented by Kanwar [77].

301 *3.2.1. Setup*

302 For the validation of the corrosion model, a simulation of a carbon steel  
 303 straight pipe of 5 m long and 0.1 m diameter filled by a mixture of oil (Ar-  
 304 copac90) and water is chosen. As previously mentioned, the flow regime de-  
 305 pends on the amount of water and oil entering the pipe, so we define  $V_{o/w}[\%]$   
 306 as the volume flow rate of oil over the total volume flow rate and  $V_{w/o}[\%]$   
 307 as the volume flow rate of water over the total volume flow rate. Figure 7  
 308 [26] illustrates the flow patterns of the oil–water two phase fluid in pipelines  
 309 with a mean flow velocity,  $U_0$ , where  $V_{os} = V_{o/w}U_0$  and  $V_{ws} = V_{w/o}U_0$  are oil  
 310 superficial velocity and water superficial velocity, respectively.

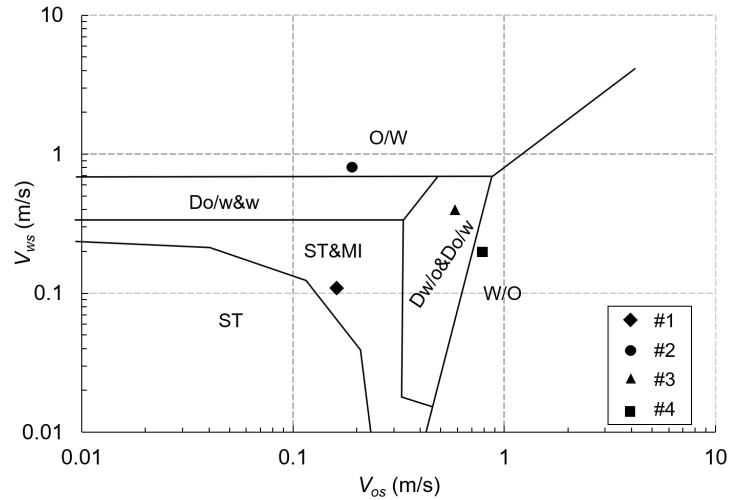


Figure 7: Flow pattern of the oil–water two phase fluid in pipelines [26] and numerical simulations. ST is stratified flow, ST & MI is stratified flow with mixing on the pipe wall surface, O/W is oil in water emulsion, W/O is water in oil emulsion,  $D_{o/w\&w}$  is dispersion of oil in water emulsion and water, and  $D_{w/o}$  &  $D_{o/w}$  is dispersion of water in oil and oil in water

311 For this validation four different regimes have been selected (labeled from

312 #1 to #4 in Figure 7). For each one, a CFD simulation will be performed  
 313 to get the flow regime and the flow variables in the domain; after that, the  
 314 corrosion rate will be computed following Section 2.3 and compared with  
 315 experimental results.

316 The fluid properties used for the simulation are shown in the Table 5.  
 317 After a detailed convergence analysis, not included here, the simulation has  
 318 been carried out on a mesh of 2800 elements (140 sections of 20 elements)  
 319 and polynomial order 4 (see Figure 8). The resolution of this mesh permits  
 320 to capture the interfaces between the two phases with an interface width,  
 321  $\varepsilon = 0.01$  m. Finally, the mobility is set to  $M_0 = 8 \cdot 10^{-7} \text{ m}^3\text{kg}^{-1}\text{s}$ , which  
 322 results in an appropriate balance between convection and chemical diffusion  
 323 effects. Additionally, a low value of this parameter is beneficial from a nu-  
 324 merical point of view, as it reduces the stiffness of the resulting system of  
 325 equations. For more details on the proper selection of the mobility parame-  
 326 ter, see [63].

327

Table 4: Parameters for the four different CFD-based corrosion prediction simulations

Simulation	#1	#2	#3	#4
Flow regime	ST & MI	O/W	$D_{w/o}$ & $D_{o/w}$	W/O
$V_{o/w}$ , [%]	60	20	60	80
$U_0$ , [ $\text{m s}^{-1}$ ]	0.28	1	1	1

Table 5: Fluid properties for the CFD-based corrosion computations

	Water	Arcopac90
Density, [ $\text{kg m}^{-3}$ ]	992.25	825.0
Kinematic viscosity, [ $\text{Pa s}^{-1}$ ]	$0.653 \cdot 10^{-3}$	$15.0 \cdot 10^{-3}$
Surface tension, [ $\text{N m}^{-1}$ ]		0.028
Contact angle, [deg]		45

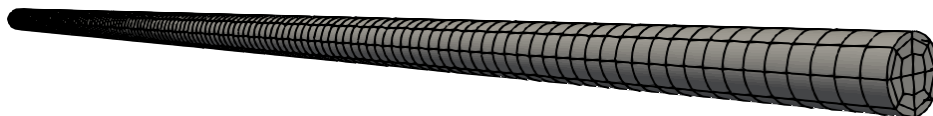


Figure 8: Detail of the high order mesh used for the CFD-based corrosion computations of a straight pipe

### 3.2.2. Results

This section includes the results for the four simulations detailed in Table 4. The results include the prediction of the flow regime and the computation of the corrosion rate.

#### *Flow simulation*

To accelerate convergence and get a representative flow structure, the simulation has been started with 100000 iterations and a polynomial order of 3, followed by 100000 iterations with polynomial order 4. The last 100000 iterations, which correspond with 4s of physical time using a time step of  $2 \cdot 10^{-5}$  s, have been used to perform the post-processing. The simulation took 48h of computational time in 20 Intel Xeon Gold 6230 cores at 2.1 GHz in the CESVIMA-UPM supercomputer center.

According to experimental results [77], for oil volumetric fractions of 60% and below, stratified flow is maintained with a continuous film of water at the bottom of the pipe, causing corrosion in that area. For oil volumetric fractions of 80% and above, the water film breaks down and the water forms a dispersed phase in a continuous oil matrix. With this flow pattern corrosion is dramatically reduced.

Figures 9, 10, 11 and 12 show the flow structures at the different flow regimes. Top view represents the *pure* water representation ( $c_w > 0.25$ ) and the bottom view represents the mixture interface between water ( $c_w < 0.25$ ) and oil ( $c_o < 0.25$ ). The parameters  $c_w$  and  $c_o$  are obtained from the Cahn-Hilliard equation (2) as  $c_w = c$  and  $c_o = 1 - c$ . All simulations consider a stratified flow in the inflow boundary condition. It is noticeable that the flow needs a distance of 3m to develop completely. This distance will not be taken into account for the corrosion analysis. Figure 9 shows a stratified

355 flow with mixing at the interface (ST & MI). Figure 10 depicts how the oil  
356 is emulsified into the water (O/W). In contrast, the water is emulsified into  
357 the oil (W/O) in Figure 12. In this case, it is remarkable how the surface  
358 of pipe steel wetted by water is very small, leading to negligible corrosion.  
359 Finally, Figure 11 shows a typical dispersed flow of water in oil and oil in  
360 water ( $D_{w/o}$  &  $D_{o/w}$ ). The flow regimes were correctly predicted (according  
361 to experimental data) by our high-order phase-field CFD technique without  
362 additional information.

363

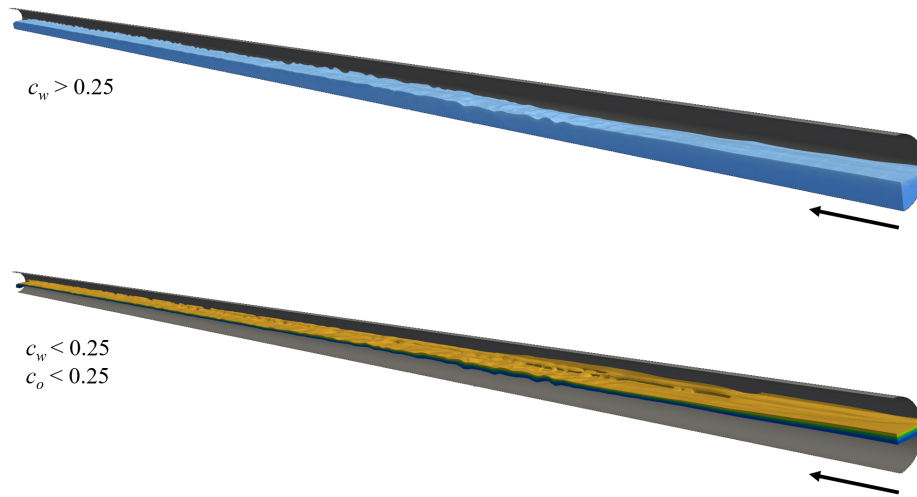


Figure 9: Simulation #1 with stratified flow with mixing on the pipe wall surface (ST&MI) flow regime for  $o/w = 60\%$  and  $U_0 = 0.28 \text{ m s}^{-1}$

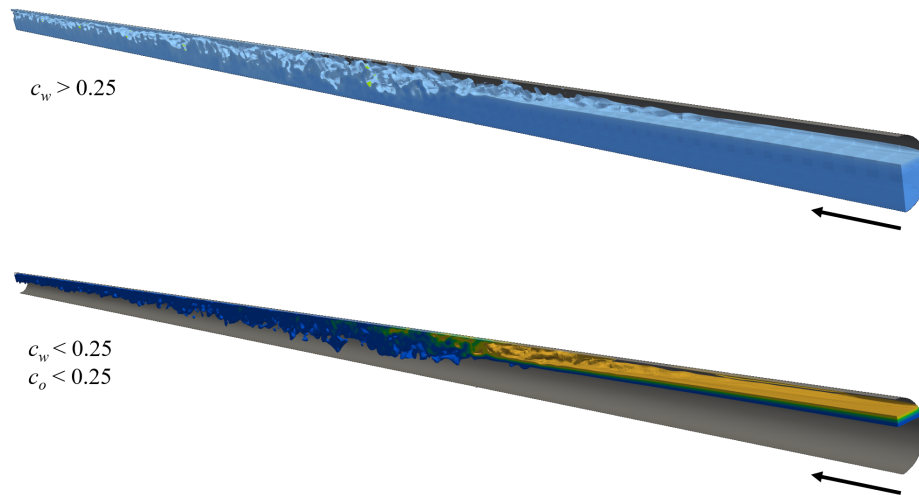


Figure 10: Simulation #2 with oil in water emulsion (O/W) flow regime for  $V_{o/w} = 20\%$  and  $U_0 = 1 \text{ m s}^{-1}$

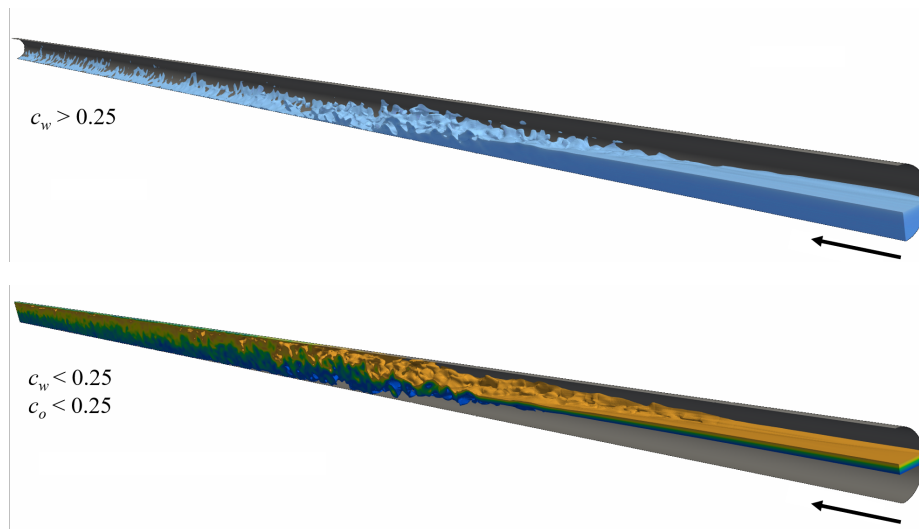


Figure 11: Simulation #3 with dispersion of water in oil and oil in water ( $D_{w/o}$  &  $D_{o/w}$ ) flow regime for  $V_{o/w} = 60\%$  and  $U_0 = 1 \text{ m s}^{-1}$

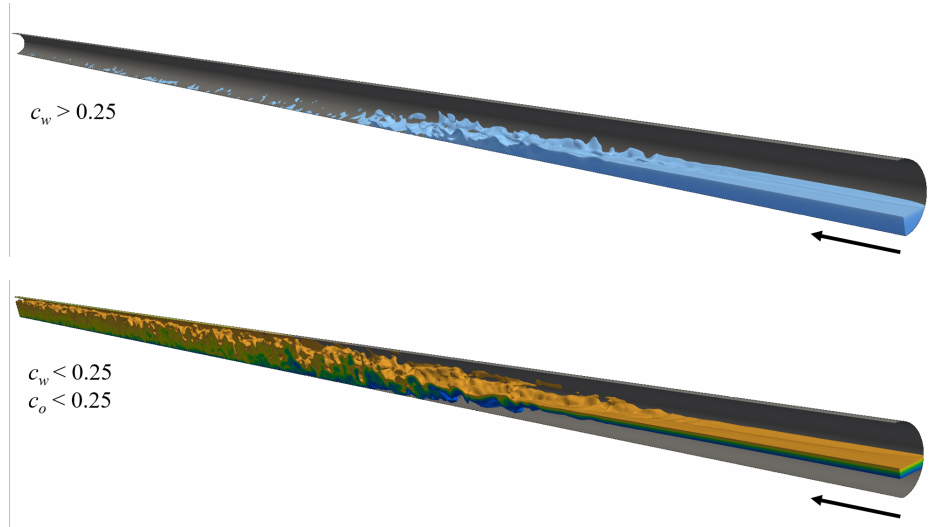


Figure 12: Simulation #4 with water in oil emulsion (W/O) flow regime for  $V_{o/w} = 80\%$  and  $U_0 = 1 \text{ m s}^{-1}$

### Corrosion calculation

As mentioned before, the estimation of the corrosion rate has been performed using (16) and (17) over 20 time instants corresponding to 4 s of physical time and the average value has been obtained. The corrosion has been analyzed in the section where the flow is developed  $3 \text{ m} < x < 4.95 \text{ m}$ . The outlet section has been excluded to avoid pollution from the boundary condition in the corrosion rate.

Table 6: Maximum corrosion rate ( $\text{CR}_{max}$ ) and shear in water wetted walls  $\tau_{max}$ . Numerical results are obtained with our methodology. Experimental results are taken from [77]

Sim.	$U_0 [\text{m s}^{-1}]$	$V_{o/w} [\%]$	$\text{CR}_{max} [\text{mm year}^{-1}]$		$\tau_{max} [\text{N m}^{-1}]$	
			Num.	Exp.	Num.	Exp.
#1	0.28	60	4.90	4.70	0.52	0.34
#2	1.00	20	5.57	6.35	6.50	6.98
#3	1.00	60	6.27	6.40	7.80	7.55
#4	1.00	80	-	negligible	-	negligible

The results obtained for the maximum corrosion rate along with the

372 maximum shear in wet walls is represented in Table 6. The results are  
373 in good agreement with the experimental results, showing the validity of  
374 the methodology. It is especially noticeable the accuracy in Simulation #4  
375 ( $V_{o/w} = 80\%$ ). In this case the water film breaks and the water goes on  
376 to form a dispersed phase in the oil and the corrosion is negligible. It has  
377 been confirmed that the correct simulation of the flow regime carried with  
378 our phase-field high-order CFD technique is paramount for the accuracy of  
379 the corrosion rate computation.

#### 380 4. Conclusions

381 This work has shown that the methodology presented is able to predict  
382 erosion-corrosion in pipes carrying multiphase mixtures. Two test cases rele-  
383 vant to oil and gas industry have been studied, showing the reliability of the  
384 incompressible Navier-Stokes/Cahn-Hilliard equations approximated with a  
385 high-order Discontinuous Galerkin Spectral Element Method (DGSEM) to  
386 solve hydrocarbon transport pipes problems.

387 As far as erosion modeling is concerned, the results for a one phase com-  
388 plex configuration are very satisfactory. The particles reach the study zone  
389 after multiple bounces and an upward path. With the first bounces the par-  
390 ticles lose speed and their trajectories become oblique to the main flow. In  
391 the upward section the particles align again with the flow and gain speed.  
392 Finally they collide with the second elbow eroding the wall section under  
393 study. The results show a good behavior of the particle drag model, the  
394 rebound model and the erosion model.

395 As far as corrosion modeling is concerned, the results are also in good  
396 agreement with the experiments. In this case, the complexity is not driven  
397 by the geometry (a straight pipe is considered) but by the flow pattern. A  
398 multiphase experiment is considered where different volumetric fractions of  
399 water and oil are used in the inlet boundary. Depending on these volumet-  
400 ric fractions, different flow regimes can be experimentally observed. These  
401 regimes are appropriately captured by our CFD methodology, resulting in an  
402 accurate computation of the corrosion rate.



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## 413 **Appendix A. High order representation of the surface**

414 Figure A.13 illustrates the effect of the CFD method used to compute the  
415 erosion. The usage of a discontinuous Galerkin method, permits a high order  
416 representation of the surface. Figure A.13 compares the distribution of the  
417 particles impacts using first order surface representation (right) and fourth  
418 order surface representation (left). An accurate prediction of the impact  
419 angle is critical to estimate the erosion pattern caused by secondary rebounds  
420 along the elbow and the single particle erosion that has a highly nonlinear  
421 dependency with the incoming angle. As can be seen, the high order method  
422 permits to obtain an accurate solution with a low number of elements. This  
423 fact is especially relevant in multiphase simulations of complex geometries  
424 where a nearly constant mesh refinement is required to capture the interface  
425 between the phases.

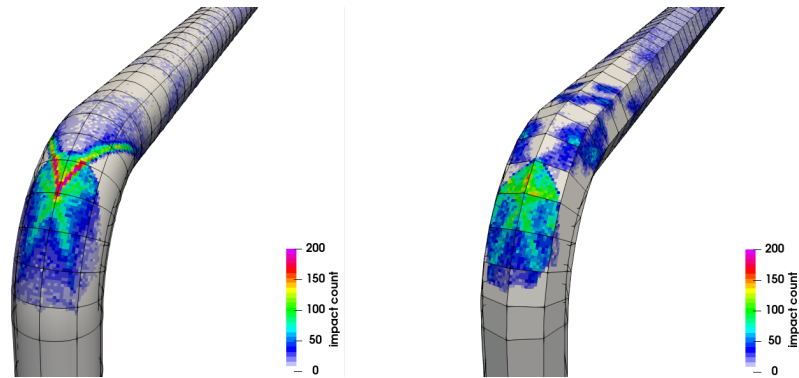


Figure A.13: Influence of the surface polynomial order representation (1st order right, 4th order left) on the impact distribution along the first elbow

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