



Titre: Numerical simulations with the P-Hydroslag model to predict Title: phosphorus removal by steel slag filters

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1 2	Numerical simulations with the P-Hydroslag model to predict phosphorus removal by steel slag filters
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7	ABSTRACT
8	The first version of the P-Hydroslag model for numerical simulations of steel slag filters is presented.
9	This model main original feature is the implementation of slag exhaustion behavior, crystal growth and
10	crystal size effect on crystal solubility, and crystal accumulation effect on slag dissolution. The model
11	includes four mineral phases: calcite, monetite, homogeneous hydroxyapatite (constant size and
12	solubility) and heterogeneous hydroxyapatite (increasing size and decreasing solubility). In the proposed
13	model, slag behavior is represented by CaO dissolution kinetic rate and exhaustion equations; while slag
14	dissolution is limited by a diffusion rate through a crystal layer. An experimental test for measurement
15	of exhaustion equations is provided. The model was calibrated and validated with an experimental
16	program made of three phases. Firstly, batch tests with 300g slag sample in synthetic solutions were
17	conducted for the determination of exhaustion equation. Secondly, a slag filter column test fed with
18	synthetic solution was run for 623 days, divided into 9 cells and sampled at the end of the experiment.
19	Finally, the column was dismantled, sampled and analyzed with XRD, TEM and SEM. Experimental
20	column curves for pH, oPO ₄ , Ca and inorganic carbon were well predicted by the model. Crystal sizes
21	measured by XRD and TEM validated the hypothesis for homogeneous precipitation while SEM
22	observations validated the thin crystal layer hypothesis.
23	KEYWORDS

slag, phosphorus, wastewater treatment modelling, PHREEQC, hydroxyapatite, calcite, precipitation

25 ABBREVIATIONS

26	<u>Symbol</u>	Description
27	General abbre	eviations
28	BOF	Basic oxygen furnace
29	CW	Constructed wetands
30	EAF	Electric arc furnace
31	HRT_{V}	Hydraulic retention time of voids
32	MONtoHAP	Transformation of MON into HAP
33	o-PO ₄	Ortho-phosphates
34	SEM	Scanning electron microscope
35	TEM	Transmission electron microscope
36	TIC	Total inorganic carbon
37	WW	Wastewater
38	XRD	X-Ray diffraction
39	Abbreviations	s for mineral phases
40	CAL	Calcite CaCO ₃
41	HAP	Hydroxyapatite Ca ₅ OH(PO ₄) ₃
42	HAP_HO	Primary hydroxyapatite via homogeneous precipitation Ca ₅ OH(PO ₄) ₃
43	HAP_HE	Primary hydroxyapatite via heterogeneous precipitation $Ca_5OH(PO_4)_3$
44	HAP2	Secondary hydroxyapatite via monetite transformation $Ca_4OH_2(PO_4)_2$
45	MON	Monetite CaHPO ₄
46	Symbol	Description Units (value)
47	Constants	
48	a_{HAP_0}	HAP crystal size in homogeneous precipitation [m]
49	B_1 and B_2	Regression coefficients in k_{diss} exhaustion function
50	D^*	Dispersivity (transport model) [cm]
51	D_n Exchange fa	actor between effective and immobile porosity (transport model) $[s^{-1}]$
52	k _{CAL}	CAL precipitation constant [mol CAL/s m ² slag]
53	k_{HAP}	HAP precipitation constant [mol HAP/s m ² slag]
54	k _{MON}	MON precipitation constant [mol MON/s m ² slag]
55	k _{MONtoHAP}	MONtoHAP precipitation constant [M HAP2/(M MON s)]
56	K _{spCAL}	Solubility product for CAL [M ²]

57	K _{spHAP_bulk}	Bulk solubility product for HAP	[10 ⁻⁵⁷ M ⁹]
58	K _{spHAP_HO}	Solubility product for HAP_HO	[M ⁹]
59	K _{spMON}	Solubility product for MON	[M ²]
60	L _{HAP}	L/D ratio for columnar HAP crystals	[-]
61	mv_{exp}	Slag mass to water volume ratio in a batch test	[g/mL]
62	MW _{HAP}	HAP Molecular weight	[502 g/mol]
63	n	Total porosity in the slag filter	[-]
64	n _e	Effective porosity in the slag filter	[-]
65	n _{im}	Immobile porosity in the slag filter	[-]
66	P_{1}, P_{2}, P_{3} ar	nd P_4 Regression coefficients in pH_{sat} exhaustion function	
67	R	Ideal gas constant	[8.31 J mol ⁻¹ K ⁻¹]
68	S	Slag specific surface	[m ² /m ³]
69	se _{HAP_0}	initial HAP seeds concentration	[seeds/L]
70	SIc	Critical saturation index between HAP_HE and HAP_HO	[-]
71	Т	Temperature	[K]
72	γ	HAP mean free surface energy	[87 mJ/m ³]
73	$ ho_{barr}$	Crystal concentration in the crystal barrier	[g crystal/m ³]
74	$ ho_{HAP}$	HAP crystal density [3 60	000 g/m³]
75	$ ho_{slag}$	Slag grain density	[3.8 g/mL]
76	Rates, func	tions and variables	
77	a_{HAP}	Mean HAP crystal size	[m]
78	b _{im}	Moles of an element in immobile porosity (transport model)	[mol]
79	C -	Total dissolved concentration for an element (transport model)	[mol/kgw]
80	C _e	C in effective porosity (transport model)	[mol/kgw]
81	C _{im}	C in immobile porosity (transport model)	[mol/kgw]
82	<i>СаОІ</i> _{ВАТН}	Leached CaO in a acid bath	[mol/g]
83	CaO _{KTEST}	Leached CaO in a batch test	[mol/g]
84	$CaOl_{TOT}$	Cumulative leached CaO in a batch test	[mol/g]
85	D _{barr}	Diffusion coefficient in the crystal barrier	[m ² /s]
86	d_{barr}	Thickness of the crystal barrier	[m]
87	k _{diss}	Slag dissolution constant [mol	CaO/m ² slag]

88	K _{spHAP_HE}	Solubility product for HAP_HE	[M ⁹]
89	pH _{sat}	Saturation pH in the slag filter	[-]
90	q Concent	tration in the solid phase for an element (transport model)	[mol/kgw]
91	r _{CAL}	CAL precipitation rate	[M CAL/s]
92	r _{diff}	CaO diffusion rate through crystal barrier	[M CaO/s]
93	r _{diss}	Slag dissolution rate	[M CaO/s]
94	r _{HAP_HE}	Primary heterogeneous HAP precipitation rate	[M HAP/s]
95	r _{HAP_HO}	Primary homogenous HAP precipitation rate	[M HAP/s]
96	r _{MON}	MON precipitation rate	[M MON/s]
97	r _{MONtoHAP}	Secondary HAP precipitation rate	[M HAP2/s]
98	S _{HAP}	HAP molar specific surface	[m²/mol]
99	se _{HAP}	HAP seeds concentration	[units/L]
100	SF _{diff}	Step function in diffusion rate	[-]
101	SF _{diss}	Step function in dissolution rate	[-]
102	SF _{HAP_HE}	Step function in HAP_HE rate	[-]
103	SF _{HAP_HO}	Step function in HAP_HO rate	[-]
104	SI _{HAP_HE}	Saturation index for HAP_HE	[-]
105	SI _{HAP_HO}	Saturation index for HAP_HO	[-]
106	t	Time (transport model)	[s]
107	v	Pore water flow velocity (transport model)	[m/s]
108	x	1D distance (transport model)	[m]
109	X _{CaO}	Total leached CaO in the slag filter	[M]

110

111 1 Introduction

112 Steel slag filters are an effective and passive technology for phosphorus removal from wastewater,

allowing typical municipal effluent o-PO₄ concentration below 0.5 mg P/L (Koiv et al., 2016). Design tools

114 for slag filters are not yet developed and full scale slag filters cannot be implemented without expensive

pilot tests. The main issue related to steel slag filter operation relies on filter exhaustion and a relatively

rapid drop of removal efficiency (Chazarenc et al., 2008). A tool providing the effect of influent

composition and operational conditions (type and size of slag, influent flowrate, filter geometry) on slag
filter effluent o-PO₄ concentration and longevity would facilitate the design of these systems. This paper
presents the P-Hydroslag model, a new model adapted for steel slag filter simulations considering
influent composition and void hydraulic retention time (HRT_v) while being compatible with accepted
physicochemical modeling frameworks (Lizarralde et al., 2015; Mbamba, Batstone et al., 2015; Mbamba,
Tait et al., 2015).

123 Sorption isotherms were largely proposed as a design tool for steel slag filters (Vohla et al., 2011), but 124 this method does not consider void precipitation and long-term changes in the material properties, explaining why isotherms could not yet predict correctly full-scale behavior. The k-C* model traditionally 125 126 used for constructed wetlands (Kadlec and Wallace, 2009) successfully predicted steel slag filter 127 performance (Barca et al., 2013). This method may be suitable for design, but it cannot estimate the 128 lifetime of the filter. A general correlation between material CaO content and P retention capacity based 129 on several studies was proposed by Vohla et al. (2011), highlighting the importance of CaO dissolution in 130 retention mechanisms. Such a design tool, however, does not consider important aspects as CaO 131 availability, HRT_v or influent composition, leading to a prediction uncertainty that is not acceptable for 132 design purpose. Finally, a predictive model based on several material properties, HRT_V and inlet P 133 concentration was proposed (Penn et al., 2016). This model's strengths were to consider both Fe-Al and 134 Ca based materials, propose an empiric relationship between material buffering capacity and P 135 retention capacity, and predict P retention for both lab-scale and pilot-scale systems. This model, 136 however, did not include direct measurement of kinetic rates, and would not be compatible with 137 general physicochemical modeling frameworks in wastewater treatment. 138 Two previous modeling studies were published in the recent years. The first study (Claveau-Mallet et al.,

139 2012) qualitatively described concepts forming the basis of the model, including slag dissolution,

140 hydroxyapatite precipitation, crystal formation and accumulation in voids, and effect of velocity on

crystal accumulation. In the second study (Claveau-Mallet et al., 2014), concepts were translated into a prototype model including mathematical equations for precipitation and slag exhaustion, and a proposition of laboratory protocol for slag characterization was presented. Numerical simulations of a slag filter were performed on the base of this prototype model without experimental program for calibration. Results were realistic but overestimated the filter longevity. Predictions from the 2014 prototype model were compared to full-scale real data in a recent study (Koiv et al., 2016) in which longevity was overestimated.

In this paper, the first full version of the P-Hydroslag (standing for Phosphorus-hydroxyapatite-slag)
model is presented. The P-Hydroslag model is similar to the 2014 prototype model, with additional
features for diffusion barrier and crystal growth, a refined characterization of exhaustion equations, and
a complete model equation matrix. The objectives were to calibrate the P-Hydroslag model with
experimental data and evaluate the validity and realism of the model.

153 2 Material and Methods

154 2.1 Slag media

155 5-10 mm electric arc furnace steel slag produced by Arcelor Mittal and provided by Minéraux Harsco

156 (Contrecoeur, Canada) was used (33% Fe₂O₃, 30% CaO, 16% SiO₂, 12% MgO. 6% Al₂O₃ and 3% other

metallic oxides). Its density (3.8) and specific surface (0.308 m^2/g) were determined according to the

ASTM C127-04 standard (ASTM, 2004) and the Brunauer, Emmet and Teller method (Lowell et al., 2004).

159 Slag from the same source was previously studied by the authors' research team for wastewater

treatment applications (Claveau-Mallet et al., 2015; Claveau-Mallet et al., 2013; Koiv et al., 2016) or

161 modeling studies (Claveau-Mallet et al., 2014; Claveau-Mallet et al., 2012).

162 2.2 Column test

A vertical filter column filled with slag was fed from its base with a synthetic wastewater in a saturated mode for a total duration of 623 days at approximatively 25°C. The column size was 159 cm in length and 10 cm in internal diameter. The synthetic wastewater solution consisted of K₂HPO₄, KH₂PO₄, NaHCO₃ and CaCl₂ in tap water. The influent mean composition was pH of 7.80 ± 0.2, ortho-phosphates (o-PO₄) of 8.9 ± 2.0 mg P/L, total inorganic carbon (TIC) of 22 ± 2 mg C/L, Ca of 54 ± 14 mg/L and alkalinity of 102 ± 3 mg CaCO₃/L. The influent flowrate was 6.9 ± 1.0 mL/min for the first 517 days and 3.4 ± 0.5 mL/min for the remaining 106 days of operation.

The column was divided into 11 virtual cells for the filling step, identified #0 to #10, with #0 at bottom (inlet) and #10 at top (outlet). Cells #1 to #9 were 15 cm long and had a sampling hole in the middle. Cells #0 and #10 were 7.5 cm long and had no sampling hole to provide a slag transition zone between the inlet/outlet tubing and sampling zones. While filling the column, two 300 g slag samples were taken from each cell using a standard sampling procedure for aggregate materials (ASTM C702, 2011). Slag samples were used in batch kinetic tests (presented in section 2.3). The total slag mass in the column was 24.24 kg, resulting in a 49.2% porosity.

177 The feeding barrel, column effluent and cells were sampled and analyzed periodically for pH, o-PO₄, 178 filtered Ca, settled TIC, total P and alkalinity, using standard procedures (APHA, AWWA and WEF, 2005). 179 A maximum of 3 cells were sampled in the same day to minimize perturbation, resulting in a monthly 180 sampling frequency for each cell (twice a month in the second feeding phase). The feeding barrel and 181 effluent column sampling frequency was weekly for pH and once or twice a month for the other 182 parameters. Tracer tests were conducted after 12, 69, 82, 107, 187, 271, 376 and 558 days. Rhodamine 183 at a concentration of 20 mg/L was used as a tracer and measured in the effluent using 184 spectrofluorometry.

At the end of operation, feeding was stopped and the column was kept saturated for 6 days before
 dismantling. Upon dismantling, pore water was first sampled and analyzed, then the column was cut

187 into 4 sections to ensure efficient solids sampling. For each cell, three samples were taken: first, several 188 slag particles sampled before doing any major disturbance of the slag media (for scanning electron 189 microscope (SEM) analysis); then slag was washed with water in a large pan and precipitates were 190 sampled by sedimentation (for X-ray diffraction (XRD) and transmission electron microscope (TEM) 191 analyses); finally, a 300 g of washed slag sample was taken for kinetic tests (described in section 2.3). 192 Precipitates were air-dried for 3 days, sieved at mesh 60 and cleaned from slag dust with a strong 193 magnet. Precipitates were analyzed with XRD using a Philipps X'Pert diffractometer operated at 50 kV 194 and 40 mA, using the Bragg-Brentano geometry and a Cuk_{α} radiation. The Scherrer equation (Cullity, 195 2001) was used to estimate mean crystal sizes from diffractograms, using the ~26.1° peak for 196 hydroxyapatite (HAP) and ~29.4° peak for calcite (CAL). Precipitates of cells 1, 2, 3 and 8 were analyzed 197 using TEM with the bright field imaging technique (Jeol JEM-2100f field emission gun microscope, 200 198 kV). Before TEM analysis, samples were prepared with a 30-s ultrasound bath in methanol, and placed 199 on a copper grid covered with Formwar lightly coated with amorphous carbon. Undisturbed slag 200 particles of cells 1, 2, 3, 5 and 8 were analyzed with SEM using a Jeol JSM-7600F microscope (2.0 kV, LEI 201 or SEI detector).

202 2.3 Batch kinetic tests

203 The batch kinetic test method is described in another reference (Claveau-Mallet et al., 2014) and is 204 intended to produce exhaustion equations. The batch test included 5 identical phases. In a phase, the 205 slag sample was placed in a 1L Erlenmeyer flask containing 350 or 700 mL of a wastewater solution. The 206 Erlenmeyer flask was placed in a gyratory shaker at 160 rpm. The flask was closed with a rubber cap that 207 contained three airtight holes; one for a pH probe, one for a sampling tubing and one for a tubing 208 connected to a N₂ gas balloon. The synthetic solution was composed of KH₂PO₄, K₂HPO₄, NaHCO₃ and 209 CaCl₂ dissolved in tap or distilled water. Four solutions with different concentrations were used to test 210 the method in a realistic range of wastewater types (pH of 6.5 to 7.9, o-PO₄ of 8 to 24 m P/L, Ca of 17 to

211 50 mg/L, TIC of 0.5 to 24 mg C/L and alkalinity of 3 to 107 mg CaCO₃/L). At time zero, slag was inserted. 212 pH was monitored for 3 to 4 days. Three intermediary 20-mL samples were taken and analyzed for o-213 PO₄, filtered Ca and filtered TIC. When necessary, a linear correction against time was applied to pH 214 measurements to account for probe drift. After this test, the slag sample was rinsed and immediately 215 transferred to a 160 rpm shaken HNO₃ acid bath of known volume and concentration for 3 to 5 days. 216 After the acid bath, pH was measured and the corresponding leached CaO from the slag was computed 217 using numerical simulations (explained in section 2.5.2). After the acid bath, the slag sample was 218 carefully rinsed and used again for a subsequent phase.

Each 300-g slag sample from cells 1 to 8 was used for a 5-phase kinetic test, resulting in 16 kinetic tests
(2 replicates per cell). One-phase batch tests on dismantled column slag samples were performed for
cells 1 to 9.

222 2.4 Model description

223 2.4.1 Precipitation

224 The model's Gujer matrix is presented in appendix. Three mineral phases are included: HAP typically 225 found in slag filters (Baker et al., 1998), monetite (MON) as intermediary phase and CAL. The 226 transformation of MON into HAP (MONtoHAP) was modelled as precipitation of HAP2, an artificial phase 227 composed of ions missing from MON before being HAP. Precipitation rates for HAP, MON and CAL were 228 formulated with a basic expression rate = k x SI, with k being a constant normalized with slag surface and 229 SI the saturation index. The bulk solubility constant for HAP ($K_{spHAP \ bulk}$) was set at 10⁻⁵⁷ (Stumm and 230 Morgan, 1996a), within the 10⁻⁵⁵ to 10⁻⁶³ range reported in the literature (Lundager Madsen, 2008; 231 Oelkers et al., 2009; Parkhurst and Appelo, 1999; Stumm and Morgan, 1996a). K_{spCAL} was set at 10^{-7.5}, assuming an intermediary state between crystalline calcite (10^{-8.48}, PHREEQC database (Parkhurst and 232

Appelo, 1999)) and hydrated calcium carbonate (10^{-7.144}, MINTEQ database (Allison et al., 1991)).

234 K_{spMON} was set at 10⁻⁷ from Valsami-Jones (Valsami-Jones, 2001).

235 Two types of HAP were included to account for two types of precipitation. HAP_HO was HAP formed by 236 homogeneous precipitation (new seeds precipitated in voids, or spontaneous precipitation) while HAP HE was HAP formed by heterogeneous precipitation (on existing surfaces - crystal growth). HAP HE 237 238 occurred below a critical saturation index (SI_c) while HAP_HO occurred over SI_c . In homogeneous 239 precipitation, crystal size was assumed to be constant. Using equation 1 for solubility of fine particles (Stumm and Morgan, 1996b), $K_{spHAP HO}$ was set at 10⁻⁴⁶. Temperature was set at 298 K and specific 240 241 surface (S_{HAP}) was calculated assuming crystal size $a_{HAP 0} = 31.3$ nm, columnar shape for HAP and a 242 L/D ratio (L_{HAP}) of 50. The value assumed for $a_{HAP 0}$ is close to crystal size measured in this study 243 (presented in a following section) and measurements made in previous studies (Claveau-Mallet et al., 2012, 2013). A value of 10^{-46} for K_{spHAP_HO} is consistent with equilibrium state generally observed in 244 slag filters, assumed when effluent pH is over 10 with high HRT_V (Table 1 and Figure 1). In Table 1, Ca^{2+} , 245 OH^{-} and PO_{4}^{3-} activities were determined with PHREEQC and used for apparent solubility calculation. 246 247 Studies conducted with hydrated oil shale ash were included in Table 1 as they behave in a similar way then slag. Resulting mean and median were 10^{-45,7} and 10^{-46,0} from 389 data points. 248

249
$$log(K_{spHAP_{HO}}) = log(K_{spHAP_{bulk}}) + \frac{\frac{2}{3}\gamma S_{HAP}}{2.3RT}$$
[1]

Table 1. Apparent HAP solubility from reported alkaline filter effluent with pH≥10, based on reported
 pH, Ca and o-PO₄ concentration.

Matorial	Size	Influent	Type of	Apparent HAP	$\log K_{sp}$ of	f Re f	Nb of data
Wateria	(mm)	linuent	study	mean	median	f	point s
			lab –				
Sas	5-10	Solution	column	-45.55	-45.94	1	35

			lab —				
EAF slag	5-10	Solution	column	-45.18	-45.72	2	103
			lab –				
Sas	2.5-5	Solution	column	-48.06	-48.17	3	6
			pilot –				
Sas	5-10	real fishfarm WW	column	-45.23	-45.39	4	46
Hydr. Oil shale ash	5-20	real domestic WW	pilot – CW	-46.32	-46.52	5	62
		real landfill					
Hydr. Oil shale ash	5-20	leachate	pilot – CW	-46.55	-46.42	5	97
		reconst. fishfarm					
ladle furnace slag	0-1	WW	lab – CW	-44.86	-45.28	6	9
Sas mixed with		reconst. fishfarm					
limestone	5-15	WW	lab – CW	-46.04	-46.01	6	12
Sas mixed with		reconst. fishfarm					
limestone	20-40	WW	lab – CW	-44.48	-44.56	6	8
		reconst. domestic					
Sas	10-30	WW	lab – CW	-40.85	-41.01	7	11
		reconst. domestic					
Sas	30-100	WW	lab – CW	-45.16	-45.26	7	8*
	0.003-		lab –				
BOF oxide mixture	0.1	Solution	column	-49.51	-49.47	8	8*
EAF slag	20-40	real domestic WW	pilot – CW	-44.07	-	9	* +
BOF slag	20-40	real domestic WW	pilot – CW	-43.52	-	9	* +

252 *: all data below pH 10 for these studies. Data was not considered for calculation of global K_{sp}

253 +: single K_{sp} calculated from reported mean values for pH, o-PO₄ and Ca

254 Sas: same as present study; WW: wastewater; CW: constructed wetland

255 Ref #1: Claveau-Mallet et al., 2012. Ref #2: Claveau-Mallet et al., 2013. Ref #3: Forget, 2001. Ref #4 :

256 Koiv et al., 2016. Ref #5 : Koiv et al., 2010. Ref #6 : Abderraja Anjab, 2009. Ref #7 : Stangart, 2012. Ref

257 #8: Baker et al., 1998. Ref #9: Barca et al., 2013.





- 260 In heterogeneous precipitation, crystal size is a variable following crystal growth on existing seeds,
- assuming that all crystals have the same size. The number of seeds (se_{HAP}) increases as homogeneous
- precipitation takes place, assuming an initial number of seeds (se_{HAP_0}) and columnar shape. Equations
- 263 for HAP_HE specific surface and solubility product are provided in equations 2 and 3.

264
$$log(K_{spHAP_HE}) = log(K_{spHAP_bulk}) + \frac{\frac{2}{3}\gamma S_{HAP}}{2.3RT}$$
 [2]

$$265 \qquad S_{HAP} = \frac{(4L_{HAP} + 2)MW_{HAP}}{a_{HAP}\rho_{HAP}}$$
[3]

266 2.4.2 Slag dissolution

Slag composition was simplified to the chemical formula CaO-0.3CaCl₂. Exhaustion equations were determined experimentally (described later), resulting in decreasing functions for saturation pH (pH_{sat}) and dissolution kinetic constant (k_{diss}). k_{diss} was normalized with slag surface as for precipitation constants. The proposed approach gives flexibility to the model and every specific slag has its own exhaustion parameters determined from batch tests by regression.

272 In this model, slag dissolution is assumed to be limited by Fick's law of diffusion (Domenico and

273 Schwartz, 1998) through a crystal barrier that forms uniformly on the slag surface in a thin layer. The

thickness of the crystal barrier (d_{barr}) increased according to CAL, HAP and MON precipitation,

assuming a constant specific surface (S) for slag (equation 4). It was assumed that the type of

- precipitation has an influence on the diffusion coefficient (D_{barr}) and that diffusion is easier in a large
- and organized crystals framework, compared to numerous small crystals. Mathematically, *D*_{barr} was
- 278 defined with a step function set initially at a high value, and to a lower value when the seed
- 279 concentration was doubled. As either dissolution or diffusion rate is the limiting process (the smallest),
- the step function was added to consider the passage from dissolution-limiting to diffusion-limiting.

281
$$d_{barr} = \frac{(m_{CAL} + m_{HAP} + m_{MON}) \times n}{\rho_{barr} S \times 0.001(1-n)}$$
[4]

282 2.4.3 Hydraulic model

In continuous flow column simulations, the Advection-Reaction-Dispersion (ARD) equation for 1D flow
was used (equation 5). A first-order exchange approximation was added to account for diffusion
between effective and immobile porosity (equation 6). The hydraulic model is available in the PHREEQC
software (Parkhurst and Appelo, 1999).

$$287 \qquad \frac{\partial C}{\partial t} = -v \frac{\partial C}{\partial x} + D^* v \frac{\partial^2 C}{\partial x^2} - \frac{\partial q}{\partial t}$$
[5]

$$288 \qquad \frac{db_{im}}{dt} = n_{im} \left(1 + \frac{dq}{dC} \right) \frac{dC_{im}}{dt} = D_n (C_e - C_{im}) \tag{6}$$

289 2.5 Numerical simulations

Numerical simulations were performed using the PHREEQC software with its IPHREEQC modules for
 interfacing with MATLAB (Charlton and Parkhurst, 2011).

292 2.5.1 Batch tests

The initial solution was simulated with KH₂PO₄, K₂HPO₄, CaCl₂ and NaHCO₃ added to pure water. A small 293 294 amount of HCl or NaOH was included to reproduce the precise pH of the experimental solution. Solutions were equilibrated with HAP and CAL (but no MON) in the EQUILIBRIUM_PHASES block prior to 295 296 the simulated batch test. Simulated and experimental alkalinity were used for calibration of initial 297 solutions. Slag exhaustion was considered to be constant, therefore, pH_{sat} and k_{diss} were constant 298 instead of being adjusted according to exhaustion equations. pH_{sat} was set as the maximum pH value 299 reached in the experimental batch test. Calcite precipitation was removed from the batch test model, as 300 surprisingly no calcite precipitation occurred in experimental batch tests (no TIC reduction). 301 The model constants were identified by minimizing the misfit between the simulated output and the experimental pH and oPO₄ measurements. The objective function optimization was performed on the 302 303 log transformed constants k_{diss} , k_{HAP} , k_{MON} , $k_{MONtoHAP}$ and CaO_{in} with the conjugate gradient

method and the golden-section search method (Press, 2007). As shown in Table 2, a two-step strategy was used to achieve satisfactory results and speed-up the calibration process. In the first step, the mean absolute error between the experimental and simulated pH (F₁) was minimized with large tolerances for PHREEQC (1E-11) and the optimization algorithm (1E-7 for line search and 0.01 as F stop criteria). Then, the objective function F_2 was minimized with more demanding tolerances (1E-12 for PHREEQC and line search, and 0.001 as F stop criteria). Note that CaO_{in} was a little amount of CaO instantaneously released at the water/slag contact, added for improving the calibration.

311 Table 2. Batch test inversion parameters for conjugate gradient method

Step	Initial values (log)	Objective function
1	-7 for k_{diss} and CaO_{in}	$F_1 = \frac{1}{2} \sum_{i=1}^{n} \left pH_{i,eyn} - pH_{i,sim} \right $
	-6.5 for k_{MON}	$n \sum_{i=1}^{n} \sum_{j=1}^{n} n_{ij} \sum_{j=1}^{n} n_{i$
	-9 for k_{HAP} and $k_{MONtOHAP}$	
2	Solution from inversion 1	$F_{2} = F_{1} + \frac{0.2}{m} \sum_{1}^{m} \left oPO_{4_{m,exp}} - oPO_{4_{m,sim}} \right $

312

313 2.5.2 Exhaustion functions

Exhaustion functions were produced by plotting pH_{sat} and k_{diss} against total leached CaO ($CaOl_{TOT}$). For a given phase *i*, $CaOl_{TOT}^{i}$ was calculated by cumulating leached CaO in preceding kinetic tests and acid bath (equation 7). $CaOl_{BATH}$ was determined by simulating acid bath with PHREEQC, following the final pH of the acid bath as a target value.

318
$$CaOl_{TOT}^{i} = 0.5 \times CaO_{KTEST}^{i} + \sum_{n=1}^{i-1} (CaO_{KTEST}^{n} + CaO_{BATH}^{n})$$
 [7]

Exhaustion functions coefficients were determined by linear regression of pH_{sat} vs $CaOl_{TOT}$ (equation 8) and logistic function regression of k_{diss} vs $CaOl_{TOT}$ (equation 9). Mean regression coefficients were kept for k_{diss} , but coefficients following the top of the graphical data cloud were kept for pH_{sat} , as 322 pH_{sat} is a saturation state, and we can assume that saturation is controlled by the most reactive 323 particles.

324
$$pH_{sat} = P_2 - \frac{P_2 - P_1}{\left(1 + e^{-P_3(CaOl_{TOT} - P_4)}\right)}$$
 [8]

$$325 k_{diss} = B_1 + B_2 CaOl_{TOT} [9]$$

In the Gujer matrix, exhaustion functions included additional terms involving porosity and slag density to account for the conversion of $CaOl_{TOT}$ (units of mol/g slag) into X_{CaO} (units of mol/L water) for filter numerical simulations.

329 2.5.3 Column test

- 330 The simulated influent was prepared according to the procedure described in the batch tests section.
- 331 The column test was simulated within KINETIC and TRANSPORT blocks, with 50 numerical cells and a
- tolerance of 1E-6. Kinetic rates were applied to both mobile and immobile cells. Hydraulic parameters
- 333 n_e , D^* and D_n were calibrated with each tracer test.

334 3 Results and discussion

- 335 3.1 Determination of exhaustion equations and precipitation constants
- In general, batch test calibration was excellent for pH and good for o-PO₄, except for the period 0 to 100
- 337 min where the model overestimated slightly the o-PO₄ concentration. An example of a well-calibrated
- batch test (rank 7 out of 84 for global error function) is shown in Figure 2. No TIC reduction was
- observed. Absence of CAL precipitation in batch tests was not expected, as calcite was precipitated in
- column tests and is frequently observed in slag filters (Claveau-Mallet et al., 2013; Liira et al., 2009).
- 341 Mean precipitation constants were $k_{HAP} = 10^{-11.03} mol HAP/s m^2 slag$, $k_{MON} =$
- 342 $10^{-8.67} \text{ mol } MON/s \text{ } m^2 \text{ slag}$ and $k_{MONtoHAP} = 10^{-8.01} \text{ mol } HAP2/s \text{ mol } MON$.



Figure 2. Example of a batch test calibration. Simulated data is shown with lines. Batch test error
functions: 0.09 for pH and 0.15 for o-PO₄.

pH_{sat} and k_{diss} obtained from all batch tests were plotted against CaOl_{tot} for the production of exhaustion functions (Figure 3). For k_{diss}, linear regression coefficients were used in column simulations (B₁ = -7.91 and B₂ = -1933 g/mol). pH_{sat} exhaustion function had to be slightly increased above the data cloud (discussed later) to improve the calibration (Figure 3B), resulting in coefficients P₁ = 9.1, P₂ = 12.1, P₃ = 6000 and P₄ = 1.2E-4.



352 **Figure 3.** Exhaustion functions for k_{diss} (A) and pH_{sat} (B). Regression coefficients are provided in text.

353 3.2 Column test calibration

An example of tracer test calibration is shown in appendix. Hydraulic parameters D^* (dispersivity) and D_n (exchange factor between mobile and immobile porosity) were roughly constant for 8 tracer tests, while effective porosity (n_e) decreased slightly following the column operation. The n_e decrease was neglected, and hydraulic parameters from tracer test at time 187 were used ($n_e = 0.359$, $D^* = 5$ cm and $D_n = 5 \times 10^{-6} \text{ s}^{-1}$).

359 Results of numerical simulations are compared to experimental data in Figure 4. pH was correctly 360 predicted for cells 1 to 6 but slightly underestimated for cells 7 to effluent. o-PO₄ was in general 361 successfully predicted, except for cells 1 and 2 in the first 100-200 days, where the o-PO₄ rise was 362 predicted too late. This could be explained by the close position of cells 1 and 2 relative to the influent 363 point resulting in a non plug flow condition for these cells and in some short-circuiting. Calibration of o-364 PO₄ from cells 5 to 7 was less accurate for the last 100 days, as the model predicted a stable 365 concentration while the experiment showed an increase of almost an order of magnitude. Calcium and 366 TIC calibration were less accurate then those for pH and $o-PO_4$, but were considered satisfactory. The 367 effect of the influent rate change was correctly predicted by the model.

Calibrated constants were $k_{CAL} = 10^{-9} \frac{M}{s m^2}$, $SI_c = 0.2$ and $\rho_{barr} = 2000 \frac{kg}{m^3}$. Precipitation constants for HAP, MON and MONtoHAP were already determined in batch tests and were not changed for column simulations. SI_c was lower than reported values for calcite, which occurs in heterogeneous precipitation over $SI_c = 0.3$ and homogeneous precipitation over $SI_c=1.5$ (Mayes et al., 2006). ρ_{barr} value was similar to dry density for a natural sand. k_{CAL} was 2.5 orders of magnitude higher than the reported initial value by Mbamba, Tait et al (2015), and monetite constant was in the same order of magnitude as Mbamba's values. In this study, precipitation rates were function of crystal concentration, with initial crystal seeds of 1E-5 M, while the effect of crystal concentration was not considered in thismodel.

377 The column was divided in two zones for the calibration of $se_{HAP 0}$: cells 1 to 6 were set at 2e21 seeds/L 378 and cells 7 to 9 (and all immobile cells) were set at 5e20 seeds/L. This refinement was necessary to 379 achieve both o-PO₄ calibration of first cells (mainly homogeneous precipitation) and last cells (mainly 380 heterogeneous precipitation). Attributing different se_{HAP 0} values for two zones was considered realistic 381 because it represents a crystal behavior in which at a very low supersaturation index, fewer but bigger 382 crystals are formed. This behavior was confirmed during column dismantling. In cells 8 and 9, a very 383 small amount of precipitates was observed and sampled, with fresh- and unused-looking slag. Several 384 well-formed crystals could be seen by naked eyes only in cells 8 and 9 (3-4 mm in length). 385 The implementation of the diffusion equation is considered a major improvement compared to previous 386 prototype versions of the model. Without the diffusion equation, CaO would always be leached at its 387 maximum capacity (pH_{sat}) and longevity would be highly overestimated. D_{barr} calibration was 1E-10 m²/s at first and was decreased to 5E-16 m²/s when the seed concentration was doubled. Calibrated 388 389 values of D_{barr} for the two steps are similar to diffusion coefficients observed for clays. A large range was reported for radioactive waste storage applications: from 10^{-17} to > 10^{-13} m²/s in consolidated clay 390 (Alonso et al., 2009) and 10^{-11} to 10^{-10} m²/s in altered bentonite (Manjanna et al., 2009). 391



392

Figure 4. Water composition in a column test for cells 1 to 9 (A) and cells 2 to effluent (B). Experimental
data is shown with dots or x and simulated data with lines.

395 3.3 Validation of model hypothesis

396 The diffusion equation can be validated by the robustness of the model regarding pH. The model could

- 397 predict pH at the end of operation, pH after the 6-day rest and pH_{sat} (Figure 5). pH was slightly
- 398 underestimated for pH_{sat} and pH after rest, but relative trends for pH and pH_{sat} were properly predicted,
- as well as the pH increase induced by the 6-day rest. The crystal layer was realistic as uniform crystal

- 400 deposits were observed by SEM (Figure 6). The presence of cracks induced by air-drying was also
- 401 suggesting that a crystal suspension layer was present onto the slag surface. Uniformity of crystal
- 402 composition was confirmed by TEM-EDS as CAL and HAP were frequently found in the same crystal at
- 403 nanometric scale.



- 405 **Figure 5.** pH distribution within column at the end of operation (623 days of feeding followed by 6 days
- 406 of rest).



- 407
- 408 **Figure 6.** SEM picture of slag grain surface from cell 2 at dismantling.
- 409 Cristal size in homogeneous precipitation was confirmed by microscope observations. In cells 1 to 3,
- 410 where the amount of precipitated HAP was sufficient for measurements, isolated crystals in TEM

411 pictures were measured, resulting in a 35 nm mean value for 505 measurements. A similar value of 24 412 nm was calculated from XRD diffractograms of cells 1 to 3. No specific increase was observed in crystal 413 size from cells 1 to 3, validating the constant crystal size hypothesis for homogeneous precipitation. HAP 414 composition was confirmed by XRD patterns (provided in supporting information). HAP composition for 415 individual crystals from TEM was confirmed by P and Ca presence with EDS. No monetite was detected 416 by XRD and no monetite precipitation occurred in the column simulations while monetite was 417 precipitated in batch test simulations, suggesting that pH rise in column was too fast for monetite 418 formation.

The relative small amount of formed HAP in heterogeneous precipitation made impossible XRD or TEM crystal size analysis for heterogeneous HAP. It was possible, however, to analyse the progression of calcite crystal size within the column using XRD. Its size was around 200 nm in cells 1 to 3, increased to 900 nm in cell 5, and was over the limit of Scherrer equation in cells 6 and higher. This suggests homogeneous precipitation and constant size in the first cells and heterogeneous precipitation and crystal growth in last cells.

Even if a distinction between homogeneous and heterogeneous precipitation was made for calcite, its solubility was kept constant for both conditions. Assuming homogeneous precipitation at 200 nm (as measured in cells 1 to 3) and spherical crystals, the computed solubility product from equation 1 is very close to bulk solubility. While HAP growth decreases its solubility, CAL homogeneous crystals are large enough to neglect this effect.

430 3.4 Model limits and recommendations

The main issue regarding the model is the number of batch kinetic tests needed to provide exhaustion
equations. The hypothesis of most reactive grains in batch tests conditioning column tests should also
be examined, because even if pH_{sat} exhaustion equation was overestimating experimental data, pH in

the last cells of columns were underestimated. Work should be done using statistical analysis and theory

435 of artificial granular material sampling (Gy, 1979) for reducing the number of batch tests and

436 transposing correctly batch to column conditions.

Calibration of heterogeneous precipitation was limited by experimental data, as last cells did not reach
their longevity. Additional studies involving long-term operation of slag filters and breakthrough of last
sections of filters, in which heterogeneous precipitation occurred for a while, would be needed for
accurate calibration and in-depth study of seeds concentration and type of precipitation. In this paper,
two layers of different seeds concentration were proposed, but other formulations would be possible
including step functions for seeds VS saturation index, increasing the number of layers or adding a third
type of HAP.

444 Further work should be conducted regarding need for refinement of the model. Additional features such 445 as interaction with atmospheric CO₂ or porosity reduction (Courcelles et al., 2011) may be needed in some cases as constructed wetlands. The model could be improved with additional P species (Mbamba, 446 447 Tait, et al., 2015) or consideration of crystal surface in rate equations. Crystal surface is obviously 448 increasing in this type of process, with long operation time without extraction, but kinetic parameters 449 may be less important than saturation parameters (pH_{sat} and K_{spHAP} HE). Sensitivity analyses would be 450 needed to assess which aspect should be studied further, slag dissolution kinetics, crystal equilibrium 451 parameters or crystal kinetic parameters.

452 4 Conclusion

The first version of the P-Hydroslag model that can be used for prediction of steel slag filter efficiency and longevity was presented in this paper. The objectives were to calibrate the model with experimental data and evaluate the validity and realism of the model. The main outcomes were as follows:

456 A complete model equation matrix was provided. The model included two main kinetic equation 457 sets: a first set for CaO dissolution and a second set for precipitation. CaO dissolution equations 458 included slag exhaustion and CaO diffusion through a uniform crystal barrier on slag particles. 459 Precipitation equations included calcite precipitation, monetite precipitation, transformation of 460 monetite into hydroxyapatite, homogeneous hydroxyapatite precipitation (constant size) and heterogeneous hydroxyapatite precipitation (increasing size). An equation for hydroxyapatite 461 462 solubility related to crystal size was included. Standard equilibrium reactions were included in the model via the PHREEQC software. The advection-diffusion-reaction model was used as the 463 464 hydraulic model (1D porous media transport).

465 - The crystal barrier hypothesis was confirmed by SEM observations of used slag grains.

466 Homogeneous and heterogeneous precipitation hypothesis was confirmed by TEM crystal size467 count and XRD measurements.

468 - The model and proposed experimental procedure for characterization of slag exhaustion
 469 behavior were successful in producing realistic results. Numerical simulations reproduced

470 experimental breakthrough curves (pH, o-PO₄, Ca, TIC) of an upward flow column slag filter.

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- 587 Appendix



Figure A1. XRD pattern of precipitates sampled in cell #2. Main peaks of HAP and CAL diffractograms areindicated in the figure

Phase	Stoichiometry - aqueous							chiomet pha	try – m ases	ineral	Rate equation	Equilibrium					
	<i>H</i> ⁺	0H ⁻	Ca ²⁺	<i>CO</i> ₃ ²⁻	P04 ³⁻	Cl-	HAP	MON	CAL	HAP2		Constants					
											$r_{HAP_HO} = 0.001k_{HAP}S\frac{(1-n)}{n} \times SI_{HAP_HO} \times SF_{HAP_HO}$						
Primary homogenous hydroxyapatite (HAP HO)		-1	-5		-3		+1				$SI_{HAP_{HO}} = log\left(\frac{\{Ca^{2+}\}^{5}\{PO_{4}^{3-}\}^{3}\{OH^{-}\}}{K_{spHAP_{HO}}}\right)$	$pK_{spHAP_HO} = 46$					
											$SF_{HAP_HO} = \frac{1}{1 + e^{-50\left(\log(SI_{HAP_HO}) - \log(SI_c)\right)}}$						
											$r_{HAP_HE} = 0.001k_{HAP}S\frac{(1-n)}{n} \times SI_{HAP_HE} \times SF_{HAP_HE}$						
Primary heterogeneous hydroxyapatite		-1	-5		-3		+1				$SI_{HAP_{HE}} = log\left(\frac{\{Ca^{2+}\}^5\{PO_4^{3-}\}^3\{OH^{-}\}}{K_{spHAP_{HE}}}\right)$	$\log(K_{spHAP_HE}) = \log(K_{spHAP_bu})$ $= \frac{2}{3}\gamma S_{HAP}$					
(nAr_nE)					l											$SF_{HAP_HE} = 1 - \frac{1}{1 + e^{-50(log(SI_{HAP_HO}) - log(SI_c))}}$	+ <u>2.3RT</u>
Monetite (MON)	-1		-1		-1			+1			$r_{MON} = 0.001 k_{MON} S \frac{(1-n)}{n} \times log \left(\frac{\{Ca^{2+}\}\{HPO_4^{2-}\}}{K_{spMON}} \right)$	$pK_{spMON} = 7$					
Calcite (CAL)			-1	-1					+1		$r_{CAL} = 0.001 k_{CAL} S \frac{(1-n)}{n} \times log\left(\frac{\{Ca^{2+}\}\{CO_3^{2-}\}\}}{K_{SpCAL}}\right)$	$pK_{spCAL} = 6.8$					
Secondary hydroxyapatite* (HAP2)		-2	-4		-2					+1	$r_{MONtoHAP} = k_{MONtoHAP} \times SI_{HAP_HO}[MON]$	$pK_{spHAP_HO} = 46$					

*: transformation of monetite into hydroxyapatite

Table A1 (followed). Complete P-Hydroslag model matrix

Slag dissolution	+2	+1.3		+0.			$r_{diss} = A \times SF_{diss}$	N/A
				3			$A = 0.001 k_{diss} S \frac{(1-n)}{n} \left(\frac{pH_{sat} - pH}{pH_{sat}} \right)$	
							$SF_{diss} = 1 - \frac{1}{1 + e^{-50(A-B)}}$	
							$r_{diff} = B \times SF_{diff}$	
CaO diffusion							$B = \frac{0.5 \times D_{barr} \times (10^{pH_{sat} - 14} - \{OH^{-}\})}{d_{barr}} \times \frac{S \times 0.001(1 - n)}{n}$	
through crystal barrier	+2	+1					$SF_{diff} = \frac{1}{1 + e^{-50(A-B)}}$	N/A
							$log(D_{barr}) = -10 - 5.3 \times \frac{1}{1 + e^{-50\left(\frac{se_{HAP}}{se_{HAP}} - 2\right)}}$	



Figure A2. Tracer test experimental data (circles) and numerical calibration (line) (started at Time = 187 d of filter operation)

Detailed equations to complement the model matrix

- 2 Advection-reaction-dispersion (ARD) equation for 1D transport and First-order exchange
- 3 approximation between effective and immobile porosity:

$$4 \qquad \frac{\partial C}{\partial t} = -v \frac{\partial C}{\partial x} + D^* v \frac{\partial^2 C}{\partial x^2} - \frac{\partial q}{\partial t}$$
[A1]

5
$$\frac{db_{im}}{dt} = n_{im} \left(1 + \frac{dq}{dc} \right) \frac{dC_{im}}{dt} = D_n (C_e - C_{im})$$
[A2]

6 Detailed equations for HAP_HE:

7
$$S_{HAP} = \frac{(4L_{HAP}+2)MW_{HAP}}{a_{HAP}\rho_{HAP}}$$
 [A3]

8
$$a_{HAP} = \frac{[HAP_{HE}]MW_{HAP}L_{HAP}^2}{\rho_{HAP}^{3}\sqrt{se_{HAP}}} + \sqrt[3]{se_{HAP}^2}a_{HAP_0}^{3}$$
 [A4]

9
$$se_{HAP} = se_{HAP_0} + \frac{[HAP_HO]MW_{HAP}L_{HAP}^2}{\rho_{HAP}a_{HAP_0}^3}$$
 [A5]

10 Detailed equations for diffusion through crystal barrier:

11
$$d_{barr} = \frac{(100[CAL] + 502[HAP] + 136[MON] + 366[HAP2]) \times n}{\rho_{barr}S(n-1)}$$
 [A6]

12 Detailed equations for exhaustion functions:

13
$$pH_{sat} = P_2 - \frac{P_2 - P_1}{\left(1 + e^{-\frac{P_2}{F}(CaOl_{TOT} - F \times P_4)}\right)}$$
 [A7]

14
$$log(k_{diss}) = B_1 + \frac{1}{F} B_2 X_{CaO} + log\left(0.001S\frac{(1-n)}{n}\right)$$
 [A8]

15
$$F = \frac{1000\rho_{slag}(1-n)}{n}$$
 [A9]

16