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 Tarek Alioua^{1,2}, Boudjemaa Agoudjil², Abderrahim Boudenne^{1*}, Karim Benzarti³
 ¹Univ Paris Est Creteil, CERTES, F-94010 Creteil, France
 ²Université Batna -1, LPEA, Les Allées 19 Mai Route de Biskra Batna, Algeria
 ³Lab Navier, Univ Gustave Eiffel, ENPC, CNRS, F77447 Marne la Vallée, France

10 Abstract

Hygrothermal mathematical models are commonly used to describe heat and moisture transfer 11 12 in porous and bio-based building construction materials. This allows to evaluate their thermal 13 insulation capacity, as well as their ability to regulate external climatic conditions and ensure 14 indoor comfort for inhabitants. In this paper, a sensitivity analysis is performed on Kunzel's 15 model, while it is applied to simulate the hygrothermal behavior of a wall structure made of a 16 new bio-based building material (cement composite including date palm fibers). In a first part, 17 the effects of finite variations of material properties/boundary conditions on the model's 18 outcome are investigated. In a second step, specific transfer modes are neglected in the model, 19 in order to study their influence on the numerical predictions. The results of this parametric 20 study show that special attention should be paid to few parameters (heat capacity and density 21 for heat transfer, sorption isotherm and water vapor resistance factors for moisture transfer) at 22 the expense of others. Uncertainties on these influent parameters may result in large error 23 accumulation, especially when modeling the moisture transfer process. Furthermore, initial 24 boundary conditions and sensors position appear to be possible sources of discrepancies in the 25 calculated RH profiles. Finally, the pure conduction model is found to provide good estimation 26 of the temperature profiles compared to the full model, whereas liquid transfer must always be taken into account in the model to ensure accurate RH predictions through a bio-based date 27 28 palm concrete wall.

Key words: heat and moisture transfer, sensitivity analysis, uncertainties, bio-based building
materials, date palm concrete.

1 *Corresponding authors:

- 2 *Pr. Abderrahim Boudenne, email: boudenne@u-pec.fr
- 3 Université Paris-Est Créteil Val de Marne, (UPEC)/CERTES, 61 Av. du Général de Gaulle

4 94010 Créteil cedex, France.

5

- 6 *Pr. Boudjemaa Agoudjil, email: boudjemaa.agoudjil@univ-batna.dz
- 7 Université Batna 1, 1rue chahid Boukhlouf Mohamed El-hadi 05000 Batna, Algeria.

8

1 Nomenclature

2	A _c	water absorption coefficient (kg $m^{-2} s^{-1/2}$)
3	b	moisture supplement of thermal conductivity (-)
4	C_p	material's heat capacity (J kg ⁻¹ K ⁻¹)
5	D_l	liquid conduction coefficient under relative humidity gradient (kg m ⁻¹ s ⁻¹)
6	D _{l,ws}	capillary transport coefficient for suction process (kg m ⁻¹ s ⁻¹)
7	$D_{l,\varphi}$	liquid conduction coefficient in hygroscopic region (kg m ⁻¹ s ⁻¹)
8	G_{Ω}	vapor flux at the boundary surface (kg m ⁻² s ⁻¹)
9	h	exterior heat transfer coefficient (W m ⁻² K ⁻¹)
10	l_v	latent heat of vaporization (J kg ⁻¹)
11	p_{sat}	vapor saturation pressure (Pa)
12	p_v	vapor pressure (Pa)
13	$Q_{arOmega}$	heat flux at the boundary surface (W m ⁻²)
14	S	relative sensitivity (%)
15	t	time (s)
16	Т	temperature (K)
17	W	water content (kg m ⁻³)
18	W	water content given by sorption isotherm (kg m ⁻³)
19	W_f	water content at free saturation (kg m ⁻³)
20	x	model's parameter
21	Y	model's solution (T or φ)
22	β	water vapor transfer coefficient at boundaries (kg m ⁻² s ⁻¹ Pa ⁻¹)
23	δ	water vapor permeability of the material (kg m ⁻¹ s ⁻¹ Pa ⁻¹)
24	δ_a	water vapor permeability of air (kg m ⁻¹ s ⁻¹ Pa ⁻¹)
25	ξ_{φ}	sorption capacity (kg m ⁻³)
26	λ	thermal conductivity (W m ⁻¹ K ⁻¹)
27	μ	vapor diffusion resistance factor in dry conditions (-)
28	μ^*	vapor diffusion resistance factor at higher humidity (-)
29	ρ	density (kg m ⁻³)
30	arphi	relative humidity (-)
31	Scripts:	

1	0	at dry state or reference value
2	amb	ambient
3	С	convection
4	l	liquid
5	r	radiation
6	ν	vapor
7	Ω	at boundaries
8		
0		

1 **1 Introduction**

Bio-composites based on vegetal fibers are still an extensive field of research in the construction
sector, as they can contribute to reduce energy consumption in buildings thanks to their low
thermal conductivity and high thermal inertia [1]. Besides, these materials have the capacity to
regulate both temperature and relative humidity (RH) of the indoor environment, offering in
most cases an improved hygrothermal comfort for inhabitants [2].

7 So far, most studies dedicated to bio-based building materials were conducted at the sample 8 scale and focused on the determination of various material characteristics, such as mechanical 9 strength or heat/moisture storage and transmissibility properties (*i.e.* thermal conductivity, 10 water vapor permeability, sorption isotherm, etc.). These works considered natural wastes of 11 about 25 different kinds of plants, which were introduced/mixed in conventional construction 12 materials [3]. Examples of well-known fibrous wastes from the biomass and used in this field 13 are hemp [4], straw [5], date palm [6], wood [7] and flax [8]. These studies highlighted the good 14 insulating properties of bio-based materials, as well as their ability to mitigate temperature and 15 RH variations.

16 Since they obtained interesting results at the material level, researchers started to investigate 17 the performance of these bio-based building materials under more realistic conditions at larger 18 scales. Nevertheless, as experimental studies at the wall and building scales require huge 19 equipment and can be very expensive, many researchers focused on the development of 20 theoretical approaches. Since classical heat transfer models have shown limits when applied 21 to bio-based materials [9], this work is focused on models considering coupled heat and mass 22 transfer processes, which were proven effective in the case of building materials, even for 23 those including natural fibers. Mathematical modeling of heat and mass transfer in porous 24 materials was first carried out by by Phillip and De Vries in the case of soil structures [10], 25 and their work served as a solid basis for the development of subsequent models. In 1995, 26 Kunzel [11] developed his own model based on a thorough investigation of heat and mass 27 transport mechanisms in porous building materials. Since then, Kunzel's model has been 28 widely used by researchers in the field of bio-based building materials, and has also been 29 validated by comparison with experimental evidences in several works [12, 13].

Hygrothermal models are commonly used to describe heat and mass transfer in building
envelopes (walls, roofs, floors ...etc.). At the building scale, these models are usually combined
with heat and mass balances of the indoor air environment. They rely on a system consisting at

1 least of two equations that describe heat and moisture transfer processes, and involve two 2 unknowns (i.e., temperature for heat transfer and RH or moisture content for moisture transfer). 3 In specific cases, air transfer is also considered in the problem, and a third equation is thus 4 added to the previous system. The coefficients of these equations are generally defined in 5 relation to actual experimental parameters. Some of these parameters are represented by single 6 constant values, while others may depend on other material properties or on the system's 7 unknowns themselves. The experimental determination of these parameters represents a source 8 of errors, which may strongly influence or introduce bias in the final solution of the system [14, 9 15]. This issue has prompted some researchers to study the sensitivity of the hygrothermal 10 models to variations of the different parameters, in order to improve as much as possible, the 11 numerical estimations of both temperature and RH.

12 Sensitivity analyses are very often adopted in almost all fields of research. The Global 13 Sensitivity Analysis (GSA) is used to determine the sensitivity of model outputs to different 14 inputs variations, and sensitivity indices are generally determined with the Monte Carlo method 15 [16]. Differently, the Local Sensitivity Analysis (LSA) is used to assess the sensitivity of a 16 model with respect to a given parameter [17]. In the field of hygroscopic building materials, the 17 simulations are simply rerun for different parameter values/conditions, and the resulting new 18 profiles are compared with the reference case. At the wall scale, Mendes et al. [18] carried out 19 a sensitivity analysis on their model which is based on Phillip and De Vries theory. They defined 20 different simplified sub-models in which they neglected or set constant specific transport 21 coefficients, in order to evaluate their influence on the numerical results. The authors 22 highlighted that neglecting moisture transfer in the model may lead to an underestimation of up 23 to 59% of the yearly integrated heat flux, hence resulting in an underestimation of the overall 24 energy consumption. Another work conducted by Bart et al. [14] studied the sensitivity of 25 Kunzel's model to variations of the material density, thermal conductivity, heat capacity, vapor 26 diffusion resistance factor, sorption curve and sensors position for the case of a hemp concrete 27 wall. Results showed that theoretical RH profiles are mostly influenced by two parameters, *i.e.*, 28 the sorption isotherm and the vapor resistance factor. Kunzel's model was also studied by 29 Oumeziane et al. [19], who investigated the effects of the material density and the initial 30 conditions on the numerical results of a hemp concrete wall. It was concluded from this work, 31 that these parameters influence both the theoretical temperature and RH profiles, which was 32 explained by the coupling effects of heat and moisture transfer processes. Othmen et al. [15] 33 carried out a sensitivity analysis on Kunzel's model at the wall scale as well using the LSA

1 method. These authors studied the effect of several material properties, as well as the influence 2 of boundary and initial conditions in the case of two walls made of tuffeau and hemp concrete. 3 It was concluded from this analysis, that convective heat transfer coefficients influence both 4 temperature and water content profiles, in addition to the adsorption isotherm that affects 5 directly the water content distribution in the wall. At the building scale, Le et al. [2] performed 6 another sensitivity analysis on Mendes model. They studied the impact of changes in the 7 transport coefficients, sorption isotherm, ventilation rate and habitant's existence, on both 8 indoor conditions and energy consumption. Once again, considering moisture transfer in the 9 calculations had a significant effect on energy consumption, and this latter was also highly 10 influenced by the ventilation rate.

11 Each of these works studied the sensitivity of hygrothermal models to various parameters or 12 conditions, by considering a certain percentage of variation of these parameters/conditions with 13 respect to their reference values. Moreover, some properties were considered constant, such as 14 the thermal conductivity and sorption capacity. On the other hand, it was noticed that the results 15 of the sensitivity analyses were sometimes different or contradictory if the models were applied 16 to different materials under some specific experimental conditions [15]. This could be explained 17 by the changes in material properties and conditions which may affect the kinetics of 18 heat/moisture transfer. As a consequence, conclusions drawn from a sensitivity analysis should 19 be restricted to the material under study.

20 In the present work, a comprehensive sensitivity analysis is carried out on Kunzel's 21 hygrothermal model, while this latter is applied to Date Palm Concrete (DPC). To our very 22 best knowledge, it is the first time this analysis is performed for this peculiar building material 23 at the wall scale. The study is based on previous results obtained from an experimental 24 campaign on a DPC at material and wall scales. DPC properties measured at material scale 25 are used as reference input parameters in the model, while temperature and RH profiles 26 collected from the DPC wall were used to validate the model's outcome that will be used in 27 the sensitivity analysis as the reference solution. In this approach, the influence of a wide 28 range of parameters (heat capacity C_p , exterior heat transfer coefficient h, thermal conductivity λ , vapor diffusion resistance factors μ , density ρ and sorption isotherm W), 29 30 conditions and variation percentages are explored. A local sensitivity analysis (LSA) method 31 is used to capture separately the influence of each parameter on the model outcome, and the 32 global influence of the whole set of parameters is then evaluated as well. The effects of initial/ boundary conditions and sensors

positions on the values of temperature/RH calculated at different depths of the DPC wall are also investigated. Finally, some terms of the hygrothermal model related to specific heat/moisture transfer modes were neglected in order to investigate the possibility of simplifying the model without affecting its accuracy.

5

6 2 Materials and methods

7 2.1 DPC properties

8 The material considered in this study is a new bio-based composite incorporating 15 wt. % date 9 palm fibers in a classical mortar (cement + sand + water), and commonly referred to as Date 10 Palm Concrete (DPC). This composite formulation was designed in previous studies [20-22], and was found to provide a good compromise between mechanical, thermo-physical and hygric 11 12 properties. This DPC material offers both good thermal insulation/moisture buffering capacity, and also meets RILEM requirements for structural and insulating autoclaved aerated concretes 13 (Rc > 2.5 MPa, $\lambda < 0.75$ W m⁻¹ K⁻¹) [21]. The main hygrothermal properties of this new DPC 14 bio-composite were experimentally measured in our previous works and are reported in Table 15 1; these data will serve as reference inputs of the mathematical Kunzel's model in the 16 17 framework of the sensitivity analysis.

18 The following sub-sections provide a brief description of the experimental methods used to 19 determine these hygrothermal properties at the material scale.

20 2.1.1 Vapor resistance factors

Water vapor diffusion resistance factors μ and μ^* characterize the ability of the material to oppose vapor transfer at different ranges of relative humidity. In the case of DPC, typical values were obtained using the dry and wet cup test method (EN ISO 12572 [23]).

24 2.1.2 Capillary suction and free saturation

Capillary suction is a liquid transfer mode which occurs mainly in the capillary region (95% > RH > 100%). This process can be taken into account in the Kunzel's model through a water absorption coefficient (A_c), whose value can be determined experimentally by suction process (EN ISO 15148 [24]). Free saturation defines the maximum liquid water storage capacity of a material after a total immersion, and is noted W_f .

1 2.1.3 Thermal conductivity, heat capacity and dry density

2 Thermal conductivity (λ) is an important property with regard to the insulation capacity of
3 building materials, as it defines the ability of the medium to transmit a heat flux by conduction.
4 In this work, it is taken dependent on the moisture content (w) according to [11]:

5

$$\lambda(w) = \lambda_0 (1 + b.\frac{w}{\rho_0}) \tag{1}$$

6 Were λ_0 is the thermal conductivity at dry state and *b* is the moisture supplement of thermal 7 conductivity. In the case of DPC, values of these coefficients were calculated from thermal 8 conductivity measurements. DPC density at dry state (ρ_0) was also determined.

9 Heat capacity C_p defines the amount of heat needed to change a material's temperature. It was

10 determined experimentally for DPC.

11 Values collected for all these coefficients/properties in the case of DPC material are reported 12 in Table 1, along with the references from which they are taken. They will be considered as 13 reference values in the sensitivity analysis.

14 2.1.4 Sorption isotherm

15 Sorption isotherm represents the water retention evolution inside a material, during adsorption

16 or desorption processes, as a function of relative humidity at a known temperature. For DPC,

17 adsorption isotherm *W* was obtained experimentally (EN ISO 12571 [24]) by Chennouf *et al.*

18 [22] at 23 °C and fitted with the GAB model (Guggenheim-Anderson-de Boer [26]).

19 Table 1: Reference values of the DPC	properties
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Property	Value	Reference	Property	Value	Reference
Dry density (ρ_0) [kg.m ⁻³]	954	[20]	Dry specific heat (C_{p_0}) [J.kg ⁻¹ K ⁻¹]	1500	[21]
Dry thermal conductivity (λ_0) [W.m ⁻¹ K ⁻¹]	0.185	[20]	Moisture supplement of thermal conductivity (<i>b</i>) [-]	10.190	[20]
Vapor resistance factor (dry cup) (µ) [-]	6.310	[22]	Vapor resistance factor (wet cup) (μ [*]) [-]	5.570	[22]

Water absorption			Water content at free		
coefficient (A_c)	0.165	[20]	saturation (W_f)	429	[20]
$[kg.m^{-2}s^{-0.5}]$			[kg.m ⁻³]		

2 2.2 *Experimental test at the wall scale*

3 For more realistic tests, boundary conditions to be used in the sensitivity analysis were imported from an actual experimental campaign. A DPC wall of dimensions 50×40×15cm³ (Fig. 1a) was 4 5 fabricated and then insulated from lateral surfaces using polystyrene panels and plastic film in 6 order prevent heat/moisture transfer through these surfaces and get a one-dimensional transfer 7 across the thickness of the wall. Polystyrene panels were used to create a passive cell at the 8 right side of the DPC wall, which represents the indoor environment (Fig. 1b). Three 9 humidity/temperature sensors model DKRF400 from Driesen + Kern, Germany (measurement 10 range of RH: 0-100% with an accuracy of ± 1.8 %, and measurement range of T: -40°C to +45°C 11 with an accuracy of $\pm 0.5^{\circ}$ C) were inserted at three different depths of the wall for data 12 acquisition (at 3 cm, 7.5 cm and 12.5 cm from the outer surface). The reason behind choosing 13 these specific depths was to enable measurements of temperature and RH close to the inner / 14 outer environments and at the middle of the wall as well, in order to evaluate the response of 15 the material at these locations. The overall setup was placed in a climatic chamber that 16 represents the outdoor environment and enables to apply dynamic boundary conditions to the 17 external side of the wall (Fig. 1.b). Two additional sensors were used to monitor both 18 temperature and RH values in the indoor and outdoor environments.

Hygrothermal scenarios applied to the DPC wall using the regulation system of the climaticchamber are described as follows:

1) An initial step change of temperature from 23°C to 40°C, followed by 4 repeated temperature
cycles of 24 hours (hot/cold periods of 12 hours with step changes of temperature between 40°C
and 18°C) at constant humidity level of 50% RH. This scenario simulates the daily temperature
variation during summertime,

- 25 2) An initial step change of RH from 23% to 75% followed by a prolonged exposure to 75%
 26 RH for 9 days at constant temperature (23°C). Afterwards, another step RH change was applied
 27 from 75% to 33%, and this latter level was maintained for another 9 days, still at constant
- 28 temperature of 23°C. The total duration of this scenario is 18 days.

1 These scenarios were applied as outdoor conditions using the climatic chamber, while indoor 2 conditions were left uncontrolled. After every test, the wall was reconditioned at relative 3 humidity $\varphi = 50\%$ and T = 23 °C which represent the initial conditions for the next test.

4 Details of the used scenarios are displayed in Table 2. Additional details regarding the 5 experimental setup and the protocols can be found in [27, 28].

6

- 7 **Table 2:** Outdoor conditions (values in brackets correspond to the initial conditions at the beginning
- 8 of the experiment, *i.e.* 23 °C and 50% HR)

Soonamiog	Outdoor	conditions	Duration
Scenarios	T [°C]	φ [%]	
Seconario 1	(23) →40 →18	50	4 cycles of 24 hours (12 h at 40 °C,
Scenario 1		50	12 h at 18 °C)
Seconario 2	23	(50) 75 22	18 days (9 days at 75% RH, 9 days at
Scenario 2		(30) 775 753	33% RH)

9



10

Fig. 1: The DPC wall (a) and schematic view of the experimental setup (b)

1 **3 Mathematical formulation**

2 3.1 Mathematical model

All mathematical models describing heat and mass transfer in porous material are derived from
basic conservation equations. In porous media, some hypotheses are usually applied in order to
simplify the models [29], such as:

- 6 The studied material is considered isotropic,
- 7 The local thermodynamic equilibrium is reached between the different phases,
- 8 Air and vapor are considered as ideal gases,
- 9 Radiative heat transfer and chemical reactions are neglected inside the studied material,

10 The mathematical model describing heat and moisture transfer in porous building materials and 11 used in this study refers to Kunzel [11]. Based on the previous assumptions, the model is written 12 as follows:

- 13 Moisture transfer
- 14

$$\xi_{\varphi} \frac{d\varphi}{dt} = \frac{\partial}{\partial x} \left(\delta \frac{\partial (p_{sat}\varphi)}{\partial x} + D_l \frac{\partial \varphi}{\partial x} \right)$$
(2)

15 Heat transfer

16
$$\left(\rho_0 C_{p_0} + C_{p_l} w\right) \frac{dT}{dt} = \frac{\partial}{\partial x} \left(\lambda \frac{\partial T}{\partial x}\right) + l_v \frac{\partial}{\partial x} \left(\delta \frac{\partial p_{sat} \varphi}{\partial x}\right)$$
(3)

17 The transfer modes considered in this model are: diffusion of liquid water and vapor in the case 18 of mass transfer, heat conduction and phase change heat for heat transfer. This model neglects 19 air transfer because of the small variations of the total pressure in the case of building 20 components [11].

The liquid water transport takes place according to several processes that are expressed in the model by the following coefficients [11]:

a) The liquid conduction coefficient (acts mainly in the hygroscopic region), defined asfollows:

25
$$\boldsymbol{D}_{l,\varphi} = \boldsymbol{P}_{sat} \boldsymbol{\delta}_{a} \left(\frac{1}{\boldsymbol{\mu}^{*}(\varphi)} - \frac{1}{\boldsymbol{\mu}} \right)$$
(4)

26 Where μ and $\mu^*(\varphi)$ are vapor resistance factors obtained from dry and wet cup tests 27 respectively. 1 b) The capillary transport coefficient for suction process (acts mainly in capillary region):

- 2 $\boldsymbol{D}_{l,ws} = \left(\boldsymbol{3}.\boldsymbol{8} \left(\frac{A_c}{W_f}\right)^2 \cdot \boldsymbol{1000}^{\frac{w}{wf}-1}\right) \boldsymbol{\xi}_{\boldsymbol{\varphi}}$ (5)
- 3 A_c and W_f are measured experimentally (cf. section 2.1).

4 c) Liquid transport coefficient of redistribution; it is estimated as one decimal power bellow

5 the capillary transport coefficient $D_{l,ws}$.

6 3.1.1 Boundary conditions

7 Experimental boundary conditions were defined and simulated according to Kunzel's work8 [11]. Heat and mass fluxes are equal to:

- 9 $\boldsymbol{Q}_{\boldsymbol{\varOmega}} = \boldsymbol{h}(\boldsymbol{T}_{\boldsymbol{\varOmega}} \boldsymbol{T}_{amb}) \tag{6}$
- 10 $\boldsymbol{G}_{\boldsymbol{\varrho}} = \boldsymbol{\beta} (\boldsymbol{P}_{\boldsymbol{\nu},\boldsymbol{\varrho}} \boldsymbol{P}_{\boldsymbol{\nu},\boldsymbol{amb}})$

11 *h* takes into account both convection and radiation through the following relation:

 $h = h_r + h_c \tag{8}$

13 And β is the water vapor transfer coefficient at boundaries, defined as:

14

$$\boldsymbol{\beta} = 7.\,\mathbf{10^{-9}}.\,\boldsymbol{h}_c \tag{9}$$

(7)

15 3.2 Model validation

16 Kunzel's model was widely used and compared with experimental data in previous studies from 17 the literature. The Fraunhofer Institute for Building Physics has developed a software (Wufi) to 18 describe the hygrothermal behavior of building envelops based on Kunzel's, which was 19 validated by comparison with experimental evidences. A good agreement was reported between 20 theoretical/experimental data for both temperature and RH profiles [13].



Fig. 2: Kunzel's model validation: comparison of experimental and numerical profiles at 3 cm depth inside the DPC wall, for temperature (a) and relative humidity (b)

3

4 Recently, we carried out a detailed validation of Kunzel's model using COMSOL Multiphysics 5 software in the case of a DPC wall, using the same setup and scenarios described in section 2.2 [12]. As an illustration, theoretical and experimental profiles obtained at a depth of 3cm in the 6 7 DPC wall for temperature (scenario 1) and RH variations (scenario 2), are presented in Fig. 2.a 8 and 2.b, respectively. A relatively good agreement was obtained between Kunzel's theory and 9 experimental data in both cases. Nevertheless, slight differences were noticed between 10 numerical/experimental curves, which could relate to various sources of errors: uncertainties 11 on test conditions, accuracy/position of the sensors or values of input parameters that were 12 injected into the model.

13

14 **4** Assumptions and sensitivity analysis

The goal of this work is to assess the influence of a wide range parameters and conditions on the response of the hygrothermal Kunzel's model. In this line, the impact of finite variations of several material parameters on the numerical results was evaluated (considering different percentages of variations of these inputs with respect to their reference values), as well as the effects of time, depth location inside the wall, initial conditions, sensors position, liquid transfer, phase change heat, etc. This is done for the case of a date palm concrete wall.

21 4.1 Uncertainties on the supposed known model parameters

22 The first part of the study deals with the influence of input material parameters on the model's 23 outcome. Significant errors are usually present on these parameters, due to measurement 24 uncertainties at the material scale, or to the use of these parameters at higher scales as mentioned 25 earlier (wall and building scales). This latter reason (transition from the material to the wall 26 scale) may be responsible for much higher uncertainties on input values, because manufacturing 27 several blocks or a sample of large dimensions can probably generate a wider range of 28 uncertainties on the materials properties and on the homogeneity of the system [19]. As 29 mentioned before, different sensitivity analysis methods are available depending on the number 30 of variables set to change together, such as the Global Sensitivity Analysis (GSA) and the Local 31 Sensitivity Analysis (LSA). The GSA consists in determining the sensitivity of output quantities to the variations of different input parameters, while the LSA evaluates the sensitivity of an 32

output quantity to the variation of a given input. In the field of hygroscopic building materials,
we are mainly interested in local methods where the simulations are simply rerun for different
parameters values and conditions, and the new profiles are compared with the reference case.
In the present work, to evaluate the influence of a varying parameter on the resulting
temperature and RH profiles, the relative sensitivity index is calculated as follows [30]:

$$S(\%) = \frac{|Y_{x-\Delta x} - Y_{x+\Delta x}|}{2Y_0} . 100$$
(10)

7 Where Y is the solution of the model (T or φ), Y_0 is the reference solution obtained when no 8 parameter is changed (all parameters are set to their reference values), x indicates the variable 9 parameter and Δx the variation range.

This relative sensitivity index is calculated considering different percentages of variation of input parameters with respect to their reference values, ranging from 5% to 25% [2, 14, 19]. Low percentages represent errors that may occur at the material scale, while high percentages are related to errors at higher scales due to the accumulation of small errors. The evolution of the sensitivity index will be discussed as a function of: time, depth location and percentage of variation of the parameters compared to their reference values.

16 The sensitivity of Kunzel's model has been investigated with respect to the following 17 parameters: heat capacity C_p , exterior heat transfer coefficient *h*, thermal conductivity λ , vapor 18 diffusion resistance factors μ , density ρ and sorption isotherm *W* that is used to define the water 19 content *w* in the hygrothermal model (Eq. 3) under the thermodynamic equilibrium hypothesis.

20 In a first step, each parameter was changed separately considering different percentages of 21 variation (Δx : -5%, +5%, -15%, +15%, -25% and +25% of the reference value), while the other 22 parameters were kept constant at their initial values. In a second stage, all the parameters were 23 changed simultaneously using the same percentages of variation, and the sensitivity index was 24 calculated for the variation of all these parameters together in order to reveal possible 25 interactions between parameters. For example, for this last case, all parameters were varied with 26 -5% relating to their initial value, to obtain $Y_{x-\Delta x}$ that is defined in Eq. 10. Then, all parameters were also varied with +5% to obtain $Y_{x+\Delta x}$, Finally, these two last results allow to obtain the 27 28 corresponding sensitivity index S. This procedure was performed as well for the rest of 29 percentages variation $\pm 15\%$ and $\pm 25\%$.

The different configurations of this sensitivity analysis are summarized in Table 3 (including the reference case). Calculations were carried out for the two scenarios previously described in section 2.2 (cycles of temperature and RH, respectively). 43 different simulations were run for each scenario, representing a total of 86 simulations for the overall study.

5 Some of these parameters represent independent coefficients (C_p , h, μ and ρ) while other parameters are defined as functions (λ defined by Eq. 1 and W defined by the GAB model) and 6 7 depend on other parameters or/and on the system's unknowns (temperature and relative 8 humidity RH). It is worth to mention that changing some parameters may imply the change of others, *i.e.* changing the adsorption isotherm W implies a systematical recalculation of ξ_{φ} , the 9 sorption capacity that represents the derivative of the sorption isotherm. Moreover, the vapor 10 resistance factors obtained from wet and dry cup tests (μ^* and μ), are both changed 11 12 simultaneously with the same percentages and are used to calculate liquid conduction coefficient (D_l) according to Eq. (4). In the same way, varying the heat exchange coefficient h 13 14 with a certain percentage leads to a variation with the same percentage of the convective heat transfer coefficient h_c , which is used to calculate the water vapor transfer coefficient at 15 16 boundaries β (see Eq. 9).

17 Besides, real boundary conditions imported from the actual experiment will be introduced in 18 the model (instead of creating uniform functions numerically), in order to perform the 19 sensitivity analysis under more realistic conditions. This choice will also facilitate future 20 comparisons between numerical simulations and experimental measurements.

21	Table 3: Study cases of the sensitivity analysis (for each case, 6 different percentages of
22	variations and two scenarios were considered)

Case	Varying parameters	Variation percentage	Unchanged parameters (set to their reference values indicated in Table 1)
Reference case			$C_p, h, \lambda, \mu, \rho, W$
Case 1	C_p	±5%, ±15%, ±25%	h, λ, μ, ρ, W
Case 2	h	±5%, ±15%, ±25%	C _p , λ, μ, ρ, W
Case 3	λ	±5%, ±15%, ±25%	C_p, h, μ, ρ, W
Case 4	μ	±5%, ±15%, ±25%	C_p, h, λ, ρ, W
Case 5	ρ	±5%, ±15%, ±25%	C_p, h, λ, μ, W

Case 6	W	±5%, ±15%, ±25%	$\mathcal{C}_p,\ h,\ \lambda,\ \mu,\ ho$
Case 7 (simultaneous variation)	C _p , h, λ, μ, ρ, W	±5%, ±15%, ±25%	

2 4.2 Effect of initial conditions and sensors position

Another common possible source of errors in numerical calculations is related to the initial conditions. In the present case, initial values of temperature and RH are imported from the experiment. Consequently, the sensors accuracy may be responsible for some uncertainties on the measured values. The hygrothermal sensors (model DKRF400 from Driesen + Kern, Germany) present an accuracy of ± 0.5 °C on temperature and $\pm 1.8\%$ on RH. It should be noted that these sensors have a higher accuracy compared those usually used in other works.

9 Besides, insertion of the hygrothermal sensors into the DPC wall requires drillings holes in 10 view of positioning the sensor's probe at a given depth. This is a rather tough operation, which 11 may result in significant errors on the sensor position. In order to evaluate the bias caused on 12 theoretical temperature and RH profiles, the effect of a deviation of the sensor location (up to 13 \pm 1 cm away from its reference position) on numerical results was investigated.

14 4.3 Simplification of heat and moisture transfer models

15 In the next part of the work, the possibility of simplifying the description of the transfer 16 processes in the model was also investigated. In this line, some terms of the model related to 17 specific transfer modes were neglected, and the effects of these modifications on the numerical 18 outcomes were then evaluated by comparison with the reference case.

Such an investigation was already proposed in the literature [31]. Heat transfer was modeled by taking into account the conduction process only, and this simplified model proved to give good results in terms of temperature predictions.

In this part of the study, we explored the same hypothesis by neglecting the phase change term in the heat equation of the hygrothermal model (Eq. 3) and keeping only the heat conduction term. This leads to the following equation for heat transfer:

25
$$\left(\rho_0 C_{p_0} + C_{p_l} w\right) \frac{dT}{dt} = \frac{\partial}{\partial x} \left(\lambda \frac{\partial T}{\partial x}\right)$$
(11)

26 This new model was solved for scenario 1, and the results obtained at different depth of

the DPC wall were then compared to reference solutions provided by the complete model.

Likewise, as far as porous building materials are concerned, huge importance is usually given to vapor transfer at the expense of liquid water transfer while modelling transfer phenomena and RH variations. In order to estimate the contribution of liquid water diffusion to the overall moisture transfer process, this specific mode was neglected in the model by setting to zero the liquid transfer coefficients (cf. section 3.1) of the mass transfer equation (Eq. 4), hence leading to the following expression:

$$\xi_{\varphi} \frac{d\varphi}{dt} = \frac{\partial}{\partial x} \left(\delta \frac{\partial (p_{sat}\varphi)}{\partial x} \right) \tag{12}$$

8 This simplified model, based on vapor diffusion only, was solved using conditions of scenario 9 2, and numerical results obtained at various depth of the DPC wall were again compared to 10 those provided by the full model.

11 4.4 Boundary conditions

To resolve the problem, Kunzel's model was implemented in COMSOL Multiphysics, a simulation software and partial differential equations solver based on the finite element (FE) method [32]. The wall presented in section 2.2 was simulated and discretized following the same procedure as in [12]. A schematic view of the simulated wall and boundary conditions is presented in Fig. 3.



17

Fig. 3: Model of the simulated wall

18 The global heat transfer coefficients at boundaries were estimated as follows, according to

19 Kunzel's approximations [11]:

20 - 8 W.m⁻².K⁻¹, with a convective contribution of 3.57 W.m⁻².K⁻¹ for the interior surface,

- 17 W.m⁻².K⁻¹, with a convective contribution of 10 W.m⁻².K⁻¹ for the exterior surface.

1

2

5 Results and discussion

3 5.1 Uncertainties on supposed known model parameters

4 5.1.1 Reference solutions

5 Reference solutions obtained at various depth of the DPC wall for scenarios 1 and 2 (heat and 6 moisture transfers) are respectively presented in Figs. 4 and 5. These results were obtained by 7 introducing into the model the reference values of input parameters (i.e., measured material 8 properties or estimated value in the case of the heat transfer coefficient h). These temperature 9 and RH profiles will be used as a reference curves for the sensitivity analysis.



Fig. 4: Reference solutions of the temperature profiles at various depths of the DPC wall subjected to scenario 1 (origin = outdoor side of the wall): 3 cm (a), 7.5 cm (b), 12.5 cm (c).



Fig. 5: Reference solutions of RH profiles at different depths of the DPC wall subjected to
 scenario 2 (origin at the outdoor side of the wall): 3 cm (a), 7.5 cm (b) and 12.5 cm (c).

4 5.1.2 Sensitivity as a function of time

5 In the current study, different factors may influence simultaneously the model's sensitivity for 6 the different input parameters: the percentage of variation of input parameters, time and depth 7 location. Therefore, we propose a methodology in which two of these factors are fixed, so that 8 the sensitivity evolution for the third factor can be assessed. In a first stage, the effect of time 9 evolution on the model's sensitivity to different input parameters is investigated for each 10 scenario. At first the effect of all parameters over time is displayed for a 25% variation percentage at 3 cm depth (the effect at other depths is also presented below). This percentage 11 12 was chosen in order to show the extreme case of errors.

13 Heat transfer (scenario 1)

Fig. 6 shows the evolution over time of the relative sensitivity of the model to the various input
parameters, at a depth of 3 cm in the DPC wall and considering 25% variation on parameters
with respect to their reference values.



Fig. 6: Evolution over time of the relative sensitivity of the model to various input parameters,
 at 3 cm depth in the DPC wall and considering 25% of variation on parameters with respect to
 their reference values (case of scenario 1)

15 It can be observed that the various parameters do not have the same influence on the evolution 16 curves of the temperature relative sensitivity. The most influencing parameters are respectively 17 C_p and ρ , followed by h and λ , and finally μ . Likewise, it seems obvious that the sorption 18 isotherm W and the water vapor resistance factor μ have a negligible impact in this case.

19 Besides, the curve obtained for the relative sensitivity to h shows a very different trend (and even a reverse evolution) compared to the other curves: when the relative sensitivity to h20 21 increases, the sensitivities to the other parameters are found to decrease, and vice versa. Moreover, sensitivities to C_p and ρ have almost the same evolutions. The same results were 22 obtained by Le et al. [33] for a hemp concrete wall and the same variation percentage (25%). 23 24 These two parameters are usually linked together but we plotted each of them separately 25 because their experimental values can be measured separately, and thus measurement errors 26 can be made on both properties.

Fig. 7 displays the evolution versus time of the relative sensitivity to C_p (this parameter has previously been identified as the most influencing parameter on heat transfer). Evolution curves are plotted for different depths of the DPC wall, at 3 cm, 7.5 cm and 12.5cm away from the
 outdoor side of the wall. It can be seen that the sensitivity index goes up and down
 simultaneously at all depths.



Fig. 7: Evolution versus time of the temperature relative sensitivity to a change of 25% in the heat capacity (C_p) with respect to the reference value, and at different depths of the DPC wall: 3 cm (a), 7.5 cm (b) and 12.5 cm (c)

22

Fig. 8 shows the evolutions of the new temperature profiles at different depths, which were
calculated considering variations of ± 25% on C_p with respect to the reference value
(reference curves are displayed on this graph as well). We can observe that the wall reacts in
the same way as external variations but with different amplitudes decreasing as a function of
depth. This is

due to the fact that we are moving away from the external surface where the outdoor conditions (scenarios) are applied. By comparing Figs. 7 and 8, it is found that maximum values of sensitivity are reached at the same time as the outdoor temperature reaches its minimum during the imposed temperature cycle. Indeed, the quantity $|Y_{x-\Delta x} - Y_{x+\Delta x}|$ is maximal at this point, and when divided by the minimum reference temperature, it gives maximum values of the relative sensitivity (Eq. 9). This was the case for all depths.

We can also notice that the sensitivity curves (Fig. 7) are smoother at 7.5 cm and 12.5 cm depths,
compared to 3 cm. This is due to temperature reference and new profiles which show also
smoother curves at 7.5 cm and 12.5 cm depths. Note that the first cycle was ignored in the result
analysis to avoid any possible effect of initial conditions.



Fig. 8: New temperature profiles simulated at different depths of the DPC wall, considering variations of the heat capacity by +25% (C_{p+}) or -25% (C_{p-}) compared to the reference value

3

4 Moisture transfer (scenario 2)

5 Regarding moisture transfer (scenario 2), the most influencing parameters are the sorption 6 isotherm W and vapor resistance factors (μ) whereas the other parameters have negligible 7 effects, as shown in Fig. 9.



Fig. 9: Evolution over time of the relative sensitivity of the model to various input
 parameters, at 3 cm depth in the DPC wall and considering 25% of variation on
 parameters with respect to their reference values (for scenario 2).

20 Fig. 10 depicts the new RH profiles simulated considering variations of +25% or -25% on these two influent parameters, and at different depths of the wall. It was noted that lower values of 21 22 W and μ lead to higher RH curves during adsorption phase (*i.e.* max RH differences during 23 adsorption phase for a -25% sorption isotherm W read: 0.63%, 1.29% and 0.85% at 3 cm, 7.5 24 cm and 12.5 cm depth respectively) and lower curves during desorption phase (i.e. max RH 25 differences during desorption phase for a -25% sorption isotherm W read: 1.28%, 2.31% and 26 1.39% at 3 cm, 7.5 cm and 12.5 cm depth respectively). A lower sorption isotherm corresponds 27 to lower moisture storage in the material for the same RH value, this means that at a given water 28 content, we get a higher RH with the new sorption isotherm. Thus, a decrease in water content

resulted in slowing down the moisture transfer and storage in the material, as maximum and
minimum levels of RH were obtained with a time delay on the new profiles.

In the same way, a reduction of the vapor resistance factor μ implies more vapor diffusion in
the material, hence higher RH levels during adsorption phase and lower ones during desorption.
The same observations and conclusions were also reported by [14, 34].



Fig. 10: New RH profiles simulated at different depths of the DPC wall, considering $\pm 25\%$ variations on (a) the sorption isotherm W and on (b) the vapor resistance factor μ compared to their reference values

17

18 Fig. 11 represents the evolution over time of the relative sensitivity of the model to a 25% 19 variation on the sorption isotherm W, which was previously identified as the most influencing 20 parameter on moisture transfer. Numerical curves obtained at different depths of the DPC wall 21 are displayed. Larger variations can be observed at 3 cm depth compared to 7.5 cm and 12.5 22 cm depths (this is very similar to what was observed for the temperature relative sensitivity 23 obtained in the case scenario 1). The relative sensitivity reaches minimum values (zero) at the 24 points of the curves reversing shown on Fig. 10 (the points where the higher curve becomes lower and the lower one becomes higher). This is because the new curves intersect and $|Y_{x-\Delta x} - X_{x-\Delta x}|$ 25 26 $Y_{x+\Delta x}$ is equal to zero. Almost at all depths the highest relative sensitivity values were obtained 27 at the end of the test. At these points, $|Y_{x-\Delta x} - Y_{x+\Delta x}|$ is maximum and RH_0 is minimum.



time, and considering different variation percentages of the model inputs. Simulations were carried out at different depths of the DPC wall. The time instants were chosen according to the analysis reported in section 5.1.2. The principle was to choose instants where the relative sensitivity reaches maximum values, so that the maximal influences of the various parameters can be compared. Different time instants were selected, depending on the scenario and the depth position considered.

1 Heat transfer (scenario 1)

2 As already mentioned, the first cycle was not considered in this analysis. The instants chosen 3 for the comparison are t_1 =268920s, t_2 =269760s and t_3 =267120s for 3 cm, 7.5 cm and 12.5 4 cm depth respectively. They represent the instants where the sensitivity to C_p , the most influencing parameter on heat transfer reach a maximum value. At these times and using the 5 6 relative sensitivity expression (Eq. 9), the evolution of the sensitivity as a function of the variation percentage of each parameter was plotted in Fig.12. This figure displays the graphs 7 8 determined at different depths of the DPC wall. In agreement with previous results of section 5.1.2, the most influencing parameters on the heat transfer were C_p and ρ , then h and λ , and 9 finally μ and W. A similar ranking was observed at all depths. Similar results were obtained by 10 Van Belleghem et al. [34], who studied the sensitivity of heat and mass transfer to moderate 11 changes (5%) in initial material properties, in the case of in gypsum boards. It can be noted that 12 13 the sensitivity of the model to the external heat transfer coefficient (h), decreases as a function 14 of depth. This result was expected, due to the increasing distance from the outdoor surface of 15 the wall.

16 Besides, the sensitivity of the model to a simultaneous change in the whole set of parameters 17 (using the same variation percentage for all parameters) was found much lower compared to 18 the sum of individual sensitivities. This result suggests that errors on input parameters don't 19 accumulate and don't influence each other, or that they compensate each other in some way. In 20 order to clear this up, the new temperature profiles simulated at 3 cm depth in the DPC wall considering a 25% variation on the most influencing parameters (C_p , ρ , h and λ , varied one by 21 22 one while keeping the other parameters constant at their initial value) are represented in Fig. 23 13. One can notice that C_p and ρ influence the temperature profiles in an opposite way compared to h and λ : increasing the values of C_p or ρ in the model results in a larger amplitude of the 24 simulated profiles, whereas higher values of h and λ tend to decrease this amplitude. This 25 26 opposite trend confirms that, when parameters are changed simultaneously, compensation 27 phenomena occurs and lower the resulting sensitivity of the model in comparison to the sum of 28 individual sensitivities. However, since variations of C_p and ρ have similar effects, errors on 29 these two parameters may accumulate and emphasize the difference of the model outcome compared to the reference response. The same comment can be made regarding accumulation 30 31 of errors on *h* and λ .

1 It should be noticed that some of the cases considered in this sensitivity analysis may not be 2 representative of actual experimental configurations, because varying a specific parameter may in reality imply changing one or several other parameters. For instance, Collet et al. [35] 3 4 reported that thermal conductivity of hemp concrete blocks is highly dependent on the material 5 density (λ increases as a function of ρ). Nevertheless, the present mathematical sensitivity 6 analysis remains relevant for parameters on which experimental errors can be made during 7 characterization. In other words, as long as a parameter can be measured independently, 8 experimental errors can be made on this specific parameter.





28



1

Fig. 13: New temperature profiles simulated at 3 cm depth in the DPC wall, 2 considering $\pm 25\%$ variations on (a) the material density ρ , (b) heat capacity C_p , (c) heat transfer coefficient h and (d) thermal conductivity λ , compared to their reference values 3

Moisture transfer (scenario 2) 4

5 Regarding the simulation of RH profiles (corresponding to scenario 2), the instant chosen for comparison of parametric sensitivities is $t=18\times10^5$ s for all depths. Fig. 14 shows the evolution 6 7 of the relative sensitivity versus the variation percentage of the model inputs. The graphs are 8 displayed for different depths of the DPC wall.

9 As previously noted in section 5.1.2, μ and W are the most influencing parameters on moisture 10 transfer, while the impact of the rest of parameters is negligible. This was expected, because 11 these two parameters describe respectively the moisture transfer and storage in the model. The 12 same result was obtained by Bart et al. [14] and Van Belleghem et al. [34].

It is important to note that porosity does not explicitly appear in the transfer equations. However, it is implicitly taken into account through the measured reference properties, which are determined assuming homogeneity of the DPC material. This may be the reason behind the low effect of density ρ on moisture transfer observed in this theoretical analysis. In reality, this effect should be remarkable since porosity facilitates moisture diffusion. For instance, it was shown that vapor permeability of hemp concrete blocks is influenced by the material density ρ [19, 35].



Fig. 14: Evolution of RH sensitivity versus the variation percentage of input parameters (with respect to reference values) at different depths of the wall: 3 cm (a), 7.5 cm (b) and 12.5 cm (c)

Besides, varying all the parameters together provides high values of relative sensitivity on Fig.
14, which are also very close to the sum of individual sensitivities. This result means that these
material properties influence each other and that their uncertainties can accumulate and affect

- 1 the model outcome to a large extend. It was previously shown on Fig. 10, that 25% variation 2 on the two main influencing parameters (μ and W) have similar effects on theoretical profiles, 3 which explains the accumulation of errors observed in case 7 (simultaneous change in all
- 4 parameters).
- 5
 Table 4: Comparison of sensitivity values for each case and each scenario at 7.5 cm depth
- and considering 25% variation on the parameters 6

Sensitivity per case	Varying parameters	Scenario 1: heat transfer	Scenario 2: moisture transfer
Reference case			
Case 1	Cp	5.33%	0.03%
Case 2	h	0.79%	0.01%
Case 3	λ	3.77%	0.11%
Case 4	μ	0.05%	3.92%
Case 5	ρ	5.31%	0.01%
Case 6	W	0.21%	4.13%
Case 7 (simultaneous variation)	C _p , h, λ, μ, ρ, W	5.97%	7.69%
Sum of individual sensitivities	C _p , h, λ, μ, ρ, W	15.47%	8.21%

8 Table 4 shows the sensitivity values obtained for the different cases of scenario 1 compared 9 with scenario 2 at 7.5 cm depth and 25% of variation. We notice that the global model's 10 sensitivity to heat transfer problems is higher than the sensitivity to moisture transfer ones, as 11 a higher error accumulation risk is obtained.

12 A major reason behind these uncertainties, both in the scenarios of heat and moisture transfer, relates to the material homogeneity assumption which is usually adopted at all scales in 13 14 experimental/numerical approaches. As a matter of facts, procedures used for material 15 preparation may represent a source of heterogeneity, especially at large scale (wall or building). 16

This can induce variations on measured material properties, and thus on the model inputs.

Fig. 15 shows the effects of positive/negative changes in the values of C_p and ρ (one parameter 1 2 is varied while the other is kept constant at the reference value) on the relative temperature 3 sensitivity, as well as the influence of changing the values of W and μ on RH sensitivity. The 4 effects of positive variations are found lower compared to those of negative changes. This was 5 also observed but not discussed in previous studies of the literature [2, 14, 34]. To our 6 knowledge, this is probably due to the nonlinear nature of the mathematical model, as the 7 coupled system involve several coefficients that depend on the system's unknowns and their 8 derivatives. Moreover, our study takes place in a transient state where the system is in a 9 continuous perturbation and varies over time. This is very different from a steady state study, 10 where the system is in equilibrium and the sensitivities usually tends to constant values.



¹¹

12**Fig. 15:** Effects of variations (\pm 25%) of selected parameters on the relative sensibility of the13model: (a) influence of C_p and ρ on the temperature sensibility, and (b) influence of W and μ 14on RH sensibility

- 15 5.2 Effect of initial conditions and sensors position
- 16 5.2.1 Effect of initial conditions

Figs. 16 and 17 show the effect of changing initial conditions on simulated profiles of temperature/RH at different depths. Regarding temperature profiles, and whatever the considered depth, these effects are mainly observed during the first heating phase and then rapidly vanish; the new profiles become very similar to the reference curve after a half cycle (about 82000s after the beginning of the test). Regarding RH profiles, these effects are more durable and can be observed throughout the test, but with lower influence at the end (deviation of about 0.55% on RH value compared to the reference curve at $t=18 \times 10^5$ s). This result was

1 expected, because of the slow kinetics of moisture transfer compared to heat transfer. It was 2 also reported by Oumeziane et al. [36], who studied the influence of initial moisture content 3 and relative humidity on the MBV tests; the authors concluded that the initial hygric state has a strong impact on moisture transfer prediction, especially when hysteresis effect is taken into 4 5 account. Globally, one must keep in mind that uncertainties on initial temperature conditions 6 are relatively tolerated for long tests. Differently, uncertainties on RH conditions should be 7 carefully evaluated, since errors as low as $\pm 2\%$ can have a significant impact on numerical 8 results.

9 5.2.2 Effect of sensors position

10 Fig. 18 shows the effect of changing the sensors position on simulated temperature profiles at 11 different depths of the DPC wall. The largest deviation from the reference curve is obtained at 3 cm of depth and equals to 0.92 °C, while at 7.5 cm and 12.5 cm of depths the maximum 12 13 differences are 0.34 °C and 0.56 °C respectively. The effect of sensors position on numerical 14 RH profiles at different depths is also shown in Fig. 19. Differences caused by a variation of \pm 15 1 cm from the reference position are more remarkable in this scenario (compared to scenario 16 1). The largest deviation on RH is also obtained at 3 cm depth (5.36%), while smaller deviations 17 of 1.66% and 1.65% are respectively found at 7.5 cm and 12.5 cm depths.

18 The accuracy of the sensors should also be taken into account in this analysis. In our case, the 19 sensor accuracy specified by the manufacturer is ± 0.5 °C on temperature and ± 1.8 % on RH 20 measurements. Temperature uncertainties are thus of the same order than temperature 21 deviations resulting from the variation of sensors position, whereas RH uncertainties are much 22 lower compared to RH deviations caused by errors on sensors position. Therefore, meticulous 23 care should be taken while positioning the sensors on the experimental setup, especially for 24 locations close to the outdoor surface of the wall where small errors may cause large data 25 discrepancies. Similar conclusions were reported by Bart et al. [14], who studied the effect of 26 sensors position on measured hygrothermal properties in the case of a 30 cm thick hemp 27 concrete wall.



Fig. 16: Effect of changing initial temperature conditions on numerical temperature profiles
 obtained at different depths: 3 cm (a), 7.5 cm (b) and 12.5 cm (c)



Fig. 17: Effect of changing initial RH conditions on numerical RH profiles obtained at
 different depths: 3 cm (a), 7.5 cm (b) and 12.5 cm (c)

4 **5.3** simplified models

5 5.3.1 Pure conduction model

Fig. 20 displays the numerical temperature profiles provided by the model based on pure conduction transfer only, and by the full Kunzel model (reference case), at different depths of the DPC wall. Differences between the numerical outcomes of the pure conduction and full models did not exceed 0.25 °C, whatever the considered depth. This result shows that heat conduction is the main transfer mode involved in the case of the DPC wall subjected to conditions of scenario 1. This is also validated by the fact that RH boundary conditions were

1 kept unchanged, and hence any moisture motion was the result of heat transfer due to the 2 coupling effects between heat and moisture transfers (very small effect). This is translated 3 mathematically by the phase change term in the right-hand side of Eq. 3, where the heat of phase change is related to vapor flux. In other words, this phase change term, which represents 4 5 the only difference between the pure conduction and the full models, didn't play an important 6 role in the studied case since the relative humidity was unchanged. Similar observations were 7 noted by other authors on temperature profiles in the absence of phase change phenomena inside 8 the material [31]. In such cases where only temperature changes are observed, one can conclude 9 that the pure conduction model is sufficient to predict temperature profiles through a bio-based date palm concrete wall. This simplification provides a substantial gain of about 30% in 10 11 calculation time.

12 5.3.2 Neglecting liquid transfer

13 Fig. 21 compares the numerical RH profiles obtained with the full Kunzel model (reference 14 case) with those obtained with a simplified model neglecting liquid transfer. It can be noticed 15 that the two models provide very different outcomes whatever the considered depth. Deviations 16 on RH values increase during the adsorption phase, with differences up to 3.8%, 4% and 2.3% 17 at 3 cm, 7.5 cm and 12.5 cm respectively. Conversely, the gap between the two models is less important during desorption phase and becomes even negligible at 3 cm depth. This result can 18 19 be explained by the fact that liquid and vapor transfers play both important roles during the adsorption phase, while liquid transfer is less influent during the desorption phase. Liquid 20 21 transfer is naturally present in the adsorption phase, mainly through a surface diffusion process 22 that starts to appear earlier in the hygroscopic region [11]. This process contributes to moisture 23 transfer and accelerates its kinetics. Therefore, the full model provides numerical RH profiles 24 that rise faster during adsorption phase and decrease faster during desorption phase, compared 25 to the simplified model. Consequently, it is recommended not to neglect liquid transfer when 26 modeling moisture transfer in porous building materials.



Fig. 18: Effect of changing sensors position on numerical temperature profiles determined at
different depths of the DPC wall: 3 cm (a), 7.5 cm (b) and 12.5 cm (c)



Fig. 19: Effect of changing sensors position on numerical RH profiles determined at different depths of the DPC wall: 3 cm (a), 7.5 cm (b) and 12.5 cm (c)



Fig. 20: Comparison of numerical temperature profiles provided by the pure conduction
 model and by the full Kunzel model (reference case), at different depths of the DPC wall: 3
 cm (a), 7.5 cm (b) and 12.5 cm (c)



Fig. 21: Comparison of numerical RH profiles provided by the model neglecting liquid
transfer mode and by the full Kunzel model (reference case), at different depths of the DPC
wall:3 cm (a), 7.5 cm (b) and 12.5 cm (c)

1 6 Conclusion

In this work, a parametric sensitivity analysis was carried out on the transient heat and moisture transfer response of Kunzel's model. This model was applied to simulate the hygrothermal behavior a wall made of Date Palm Concrete (DPC), and boundary conditions for heat and moisture transfer scenarios were imported from real experiments. The effects of finite variations of the model inputs (reference material parameters and initial conditions) on the numerical outcome were investigated.

8 The results showed that the theoretical description of heat transfer is very influenced by changes in the heat capacity C_p and density ρ , which generates errors on temperature profiles 9 reaching about 5.3% in terms of relative sensitivity. On the other hand, parameters 10 influencing moisture transfer are mainly the sorption isotherm W and the vapor resistance 11 factors μ , which are responsible for a relative sensitivity of 4% on RH profiles. Therefore, a 12 particular attention should be paid to the experimental determination of these reference 13 properties at the material scale, and their validity at higher scales (wall or building) should be 14 verified as well. It was also noted that moisture transfer is very sensitive to any errors in the 15 initial conditions and sensors position. Uncertainties on initial RH values may influence the 16 17 results even in the case of long tests (18 days' test in our case). Furthermore, uncertainties of 18 \pm 1 cm on the sensors position may lead to 5.36% deviation on RH profiles.

Finally, some terms were neglected in the model, such as the heat of phase change and liquid diffusion, in order to explore the possibility of simplifying the model. It was shown that a pure conduction model is able to accurately describe the heat transfer process, whereas neglecting liquid diffusion leads to significant underestimations in RH profiles during the adsorption phase and should thus be proscribed.

24

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