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Experimental and numerical evaluation of the effect of micro-aeration on the thermal properties of chocolate

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Thermal properties, such as thermal conductivity, specific heat capacity and latent heat, influence the melting and solidification of chocolate. The accurate prediction of these properties for micro-aerated chocolate products with varying levels of porosity ranging from 0% to 15% is beneficial for understanding and control of heat transfer mechanisms during chocolate manufacturing and food oral processing. The former process is important for the final quality of chocolate and the latter is associated with sensorial attributes, such as grittiness, melting time and flavour. This study proposes a novel multiscale finite element model to accurately predict the temporal and spatial evolution of temperature across chocolate samples. The model is evaluated *via* heat transfer experiments at temperatures varying from 16 °C to 45 °C. Both experimental and numerical results suggest that the rate of heat transfer within the micro-aerated chocolate is reduced by 7% when the 15% micro-aerated chocolate is compared to its solid counterpart. More specifically, on average, the thermal conductivity decreased by 20% and specific heat capacity increased by 10% for 15% micro-aeration, suggesting that micro-pores act as thermal barriers to heat flow. The latter trend is unexpected for porous materials and thus the presence of a third phase at the pore's interface is proposed which might store thermal energy leading to a delayed release to the chocolate system. The developed multiscale numerical model provides a design tool to create pore structures in chocolate with optimum melting or solidifying response.

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1. Introduction

Chocolate is a popular food product, which has a complex behaviour due to its multiphase nature. Chocolate consists of hard particles such as sugar crystals, cocoa solids and fat crystals dispersed in a soft continuous matrix, the cocoa butter.¹ There is extensive literature investigating the effects of chocolate composition on material properties such as rheological effects, and on consumer perception.^{2–6} The importance of material properties has been thoroughly reported and linked to flow processes,^{2,7} oral processes^{4,8} and cooling processes.^{9–11} Carvalho da Silva¹² highlighted that the melting behaviour of chocolate plays a determinant role in the final quality of the chocolate.

Thermal properties, such as specific heat capacity, thermal conductivity and latent heat, affect the melting and solidification/crystallisation behaviour of chocolate^{13,14} and therefore,

the accurate measurement and prediction of these thermal properties is crucial. These include the cooling stage during manufacturing, which is important for the final product quality, and the consumer experience during consumption or oral processing of the chocolate, where melting is associated with several sensorial attributes, such as melting time and flavour intensity.¹⁵ These thermal properties have been evaluated experimentally and numerically.

Amongst others, Le Reverend *et al.*¹³ highlighted the importance of developing a numerical model that considers the thermal properties of chocolate, coupled with the phase transformation kinetics of fat crystals, to predict the crystallisation evolution during the cooling stage of chocolate. This can ultimately be used to control and improve the cooling process during chocolate manufacturing, ensuring the formation of the desired crystal form. Several computational modelling studies simulate the cooling process during chocolate tempering or before demoulding.^{14,16,17}

In addition, Haedelt, Beckett, and Niranjan¹⁵ reported a link between melting behaviour of chocolate during oral processing and sensory properties for different porous chocolate products with different pore sizes through a series of *in vivo*

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sensory tests. Ziegler, Mongia, and Hollender¹⁸ reported similar findings and highlighted the effect of melting time on the release of flavour attributes: the slow melting of porous chocolate will release flavour attributes more intensely compared to solid chocolate, which melts rapidly. However, no studies are available that explain the mechanisms relating the effect of porosity or pore size to the melting behaviour of chocolate and consequently to consumer perception. Oral processing of chocolate lasts approximately 20 seconds¹⁹ and within this timeframe all processes change dynamically. Chocolate melting assists the release of all flavour attributes captured inside the chocolate and is related to textural attributes such as moistness, melting time and grittiness.

Thus, the accurate prediction of the thermal properties is required to understand and control both the cooling stage during manufacturing and the melting behaviour during food oral processing. Even though the importance of the thermal properties to the crystallisation of chocolate and to the cooling stage during the manufacturing process has been highlighted previously and thorough investigation has been performed, this is not true in the case of the melting process during oral process. To the authors' knowledge, there is no literature that systematically tackles this topic in an analogous way as has been done for the cooling process.

Material properties are strongly dependent on the chocolate recipe and composition. This is the reason why thermal property values for different chocolates slightly deviate across literature.^{13,14,20,21} Across the literature, differential scanning calorimetry (DSC) devices are commonly used to investigate the crystallisation form of chocolate and measure the latent heat and the specific heat capacity against temperature and rate of heating or cooling.^{13,17,22–24} Unfortunately, such instruments are not able to provide measurements of the thermal conductivity, at least directly. Commercially available instruments or methods to measure thermal conductivity, such as the hot wire method (HWM) and the laser flash method (LFM), exhibit limitations when applied to materials which are not solid and exhibit low melting temperature points, such as chocolate. These techniques provide more accurate results when a large temperature testing range can be employed, which produces an additional limitation to their implementation in the study of chocolate. However, there are some studies that report thermal conductivity measurements on chocolate using custom-designed experimental setups coupled with numerical modelling methods calibrated on these experimental setups.^{13,14,17}

Literature typically only refers to non-aerated chocolate, either milk or dark and does not address the thermal properties of micro-aerated chocolates. Micro-aeration has been utilised in chocolate systems as a method to modify textural attributes such as hardness, crumbliness, grittiness, and stickiness.^{8,15} The micro-pores inside the chocolate system are found to have an average diameter of 40 μm , which is of the same scale as other ingredients found in chocolate such as cocoa solids and sugar particles. A micro-pore suspension is developed inside the chocolate by injecting gaseous N_2 before

the moulding process, when the chocolate is already tempered. It has been shown that micro-aeration of chocolate significantly affects mechanical properties such as Young's modulus, fracture toughness, and friction coefficient^{4,8} but there is no data available that relates the thermal properties of chocolate, such as thermal conductivity, latent heat, and specific heat capacity to the level of micro-aeration of the chocolate. These thermal properties are important parameters to determine the chocolate thermal response under transient thermal loading conditions.

A review on porous foods by Dadmohammadi and Datta²⁵ reported the significance of investigating the link between the porous structure of the resulting food product and its quality, both in terms of material properties and texture. The pores can significantly influence the microstructure of the food and therefore affect processes such as cooking, drying and freezing. In literature, experimental, numerical, and analytical approaches have been employed to investigate the thermal behaviour of porous foods.^{26–28}

Many analytical models have been developed to predict the effective thermal properties of foods.^{27,29–31} These analytical models employ a range of assumptions for the porous microstructure (such as the distribution and shape of pores *etc.*) to devise a relationship associating the thermal properties of the mixture and the thermal properties of the constituents and porosity. Semi-empirical analytical models³¹ that employ additional fitting parameters – often without any physical justification – also exist, which are not considered in the current work. Despite the large number of proposed analytical models, there is yet no conclusive model to accurately predict the effective thermal properties of multiphase foods. Therefore, the need for a conclusive and accurate model to describe the complex multiphase microstructure and behaviour of food materials remains.

Numerical modelling is an attractive approach to simulate the engineering of foods, due to the high versatility of the methods, low costs and quick result turnarounds.^{26–28,32} The ability to change parameters such as environmental (temperature, moisture content *etc.*)^{27,28,33} and geometrical (size, and shape)^{27,34} conditions and record parameters that are otherwise impossible to extract from *in vivo* tests (*e.g.* simulation of chewing), constitutes this method a powerful design tool for the food industry. However, despite the extensive use of such models, they fail to describe all the physical mechanisms taking place inside these materials. For example, no studies have been presented showing the heat transfer in porous foodstuff at the microscale and the resulting consequences on the averaged thermal properties of the bulk material.

Our recent work on the effect of micro-aeration on chocolate structure, mechanical, and sensory properties during oral processing highlighted the importance of heat transfer and chocolate melting on the structural breakdown during mastication that is expected to alter its' perceived consumer experience.³⁵ It was found that the combined action of heat transfer and fragmentation guides chocolate oral processing after the first bite. The need to compute the true properties of the



micro-porous chocolate under thermal conditions relevant to oral processing remained as previous work only used macro-scale analytical models to predict the effects of micro-aeration on the thermal properties of chocolate rather than an extensive micro-FEM calculation, *i.e.* the focus of the current paper. The Finite Element Model of the current work may prove a useful design tool to create chocolate products with controlled sensorial attributes.

The aim of this study is to investigate the thermal behaviour of three milk chocolates with varying levels of micro-aeration: one non-aerated and two micro-aerated with 10% and 15% porosity. This work targets to develop a multiscale model which can simulate the unsteady heat transfer phenomena within the chocolate. The unsteady heat transfer effects are important for the oral processing of chocolate. This effort reflects food industry's needs to investigate how small structural changes, such as low levels of porosity, can change material and sensorial properties of chocolate during oral processing that affect the consumer experience.³⁵ Measurements on a novel heat transfer rig and a commercially available DSC instrument are coupled to a multiscale finite element model to predict the thermal temporal response produced by these different porous structures. The multiscale numerical model is composed of a heat transfer model at the microscale to compute the average (effective) thermal properties of the resulting porous material which are then used in a heat transfer model at the macroscale to predict the experimental results. This multiscale numerical model can be utilised as a design tool to optimise the microstructural parameters, such as porosity level, pore size, pore distribution, to adjust the melting or solidifying behaviour of chocolate in a time and cost effective way.

The paper is structured as follows; section 2 summarises the experimental, numerical, and analytical methods used. Section 3 presents and discusses the results obtained using these methods. The paper ends with a summary of the findings.

2. Materials & methods

2.1. Materials

Three types of milk chocolates with different porosity levels of 0%, 10% and 15% were provided by Nestlé Product Technology Centre in York, UK and used for the heat transfer experiments. These samples consisted of 44 wt% sugar particles, 27 wt% cocoa fat, 10 wt% whole milk powder and 6 wt% non-fat cocoa solids. The initial geometry of the chocolate is a rectangular beam with a length of 80 mm, a width of 15 mm, and a height of 8 mm. For a detailed description of the manufacturing process and the micro-aeration process, the reader is referred to Bikos *et al.* 2021.⁸ The chocolate geometries were further modified using a razor saw to a final dimension of 50 mm length (L), and 8 mm in width (W) and height (H), as shown in Fig. 1. Blind holes of 0.6 mm diameter with a depth

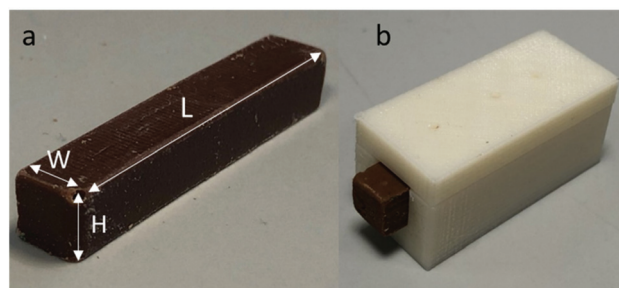


Fig. 1 (a): Non-aerated chocolate used for the heat transfer experiments. (b) Non-aerated chocolate placed in the thermally insulated sample holder.

of up to half of the height (H) were created using the tip of a needle.

Cocoa butter samples were manufactured to be tested in the DSC tests to assess the specific heat capacity of cocoa butter. Refined, untempered cocoa butter was provided by Nestlé Product Technology Centre in York which was tempered using a custom designed double bath to ensure the relatively slow and controlled heating of the material. Following the heating-cooling protocol the commonly used seeding method was employed to temper the cocoa butter, as described by Beckett.³⁶ After the tempering process the cocoa butter mass was deposited into cylindrical moulds of 25 mm diameter and height.

2.2. Experimental methods

A heat transfer rig was developed which was able to heat or cool the ends of the beam-shaped chocolate samples independently and record the spatial and temporal temperature profile. As shown in Fig. 2, the heat transfer rig consisted of the testing section, the control unit and a data logger, which recorded the temperature evolution at three locations across the chocolate sample. The testing section was the part of the rig where the chocolate samples were actively heated or cooled. The control unit consisted of a combination of two PID controllers (CN142 Omega Ltd, UK). The temperature profiles were recorded using an 8-channel thermocouple data logger (OM-HK-EH-TC Omega Ltd, UK) at different locations in the chocolate sample T_1 , T_2 , and T_3 , (see Fig. 2) placed in the centre of the sample ($H/2$).

The ends of the chocolate samples were inserted into square-shaped grooves of size 5 mm length, height, and width. The grooves were machined into the side surface of the aluminium blocks as shown in Fig. 2. The aluminium blocks had a size of 50 mm length, 50 mm height, and 40 mm width (width is not shown in Fig. 2) and used to support the chocolate samples and ensure that the heat transfer was uniform over the area inside the grooves. As shown in Fig. 2, on the opposite side of each of the aluminium blocks, Peltier Modules (ET-161-12-14-E, Adaptive Thermal Management, UK) with a maximum power of 62.3 W were placed to apply or remove heat to/from the sample. On the other side of the



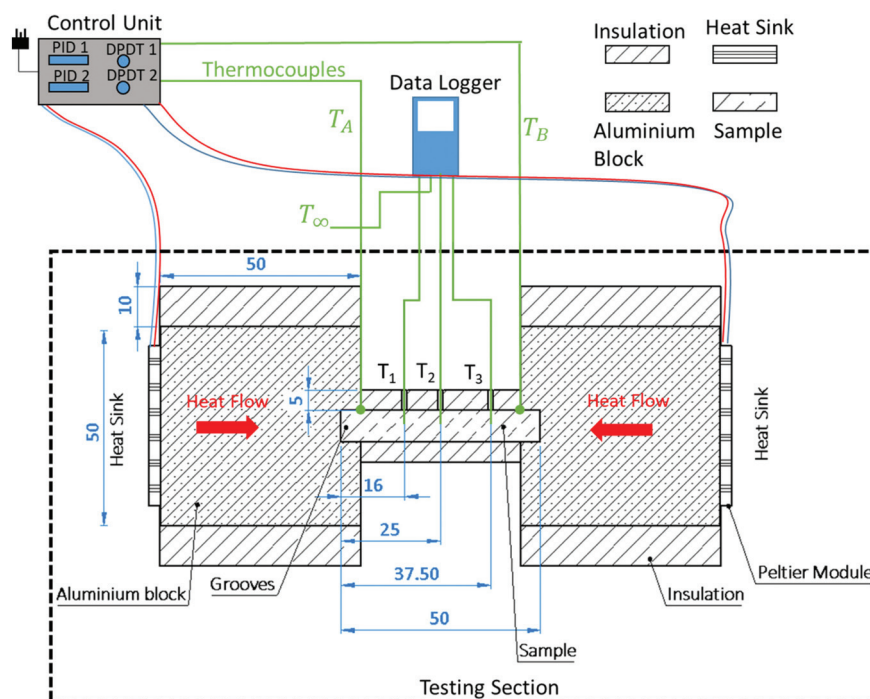


Fig. 2 Schematic of the experimental setup showing the thermal test rig and connections to control units, thermocouples and data logger, together with the testing apparatus used for the heat transfer experiments on chocolates with the thermocouple positioning. The pattern of the section lining on the upper right handed corner indicate the different components of the thermal rig.

Peltier modules, two Aluminium heatsinks (LAM 4 Fischer Elektronik, Germany) with a thermal resistance of $0.4\text{ }^{\circ}\text{C W}^{-1}$ ventilated by standard computer axial fans (Multifan S5 AC Infinity Inc., USA) were used to ensure that sufficient heat was dissipated/absorbed by the Peltier module exhausts. The fans were only used in the case where the Peltier modules were cooling the chocolate due to the required, relatively high, energy dissipation for the Peltier modules to provide effective cooling.

The Peltier modules were each controlled by individual PID controllers to ensure a constant set temperature of the sample ends when placed inside the aluminium grooves. This was achieved by monitoring the temperature of each groove using a thermocouple connected to each of the two PID controllers, see thermocouples T_A and T_B in Fig. 2. The PID controllers were modulating the power delivery to the Peltier modules according to the target temperature and actual readings of the groove thermocouples. Using a double pole double throw switch (DPDT) for each Peltier module, as shown in Fig. 2, the polarity of the Peltier module could be reversed to either heat or cool the samples. An example of the direction of heat flow inside the aluminium block and, by extend, the chocolate sample is shown in Fig. 2 for one of the scenarios investigated (*i.e.* scenario 1 and 2 are explained below, see Table 1). The aluminium block and chocolate sample were insulated using 3D-printed custom sleeves in order to limit heat losses. The material used for the insulation was ABS with micro-pores in the shape of honeycomb to increase thermal insulation performance.

Table 1 Experimental boundary conditions for all scenarios, 1, 2, 3, and 4

Scenario	$T_{\text{left}},\text{ }^{\circ}\text{C}$	$T_{\text{right}},\text{ }^{\circ}\text{C}$
1	35	35
2	45	45
3	16	16
4	45	16

The temperature of the core of the chocolate specimen was recorded at three different locations T_1 , T_2 , and T_3 , which corresponded to distances of 16 mm, 25 mm, and 37.5 mm, using k-type thermocouples. Four different thermal loading scenarios were implemented, three within the melting temperature range of the chocolate and one outside. Scenario 1 represented heating to a constant temperature of $35\text{ }^{\circ}\text{C}$ on both sides, scenario 2 rapid heating to a constant temperature of $45\text{ }^{\circ}\text{C}$ on both sides, scenario 3 cooling to a constant temperature of $16\text{ }^{\circ}\text{C}$ on both sides, and scenario 4 rapid heating to a constant temperature of $45\text{ }^{\circ}\text{C}$ at the left side and cooling to a constant temperature of $16\text{ }^{\circ}\text{C}$ at the right side of the testing section. The ambient temperature and relative humidity for all experiments was controlled and set to $23\text{ }^{\circ}\text{C}$ and 50% respectively. The boundary conditions of all scenarios are summarised in Table 1. Three repeats were performed for each loading scenario. The uncertainties of the thermocouple measurements were estimated to be approximately $0.1\text{ }^{\circ}\text{C}$ or 1% of the measured value and the sampling period was 5 s.



Preliminary experiments were performed, using the thermal rig shown in Fig. 2, on aluminium specimens of known physical and thermal properties and same geometries as the chocolate samples. The experiments were used to validate the experimental setup and calibrate the boundary conditions at the interface of the tested sample and the aluminium block and the boundary conditions at the insulated surfaces which are required for the numerical model of section 2.3.1.

Standard differential scanning calorimetry (DSC) experiments were conducted on all three types of chocolate (DSC Q2000, TA Instruments, USA) as follows. All tests were performed at a temperature range between 12 °C and 45 °C at two linear rates of 1 °C min⁻¹ (the lowest possible rate for the instrument) and 10 °C min⁻¹. Each test was performed twice for each type of chocolate. These tests were also performed for cocoa butter to provide initial indications on the specific heat capacity of the material that forms the chocolate matrix. Helium was selected as a purge gas for the calorimetric chamber. The heat flow rate data obtained from the DSC tests were then converted into specific heat capacity measurements using eqn (1):

$$C_p = \frac{\dot{Q}}{\frac{dT}{dt} m} \quad (1)$$

where C_p is the specific heat capacity of the chocolate, \dot{Q} is the heat flow rate into the chocolate measured by the DSC, m is the mass of the DSC test sample, and $\frac{dT}{dt}$ is the applied heating rate.

The DSC instrument measures the heat flow rate \dot{Q} for a given temperature and produces a \dot{Q} – temperature graph, which in turn, can be converted into a C_p – temperature curve using eqn (1). The area underneath a C_p – temperature curve, at the part of the curve where the phase occurs (discontinuity in the curve), represents the latent heat of the chocolate. The latent heat values ΔH were measured using the TA Universal Analysis software (TA Instruments, USA).

2.3. Numerical methods

In this section, the methodology of the numerical steady-state and unsteady heat transfer models at the microscale (micro-FEM) and the unsteady heat transfer model at the macroscale (macro-FEM) will be presented. The calculation of the effective material properties, such as specific heat capacity, thermal conductivity, density and thermal diffusivity, in relation to porosity will be presented using micromechanical modelling (micro-FEM). These material properties will then be used as input parameters for a macro-FEM to simulate the experimental conditions of the two porous chocolate samples. The sequence of the numerical studies is summarised in Fig. 3 and used as follows; the macro-FEM model was initially introduced for non-aerated chocolate to calibrate the thermal conductivity based on the temperature-time data extracted from thermal rig experiments. The micro-FEM models were then followed to calculate the thermal properties as a function of porosity using the thermal properties of the solid chocolate. Finally, the macro-FEM model was re-employed, but this time for the micro-aerated chocolates. The results from the macro-FEM for all chocolate materials were compared against the experimental data from the thermal rig. In all numerical models, the commercially available software Abaqus³⁷ and the Standard/Implicit Abaqus solver was used.

2.3.1. Macromechanical finite element model (macro-FEM). A 3D model was developed with the same 50 × 50 × 8 mm geometry as the chocolate samples used for the experiments (see Fig. 1a). A total number of 6200 linear heat transfer brick elements with 8 nodes (DC3D8) were used for the macro-FEM, as shown in Fig. 4a. A mesh sensitivity analysis was employed to ensure that selected mesh density showed representative results. The results showed that a mesh with a number of elements higher than 6200 delivers converged results, *i.e.* maximum 0.3% deviation, which lies within the experimental error (~1%) of the reference validation cases.

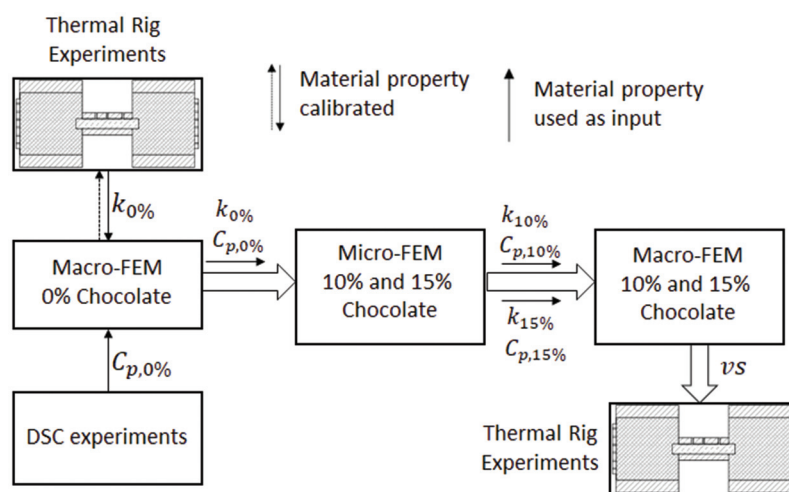


Fig. 3 Flow chart showing the sequence of the numerical studies at microscale (micro-FEM) and macroscale (macro-FEM). When a material property was calibrated based on experiments a double arrow was used to depict that.



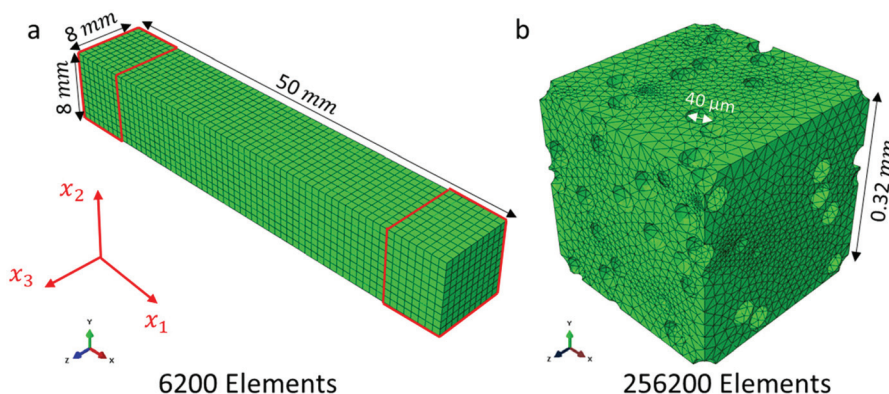


Fig. 4 Geometries of the heat transfer FE-models at (a) at macro-scale (macro-FEM) using the effective thermal properties from the micro-FEM and (b) at micro-scale for the 10% porosity chocolate (micro-FEM). The highlighted regions with red colour denote the areas where the temperature boundary conditions were applied. The red coordinate system shows the origin.

An implicit transient uncoupled heat transfer step was selected in Abaqus to simulate the experimental conditions with the thermal test rig. The transient step in Abaqus follows the backward Euler method and does not exhibit any stability issues (*e.g.* mesh density or time increment) for problems when conduction is dominant.³⁷ The time step was selected to be less than 1% of the relevant time scale observed during the heat transfer experiments. Therefore, a maximum time step of 1 s was selected for the macro-FEM model. The analysis was complete when a modelled time of 4000 s was reached, same as the overall time from the heat transfer experiments.

In Fig. 4a, the highlighted volume with red colour at both ends of the specimen indicates the volume in contact with the aluminium block, and, therefore, represents the volume of elements where the temperature boundary conditions were applied. The boundary conditions of the macro-FEM at the highlighted regions (Fig. 4a) together with ones at the insulated surfaces, which are required for the macro-FEM model, were calibrated based on the preliminary experiments, using the thermal rig of Fig. 2, on aluminium specimens. The need for these preliminary experiments was an unexpected mismatch between the applied temperature from the Peltier module and the target temperature of the grooves due to heat losses and thermal contact resistance at the contact surfaces between aluminium block and the tested sample. These preliminary experiments were employed for the four scenarios of Table 1. Using the unsteady heat transfer macro-FEM model described earlier but this time on the aluminium specimens with known material properties, the heat transfer coefficients at the boundary conditions at the aluminium block-sample interface, h_p , and at the insulated surfaces, h_{ins} , were evaluated. The same boundary conditions h_p and h_{ins} as evaluated from the preliminary tests on aluminium specimens were used for the non-aerated chocolate and two micro-aerated chocolates and shown in section 3.2.

Finally, the definition of the material properties of chocolates were required for the macro-FEM model. In the case of

the non-aerated chocolate, the C_p and ΔH values were obtained from the DSC experiments described in section 2.2. The density of the non-aerated chocolate was calculated from the weight for a known volume. However, the thermal conductivity could not be extracted directly either from the thermal rig or the DSC experiments. Therefore, a starting value of $0.3 \text{ W m}^{-1} \text{ K}^{-1}$ for the thermal conductivity both at solid and liquid state was selected, which was within the range from commonly reported values in the literature ($0.16 \text{ W m}^{-1} \text{ K}^{-1}$ – $10.7 \text{ W m}^{-1} \text{ K}^{-1}$).^{13,14,17} The thermal conductivity values at solid and liquid states were then adjusted using the macro-FEM model to match the experimental data. In a similar way to the DSC experiments, the thermal properties, such as thermal conductivity and specific heat capacity, were grouped to those for the solid state (temperatures lower than 28°C) and liquid state (temperatures higher than 34°C). During phase change of chocolate (temperatures between 28°C and 34°C), the process was driven by latent heat ΔH . In the case of the two micro-aerated products, the C_p and ΔH values were also computed from the DSC experiments and the density value was evaluated in the same way as the non-aerated chocolate. The effective thermal properties such as the thermal conductivity and the specific heat capacity were obtained from the micro-FEM.

2.3.2. Micromechanical finite element model (micro-FEM). A 3D micromechanical numerical model (micro-FEM) was developed using Abaqus with pores dispersed in the chocolate matrix to simulate the micro-aerated chocolate samples. As reported by Bikos *et al.*,⁸ the average pore size found in micro-aerated chocolate with either 10% and 15% porosity is approximately $40 \mu\text{m}$. Therefore, random pore topology coordinates were created using Macropac (MacroPac, Oxford Materials, UK) within a cubic domain to control the porosity of the resulting modelled micro-aerated chocolate samples. The coordinate data was then used to create the $40 \mu\text{m}$ three-dimensional pore spheres for the aerated models. The geometry of the spheres was then imported to Abaqus to create a



cubic 3D volume with pores dispersed within this volume as shown in Fig. 4b.

A minimum volumetric size or representative volume element (RVE) with 30 pores is needed to achieve representative thermal results when periodic boundary conditions are used.^{38,39} The number of pores inside the RVE needs to be increased at least to 60 pores to ensure that the calculated effective thermal properties remain unaffected by the type of boundary conditions employed.^{40,41} Dirichlet boundary conditions (constant surface temperature) were used in the current work. Therefore, three different 3D RVEs with different pore arrangement were created with 100 pores dispersed within the structure to enable the estimation of representative values of the thermal conductivity and the specific heat capacity.

For an RVE with 10% porosity, an average number of 256 200 4-node linear heat transfer brick elements (DC3D4) was used for the heat transfer analysis at the microscale, as shown in Fig. 4b. In this case, the RVE had a length of approximately 0.32 mm. Due to the small pore diameters (*i.e.* 40 μm) present in the chocolate, the dominant mode of heat transfer is expected to be pure conduction, whilst convection and radiation can be neglected^{42–44} because closed pores smaller than 10 mm lead to a relatively small Rayleigh number (<1000), which suggests convection has only a minor effect. More specifically, the possibility of significant convection effects to be present within a pore is expressed in terms of the Rayleigh number, which is a function of the Prandtl and Grashof numbers, and gives details regarding the flow characteristics within the pore. A critical value of the Rayleigh number close to 1000 indicates that convection effects within the pore cannot be neglected. Based on Clyne *et al.*,⁴³ for a spherical pore, this critical value is reached for a pore size of 10 mm, which is much larger than the sizes investigated in the current work (*i.e.* 10 s of microns). Based on the same studies, for temperature differences below 1000 $^{\circ}\text{C}$, radiation can also be neglected. The thermal properties of the chocolate were selected to be the same as the unsteady macro-FEM.

An implicit uncoupled steady-state and unsteady state heat transfer step was selected for the micro-FEM models to compute the k_{eff} and $C_{\text{p,eff}}$ respectively. A maximum time step of 0.01 s was selected for the unsteady micro-FEM model, which corresponded to 1% of the relevant time scale of the heat transfer within the RVE. The unsteady micro-FEM is simulated until a steady state solution is reached, which means that the temperature gradient is less than $10^{-5} \text{ }^{\circ}\text{C s}^{-1}$. Both models used the same RVE geometries, whilst the only differences were the actual step used in Abaqus and the applied boundary conditions at the surfaces. In addition, a mesh sensitivity analysis was performed both for the steady-state and unsteady steps to ensure representative results regardless of the mesh density of the model. Mesh independent results are achieved for mesh densities higher than 120 000 elements. However, in some RVEs, the mesh density was further increased to represent better the geometry of the pores within the RVE.

The relationship between heat flux and temperature is described by Fourier's law for steady-state heat conduction, see eqn (2):

$$Q_i = A k_{ij} \frac{\partial T}{\partial x_j} \quad (2)$$

where Q_i is the heat vector, k_{ij} is the thermal conductivity tensor, T is the temperature at a given point, and A is the cross sectional area of the RVE, whilst i and j indicate the directions and take the values 1, 2, and 3 for the directions x , y , and z respectively.

As boundary conditions, an arbitrary temperature difference ΔT of 10 $^{\circ}\text{C}$ between the bottom surface ($x_2 = 0$) and top surface ($x_2 = 0.32 \text{ mm}$) was applied, whilst the other surfaces were thermally insulated ($\frac{\partial T}{\partial x_1}(0, t), \frac{\partial T}{\partial x_1}(0.32, t), \frac{\partial T}{\partial x_3}(0, t), \frac{\partial T}{\partial x_3}(0.32, t) = 0$). The actual value of the input ΔT does not impact the resultant effective properties.

Under these boundary conditions and the assumption of an isotropic material ($k_{11} = k_{22} = k_{33} = k$, and $k_{ij} = 0$ when $i \neq j$), the effective thermal conductivity of porous chocolate is given by the following relationship:

$$k_{\text{eff}} = \frac{Q|_{x_2=0}}{A \Delta T} t \quad (3)$$

where k_{eff} is the effective thermal conductivity of the porous chocolate, $Q|_{x_2=0}$ is the total heat at the bottom surface of the RVE along the y -axis and t is the thickness of the RVE, *i.e.* 0.32 mm.

The two unknowns in eqn (3) are k_{eff} and $Q|_{x_2=0}$, which can be evaluated from the micro-FEM.

For the unsteady heat transfer analysis, the heat conduction equation is given by eqn (4):

$$\frac{\partial T}{\partial t} = a \frac{\partial^2 T}{\partial x_i^2} \quad (4)$$

where t is the time, a is the thermal diffusivity given by:

$$\alpha = \frac{k}{\rho C_p} \quad (5)$$

The unsteady heat transfer model was simulated similarly to a calorimetry experiment in order to measure the effective specific heat capacity of the RVE. Specifically, a pulse of heat with a total value of 0.5 mJ was applied at the bottom surface of the RVE along y -axis ($Q|_{x_2=0} = 0.5 \text{ mJ}$) for a duration of 1 s, subsequently the thermal load was removed and the surface was considered as thermally insulated. The other RVE surfaces were also thermally insulated.

$$\left(\frac{\partial T}{\partial x_1}(0, t), \frac{\partial T}{\partial x_1}(0.32, t), \frac{\partial T}{\partial x_3}(0, t), \frac{\partial T}{\partial x_3}(0.32, t), \frac{\partial T}{\partial x_2}(0.32, t) = 0 \right).$$

The unsteady heat conduction equation of eqn (4) for a uni-axial heat transfer without heat losses, uniform temperature



distribution within the material and boundary conditions described above can be simplified to eqn (6):

$$Q|_{x_2=0} = \rho_{\text{eff}} V C_p \Delta T \quad (6)$$

where ρ_{eff} is the effective density and V the volume of the RVE including the volume of the pores. Using the temperature rise predicted by the micro-FEM, following the heat pulse applied to the RVE, the $C_{p,\text{eff}}$ of the porous structure can be computed, see eqn (6). The effective mass density of the RVE can be evaluated using eqn (7):

$$\rho_{\text{eff}} = f \rho_g + (1 - f) \rho_m \quad (7)$$

Using the values of k_{eff} and $C_{p,\text{eff}}$ computed from the steady state and the transient (unsteady) micro-FEM, and the value of ρ_{eff} from eqn (7), the thermal diffusivity α_{eff} of all chocolate materials can be determined through:

$$\alpha_{\text{eff}} = \frac{k_{\text{eff}}}{\rho_{\text{eff}} C_{p,\text{eff}}} \quad (8)$$

This thermal diffusivity can be plotted against porosity to determine the “speed of heat transfer” within the chocolate materials. This property is important to understand the material's heat transfer response.

The predicted values of k_{eff} and $C_{p,\text{eff}}$ from the micro-FEM and the ρ_{eff} from eqn (7) for 10% and 15% porosity will be used as input parameters into the unsteady heat transfer model at macroscale (macro-FEM) and will be compared with the temperature profiles of the micro-aerated chocolate of 10% and 15% porosity respectively for all four scenarios in Table 1.

2.4. Analytical methods

In this section, the analytical models for the effective thermal conductivity and specific heat capacity that are commonly used in the literature for foods and general porous materials are described. The porous chocolate studied in the current work is composed of two phases, the continuous chocolate matrix and the dispersed spherical N_2 pores.

Wiener bounds are the simplest models for isotropic materials when conduction is the dominant heat transfer mechanism.^{30,31,45,46} The models represent the case that the porous material is a mixture of horizontal (series model, SM) or vertical layers (parallel model, PM) of the two phases. The weighted average of the conductivities of the two phases in each case corresponds to the effective thermal conductivity of the mixture, as shown below for eqn (9) (PM) and eqn (10) (SM):

$$k_{\text{eff}} = f k_g + (1 - f) k_m \quad (9)$$

$$\frac{1}{k_{\text{eff}}} = \frac{f}{k_g} + \frac{1 - f}{k_m} \quad (10)$$

where k_{eff} is the effective thermal conductivity of the porous material, k_g is the effective thermal conductivity of the pore, k_m is the effective thermal conductivity of the chocolate matrix, and f is the porosity.

Similarly, the Hashin–Shtrikman bounds or commonly known as Maxwell–Eucken models also apply to isotropic materials and provide predictions for porous materials with porosity levels lower than 0.25.^{30,31,45} In contrast to the Wiener bounds, these models assume that the pores are spherical. The bounds of these models refer to the case where the pores are in a dispersed phase, MEM-1, (eqn (11)) or in a continuous phase, MEM-2, (eqn (12)). Eqn (11) is relevant to this study, where the N_2 pores are dispersed in the chocolate matrix.

$$k_{\text{eff}} = k_m \frac{2k_m + k_g - 2(k_m - k_g)f}{2k_m + k_g - (k_m - k_g)f} \quad (11)$$

$$k_{\text{eff}} = k_g \frac{2k_g + k_m - 2(k_g - k_m)(1 - f)}{2k_g + k_m - (k_g - k_m)(1 - f)} \quad (12)$$

In the case of the effective medium theory model, EMT, (eqn (13)) both the pores and the matrix are mutually randomly dispersed implying a random distribution of both phases.³¹ reported that for closed-pore materials, such as the chocolate studied here, the thermal conductivity of the porous material should be between the predictions of MEM-1 and EMT.

$$k_{\text{eff}} = \frac{1}{4} \left\{ 3f(k_g - k_m) + 2k_m - k_f + \sqrt{[3f(k_g - k_m) + (2k_m - k_g)]^2 + 8k_m k_g} \right\} \quad (13)$$

In comparison to the significant number of analytical models predicting the effective thermal conductivity, analytical models for the specific heat capacity are limited to the rule of mixtures, RoM, approximation.^{27,47,48} It is unclear, according to author's knowledge, why there are no other available analytical models for $C_{p,\text{eff}}$ in the literature, whilst for k_{eff} there are plethora of analytical and semi-empirical models. One possible explanation could lie in the fact that most of the processes investigated for composites/porous materials in the literature regard heat transfer at steady state, hence the effects of specific heat capacity become irrelevant. The RoM assumes that the energy required to raise the temperature of a unit mass of the porous material corresponds to the weighted summation of the required energies from the two phases, as shown in eqn (14):

$$C_{p,\text{eff}} = w_g C_{p_g} + (1 - w_g) C_{p_m} \quad (14)$$

The predictions from the analytical models for effective thermal conductivity and effective specific heat capacity will be compared against the results from the micro-FEM.

3. Results

3.1. Experimental results

The experimental results from the thermal rig, described in section 2.2, are summarised in Fig. 5. In this figure the temporal evolution of the thermocouple readings T_1 , T_2 , and T_3 respectively are shown for the scenarios 1 (Fig. 5a), 2 (Fig. 5b), 3



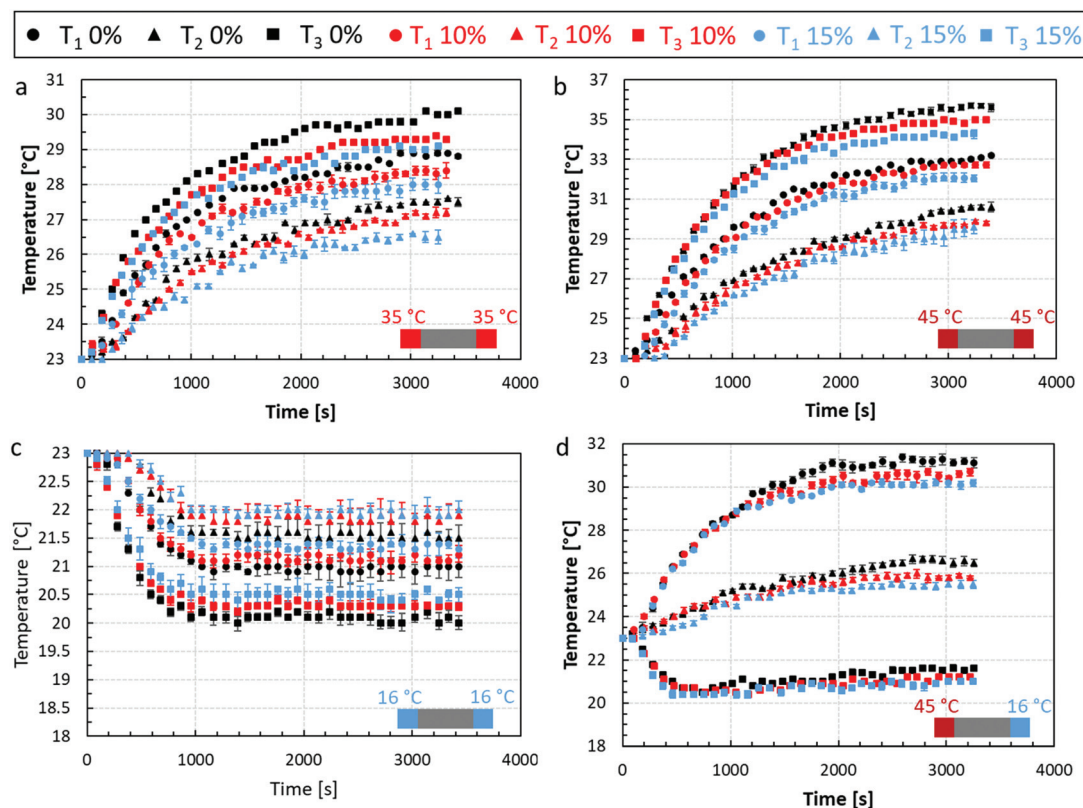


Fig. 5 Temporal temperature evolution curves from heat transfer experiments for three chocolate types at three measurements locations (T_1 , T_2 , and T_3) for: (a) scenario 1; (b) scenario 2; (c) scenario 3; (d) scenario 4 of Table 1. The schematic at the bottom right-hand corner for each graph shows the boundary conditions applied to the test section for each scenario.

(Fig. 5c), and 4 (Fig. 5d). The coloured scatter points represent the temperature response of the non-aerated chocolate (black) and the micro-aerated chocolate at 10% (red) and 15% (blue) porosity, respectively. The error bars in Fig. 5a, b and d are relatively small ($<0.5\%$), whilst those for Fig. 5c are slightly larger ($<1\%$) highlighting the reproducibility of the measurements. A possible source of added uncertainty for the cooling scenario shown in Fig. 5c is that the Peltier elements needed to be cooled using the axial fans, which may have resulted in air disturbance around the thermal rig.

For all four scenarios, the micro-aerated chocolate showed an overall slower heat propagation than the non-aerated chocolate, except for the region corresponding to the first ~ 1000 seconds, when all chocolate exhibited a similar response. The initial transient period before steady state behaviour can be assumed, is affected by both the thermal conductivity of the material and the specific heat capacity. Therefore, no conclusive arguments can be used at this region regarding the heat propagation between the three types of materials. However, when steady state is assumed in each scenario (because of the different applied thermal boundary condition scenarios 1, 2, and 4 equilibrate approximately after 3000 seconds and in scenario 3 after 1200 seconds), the heat transfer response is governed mainly by the thermal conductivity of the material. In addition, this is the region, where the maximum difference

in the heat transfer behaviour is observed amongst the three chocolate products. This means that there is a difference between the values of thermal conductivity for each type of chocolate. This observation suggests that the dispersed micropores in the chocolate suspension may affect heat propagation. This argument will be also evaluated by the micro-FEM presented in the same section.

Based on Fig. 5a, the maximum temperature difference between the curves of the non-aerated chocolate and the 10% micro-aerated chocolate is 2.7% and is recorded by the thermocouple T_2 at the centre of the sample. This difference increases to 4.3% when the non-aerated chocolate is compared to the 15% micro-aeration. For scenario 2, when the sample was heated to 45 °C the same comparison, *i.e.* 0% *vs.* 10% chocolate and 0% *vs.* 15% chocolate, leads to a maximum difference of 4.8% and 6.7% respectively. Similarly, for scenario 3, a maximum difference of 1.9% and 2.4% is observed when the non-aerated chocolate is compared to the micro-aerated chocolate samples of 10% and 15% porosity respectively. For the cooling scenario, Fig. 5c, all chocolate samples reach the steady state faster than in the other scenarios. This is attributed to the fact that the applied temperature *i.e.* 16 °C, is outside the phase change range (see Fig. 6), between 28 °C and 34 °C, of the chocolate and therefore no energy is used to melt the fat crystals.



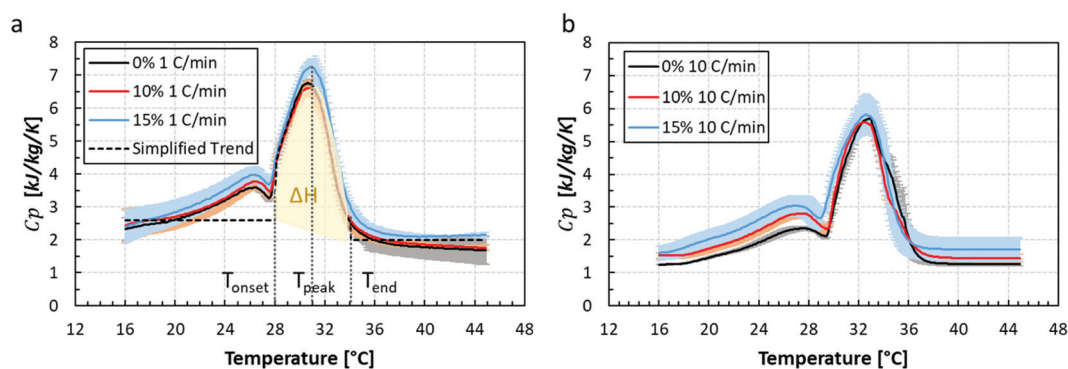


Fig. 6 DSC data for all types of chocolates at two different heating rates, $1\text{ }^{\circ}\text{C min}^{-1}$ (a) and $10\text{ }^{\circ}\text{C min}^{-1}$ (b). The dashed line in (a) shows the simplified version of the C_p trend used as an initial input for the numerical model. The dotted lines indicate the location of T_{onset} , T_{peak} , T_{end} .

Finally, Fig. 5d shows that when cooling one side and heating the other side, the temperature responses at the three locations T_1 , T_2 , and T_3 for the three types of chocolates, have a maximum difference of 0.8%, and 1.4% when the non-aerated chocolate is compared to the micro-aerated chocolates of 10% and 15% porosity respectively. After about 1000 s, the temperature from thermocouple T_3 , which is placed closer to the cooling side of the setup, starts to increase. This was consistent amongst all chocolate samples and was argued to be caused by the higher heat fluxes developed from the heated side, which eventually propagated to location T_3 and increased the temperature. The temperature rise is due to an increased heat flux as the temperature difference between the ambient ($23\text{ }^{\circ}\text{C}$) and the applied temperature at each side of the setup is larger between the heated ($45\text{ }^{\circ}\text{C}$) rather than the cooled ($16\text{ }^{\circ}\text{C}$) side. The maximum errors for all cases are summarised in Table 2.

The specific heat capacity, C_p , extracted from the DSC experiments for all chocolate samples under two constant heating rates of $1\text{ }^{\circ}\text{C min}^{-1}$ (Fig. 6a) and $10\text{ }^{\circ}\text{C min}^{-1}$ (Fig. 6b) are summarised in Fig. 6. The black curves represent the C_p values of the non-aerated chocolate and the red and blue lines the C_p values of the micro-aerated chocolate of 10% and 15% porosity respectively. Comparing the C_p data for the two heating rates, a decrease in the C_p values with heating rate is observed for all chocolate samples. In addition, a decrease in

the melting peak with the rate of heating is shown in Fig. 6, which is consistent with reports from the literature.^{13,14} However, no explanation is provided in these studies to the shift and decrease in C_p values with increasing heating rate. However, this dependency of the C_p data with the rate of heating has been reported for other composite polymers, such as PET, PTT, and PBT.^{49–51} Based on Elleithy *et al.* and Dhandapani, Nayak, and Mohanty, there is an association between crystallisation behaviour of the plastic nanocomposites and the heating rate, as shown by DSC measurements. A higher applied rate of heat transfer impacts the flexibility and the mobility of the polymeric chain of the material and by extend a shorter diffusing time is expected across the chain of molecules. This in turn would cause a faster crystallisation or melting. Similar for chocolates, the restricted mobility of the cocoa fat molecules might cause such shifting in the C_p data as well. The DSC data from the slower rate of heat transfer was eventually used for the models since this was more representative of the heat transfer rate found when both, the chocolate is orally processed as well as tested in the thermal experiments performed.

Based on Fig. 6a, the phase change for all chocolate types occurs between $28\text{ }^{\circ}\text{C}$ (T_{onset}) and $34\text{ }^{\circ}\text{C}$ (T_{end}) and the peak of the curves corresponds to a temperature of $31\text{ }^{\circ}\text{C}$ (T_{peak}). The phase change in the graph of Fig. 6b lies between $30\text{ }^{\circ}\text{C}$ and $36\text{ }^{\circ}\text{C}$ with a peak at $33\text{ }^{\circ}\text{C}$. Based on literature,^{13,14} the melting points of all crystal forms lie between $18\text{ }^{\circ}\text{C}$ and $36\text{ }^{\circ}\text{C}$, thus it can be assumed that the chocolate is solid at temperatures lower than $18\text{ }^{\circ}\text{C}$ and fully melted at temperatures higher than $36\text{ }^{\circ}\text{C}$. Based on this statement, the C_p values can be considered constant outside the $18\text{ }^{\circ}\text{C}$ – $36\text{ }^{\circ}\text{C}$ range for both heating rates. This is found to be true for both figure for temperatures higher than T_{end} whilst a decreasing trend of C_p is observed at temperatures below $20\text{ }^{\circ}\text{C}$. In the data of Fig. 6, a local maximum peak is observed at temperatures between $24\text{ }^{\circ}\text{C}$ and $28\text{ }^{\circ}\text{C}$ implying the presence of fat crystals of other crystallisation form than V form. In this work the effect of different crystal forms of cocoa fat was not considered and the specific heat capacity value which corresponds to $18\text{ }^{\circ}\text{C}$ is notated as $C_{p,\text{solid}}$ and the value of specific heat capacity at $36\text{ }^{\circ}\text{C}$

Table 2 Maximum temperature difference (%) between the non-aerated (0%) and micro-aerated (10% porosity) chocolates, and between the non-aerated (0%) and micro-aerated (15% porosity) chocolates at location T_2 where a maximum temperature difference between the samples is observed. The experimental error was estimated to approximately 1%. The maximum differences in all cases were detected at T_2 location

Scenario	ΔT [0% vs. 10%]	ΔT [0% vs. 5%]
1	2.7%	4.3%
2	4.3%	6.7%
3	1.9%	2.4%
4	0.8%	1.4%



as $C_{p\text{liquid}}$. Therefore, a simplified C_p curve is created as shown by the dashed line in Fig. 6a. The melting parameters such as C_p and melting temperature for both cases are summarised in Tables 3 and 4.

Based on the Tables 3 and 4, the latent heat ΔH remains almost unaffected of porosity whilst the values of $C_{p\text{solid}}$ and $C_{p\text{liquid}}$ slightly increased. Clearer increasing trends of specific heat capacity with porosity are observed when chocolate is at liquid state. This is true for both heating rates. This will be further discussed in the following section. Despite the significant rate dependent behaviour of specific heat capacity observed for all types of chocolates, the heating rate of interest, according to thermal rig experiments was found to be the slowest one, *i.e.* 1°C min^{-1} .

3.2. Numerical and analytical model results

This section is structured as follows; the boundary conditions of the macro-FEM model, which were calibrated by the thermal experiments on aluminium specimens of known materials properties are initially presented. Then, the calibrated thermal conductivity of non-aerated chocolate, using

Table 3 Melting parameters of all chocolate samples obtained from the DSC experiments at 1°C min^{-1}

Sample	$T_{\text{onset}}, ^\circ\text{C}$	$T_{\text{peak}}, ^\circ\text{C}$	$T_{\text{end}}, ^\circ\text{C}$	$\Delta H, \text{kJ kg}^{-1}$	$C_{p\text{solid}}, \text{kJ kg}^{-1} \text{K}^{-1}$	$C_{p\text{liquid}}, \text{kJ kg}^{-1} \text{K}^{-1}$
0%				15	2.6 ± 0.1	2.0 ± 0.3
10%	28	31	34	14	2.7 ± 0.2	2.1 ± 0.1
15%				14	2.8 ± 0.3	2.3 ± 0.1

Table 4 Melting parameters of all chocolate samples obtained from the DSC experiments at $10^\circ\text{C min}^{-1}$

Sample	$T_{\text{onset}}, ^\circ\text{C}$	$T_{\text{peak}}, ^\circ\text{C}$	$T_{\text{end}}, ^\circ\text{C}$	$\Delta H, \text{kJ kg}^{-1}$	$C_{p\text{solid}}, \text{kJ kg}^{-1} \text{K}^{-1}$	$C_{p\text{liquid}}, \text{kJ kg}^{-1} \text{K}^{-1}$
0%				16	1.5 ± 0.0	1.3 ± 0.1
10%	30	33	36	14	1.7 ± 0.1	1.5 ± 0.1
15%				15	2.0 ± 0.3	1.7 ± 0.4

Table 5 Physical and thermal properties of the non-aerated chocolate, cocoa butter and N_2 .⁵² The last column indicates the method used to evaluate each material property

Material properties	Value	Units	Method
Density of chocolate (ρ_{choco})	1300	Kg m^{-3}	Mass/volume
Specific heat capacity of solid chocolate ($C_{p\text{solid}}$)	2600	$\text{J Kg}^{-1} \text{K}^{-1}$	DSC
Thermal conductivity of solid chocolate (k_{solid})	0.45	$\text{W m}^{-1} \text{K}^{-1}$	Macro-FEM
Specific heat capacity of liquid chocolate ($C_{p\text{liquid}}$)	2000	$\text{J Kg}^{-1} \text{K}^{-1}$	DSC
Thermal conductivity of liquid chocolate (k_{liquid})	0.55	$\text{W m}^{-1} \text{K}^{-1}$	Macro-FEM
Latent heat of chocolate (ΔH)	15 000	J Kg^{-1}	DSC
Density of N_2 (ρ_{N_2})	1.16	Kg m^{-3}	
Specific heat capacity of N_2 ($C_{p\text{N}_2}$)	1006	$\text{J Kg}^{-1} \text{K}^{-1}$	Literature
Thermal conductivity of N_2 (k_{N_2})	0.026	$\text{W m}^{-1} \text{K}^{-1}$	
Density of cocoa butter (ρ_{CB})	900	Kg m^{-3}	Mass/volume
Specific heat capacity of solid cocoa butter ($C_{p\text{CB-solid}}$)	6000	$\text{J Kg}^{-1} \text{K}^{-1}$	DSC
Specific heat capacity of liquid cocoa butter ($C_{p\text{CB-liquid}}$)	4500	$\text{J Kg}^{-1} \text{K}^{-1}$	DSC

the macro-FEM model, is shown followed by the presentation of the effective thermal properties from the micro-FEM model and the analytical models, which are described in section 2.4. The comparison of the experimental data of all chocolate samples against the macro-FEM data is shown in the end of the section followed by the discussion of the importance of modelling the melting behaviour of chocolate.

The calibrated boundary conditions from the macro-FEM model and the thermal rig experiments on aluminium specimens corresponded to a heat transfer coefficient of $h_p = 70 \text{ W m}^{-2} \text{K}^{-1}$ at the interface inside the groove between the investigated sample and the aluminium block. This coefficient was found to be independent of loading scenario, *i.e.* 1, 2, 3 or 4. On the other hand, the heat transfer coefficient at the insulated surfaces was $h_{\text{ins}} = 4 \text{ W m}^{-2} \text{K}^{-1}$ for scenario 1, 2, and 4 and $h_{\text{ins}} = 8 \text{ W m}^{-2} \text{K}^{-1}$ for scenario 3. The discrepancy in h_{ins} between scenario 1, 2, and 4, and scenario 3 was attributed to the use of the double axial fans on both sides of the setup of scenario 3, which caused increased airflow disturbances around the thermal rig and increased the heat losses from the test samples to the ambient. When the double axial fans were also used for scenario 1, 2, and 4, the same value of $h_{\text{ins}} = 8 \text{ W m}^{-2} \text{K}^{-1}$ was found. The temperature values for aluminium block – chocolate interface regions were 35°C , 45°C , 16°C depending on the loading scenario as shown in Table 1, whilst the temperature at the insulated sides was the same as the ambient temperature (23°C). The same boundary conditions were also used for the macro-FEM model on chocolates.

The thermal properties of non-aerated chocolate, such as the thermal conductivity, which was calibrated from the macro-FEM model and the thermal rig experiments, the specific heat capacity and latent heat measured from the DSC tests are summarised in Table 5. The thermal conductivity of non-aerated chocolate was found to be $0.45 \text{ W m}^{-1} \text{K}^{-1}$ and $0.55 \text{ W m}^{-1} \text{K}^{-1}$ at the solid and liquid state respectively. These values are relatively within the range reported in the literature.^{13,14,20} In addition, the measured density of the non-aerated chocolate was found to be the same as the one reported in the literature.¹³

The results from the micromechanical FEM (micro-FEM) model of section 2.3 predicting the effective thermal pro-



perties, k_{eff} and $C_{p,\text{eff}}$, together with the predictions from the analytical models as a function of porosity f , as described in section 2.4, are presented in Fig. 7. Each of the four plots of Fig. 7 will be discussed individually below.

In Fig. 7a, the normalised effective thermal conductivity k_{eff} is shown against porosity as predicted by the micro-FEM steady state and the analytical models. It is evident that the analytical models, such as the EMT and MEM-1, agree with the prediction from the FE-model. This finding is also in agreement with the literature for closed pore materials³¹ acting as independent validation of the micro-FEM presented in this study. Based on Fig. 7a, decreasing trends of the thermal conductivity are observed with porosity. Specifically, a decrease in thermal conductivity of the chocolate by 14% and 21% is observed for porosities 10% and 15% respectively.

These results agree with the experimental observations of section 3.1. It is evident, as discussed in section 3.1 and the reported temperature differences amongst all types of chocolate at steady-state that micro-pores act as thermal barriers to the heat flow leading to a decrease in the values of the thermal conductivity with increasing porosity. These findings are also in agreement with reports in the literature for other porous materials. Specifically, Smith *et al.*,⁴⁴ Tavman,⁴⁵ and Skibinski *et al.*⁴⁶ reported decreasing trends with increasing porosity in materials, such as porous ceramics, granular materials, and porous copper. Similar reports were made by Carson *et al.*³¹ on porous foods, such as granular starch, milk powder, gels and cakes.

In Fig. 7b, the predictions of the specific heat capacity of chocolate against porosity are shown from the RoM analytical model and the micro-FEM. The results are in good agreement, suggesting that the specific heat capacity remains approximately the same at the investigated porosity range, *i.e.* less than 20% porosity. Specifically, the FE-model predicts a difference of 1% and 1.5% when the specific heat capacity of the non-aerated chocolate is compared to the specific heat capacity of the micro-aerated chocolate samples of 10% and 15% porosity respectively. Similar results have been reported for other porous materials.⁴⁹ Based on these studies, weaker decreasing trends with increasing porosity were observed in the specific heat capacity values compared to the thermal conductivity for concrete and aluminium alloys.

The C_p data from the DSC experiments at a constant heating rate of $1\text{ }^{\circ}\text{C min}^{-1}$, see Table 1, were also compared to the micro-FEM predictions of Fig. 7b. The comparison between the numerical model and the DSC experiments was limited to the slowest heating rate, because this rate was relevant to the heat transfer experiments of section 3.1, *i.e.* maximum heating rate of $0.7\text{ }^{\circ}\text{C min}^{-1}$. Comparing the C_p data of the 10% and 15% micro-aerated chocolates obtained from the DSC experiments, see Table 3, to the predicted C_p values from the numerical models a significant difference is observed. This is caused by the fact that the DSC tests show a slight increase of the C_p , whilst for the numerical and the analytical models the specific heat capacity remains unaffected by porosity (an increase of 1% was estimated). Specifically, an

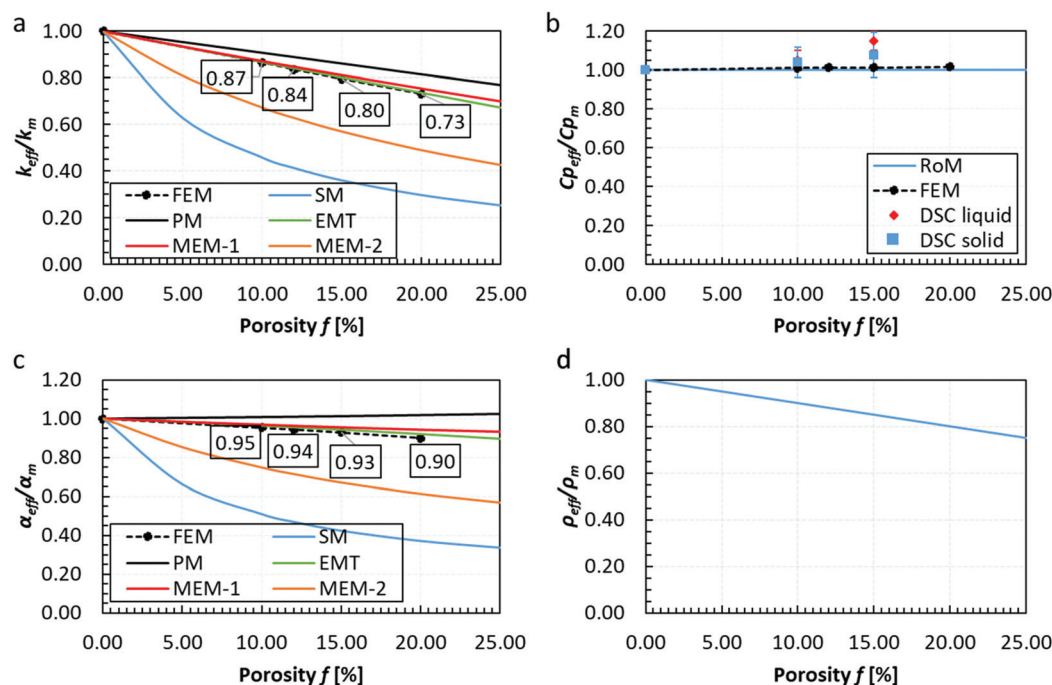


Fig. 7 Results from the micromechanical heat transfer model showing the normalised effective (a) thermal conductivity, (b) specific heat capacity, (c) thermal diffusivity and (d) density of chocolate with respect to the respective property of the matrix (solid chocolate) as a function of porosity. No experimental data for thermal conductivity are shown because this property cannot directly be obtained by either the thermal rig experiments or the DSC tests.



increase of 4% and 5% in the values of C_p is observed between the non-aerated chocolate and the micro-aerated chocolate of 10% porosity at the solid and liquid state respectively, see Table 3. Similarly, for the micro-aerated chocolate of 15% porosity, a small increase in the average value of the specific heat capacity at the regions where the chocolate is solid (7%) and liquid (13%) is shown, see Table 3. It seems that there is an increasing trend of C_p with increasing porosity, which is unexpected for porous materials. Both the analytical models and the micro-FEM in this case do not show the same trends (micro-FEM only suggests a maximum of 1.5% increase).

One possible explanation of the increase of the specific heat capacity with increasing porosity from the DSC tests could potentially be a mechanism that causes an unexpected temperature change at the chocolate matrix and the pore interface that has not been captured by the simulation. This could for example be in the form of interfacial resistance between the pore and chocolate matrix or due to the presence of a third phase within the chocolate itself. In our previous work,⁸ a third “unknown” phase is reported at the interface of the chocolate matrix and the pore as evidenced by X-ray tomography images on micro-aerated chocolate of 15% porosity, which could be responsible for changes in the bulk thermal properties, such as the specific heat capacity observed by the DSC measurements.

The thermal diffusivity a_{eff} of chocolate calculated from eqn (8) using different expressions for the effective thermal conductivity (see eqn (9)–(13)) and the RoM for the specific heat capacity (eqn (14)) are plotted against porosity in Fig. 7c. Based on the same graph, the EMT predicts most accurately the values of a_{eff} amongst all other models based on the micro-FEM predictions. Based on Fig. 7c, the thermal diffusivity decreases with increasing porosity implying that a slower heat propagation is experienced in the case of the micro-aerated chocolate samples, when the samples are either heated or cooled. Specifically, based on the results of Fig. 7c, an average decrease of 5% in thermal diffusivity for the 10% micro-aerated chocolate and 7% decrease for the 15% micro-aerated chocolate are estimated. Similar findings were reported for concrete⁴⁸ and porous aluminium alloys⁵⁶. The mass density values against porosity using eqn (7) are shown in Fig. 7d. Based on this figure, a linear decrease of the mass density of the micro-aerated chocolate with increasing porosity is estimated, as expected for porous materials.

The temperature profiles recorded by the thermocouples at locations T_1 , T_2 , and T_3 for all four scenarios and all types of chocolate are shown in Fig. 8, together with the predictions of the macro-FEM using the effective thermal properties presented in Fig. 7 from the micro-FEM. Overall, a reasonable agreement between the results from the macro-FEM model and the experimental results is observed for all types of chocolate and all four scenarios. The macro-FEM predicts accurately both the shape of the temperature recordings from the three thermocouples considering the melting behaviour of the chocolate but also the temperature of all chocolate materials during the unsteady and steady state regime. A slight overesti-

mation of the temperature profiles in the case of the two micro-aerated chocolates was observed, which was attributed to microstructural findings reported in the literature⁸ at the pore-chocolate interface, as mentioned earlier in the section. As evidenced by Bikos *et al.*,⁸ there are unknown microscopic features at the chocolate-pore interface, which create a shell around the insulating gas pore. Assuming these regions are mainly or solely cocoa butter concentrated around the pore, the material at the pore interface is capable to store thermal energy due to the higher specific heat capacity of the cocoa butter compared to chocolate, as shown in Table 5, which could thereafter explain the increase in heat capacity of the material and eventual slower heat penetration observed for the aerated vs. non aerated samples. The possible presence of a cocoa butter shell around the otherwise low thermal capacity and thermal conductivity gas pore could promote heat storage within the shell. This could also result to significant discrepancies, which should not be neglected, close to the interface and in turn affect significantly the effective thermal properties and, by extent, the thermal processes for porous chocolates (such as the cooling time during manufacturing or melting time during oral processing). This hypothesis requires further investigation both, in terms of understanding the origin and nature of the material concentrated around the pore and its impact on the effective thermal properties of chocolates.

However, recent studies have shown that TAG polymorphs of high melting point, which are the same as the ones found in cocoa butter, are formed on the surface of a bubble creating a crystal coating⁵⁴ around the bubble when crystallised. These studies have been performed on various edible oils, such as vegetable and rapeseed oil. In addition, similar evidence has been reported in whisked cocoa butter, where laminar shaped fat crystals are formed on the surface of the bubble.⁵⁵ It is argued that it might be oriented parallel to the oil-fat interface to reduce the surface energy in that region.

Based on the results of Fig. 8a, the macro-FEM predicts reasonably accurate the temperature profiles at T_1 , T_2 , and T_3 for all chocolate materials as evidenced by the thermal experiments. In the case of the non-aerated chocolate, a slight underestimation in temperature values is observed at location T_3 with a maximum difference less than 2%, while at locations T_1 and T_3 slightly higher maximum differences are depicted of 3.6% and 4%. Similar trends are observed for the non-aerated and the micro-aerated chocolates of 10% and 15% porosity.

The temporal evolution of the temperature for scenario 2 (fast heating) at locations T_1 , T_2 , and T_3 for all chocolate samples is shown in Fig. 8b. The maximum differences between the experimental readings and the numerical predictions increase to a maximum of 4.6%, 3.7%, and 4.8% for the non-aerated and micro-aerated chocolates of 10% and 15% porosity respectively.

For scenario 3, where the temperature profile is outside the melting range of the chocolate (see Fig. 8c), higher accuracy predictions are presented for all chocolate products. Specifically, a maximum difference of less than 2.5% is indicated in the case of the non-aerated chocolate, whilst a



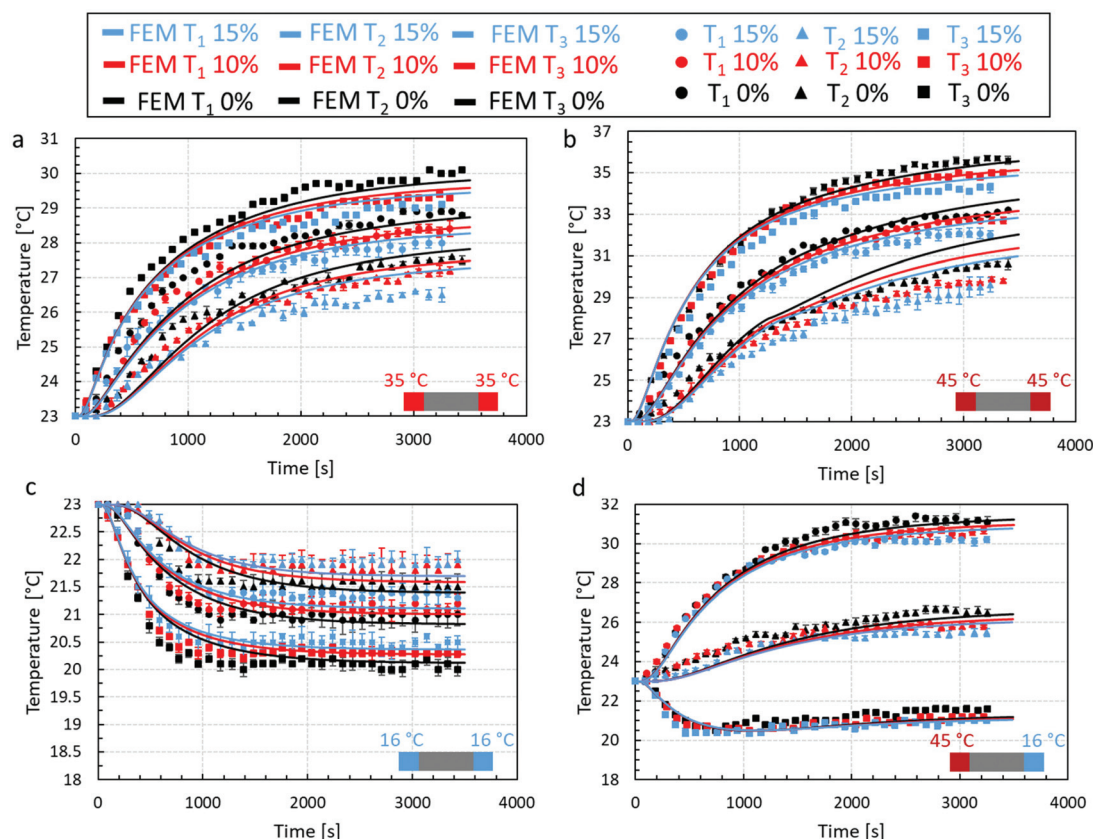


Fig. 8 Temporal temperature evolution predicted from the heat transfer macro-FEM compared to the heat transfer experiments for three chocolate porosities. The images show macro-FEM predictions for (a) scenario 1, (b) scenario 2, (c) scenario 3, and (d) scenario 4.

maximum 1.8% and 1.4% difference between the experimental and macro-FEM results are observed for the micro-aerated cases. Since no phase change occurs, the steady state is reached relatively fast after approximately 2000 seconds.

Although scenario 4 (see Fig. 8d), where the chocolate sample was heated from one side and cooled from the other, produced a more complex temperature profile across the samples, the macro-FEM estimates the temperature evolution for all three locations relatively accurate. In fact, the macro-FEM captured the increasing trends of the experimental data at location T_3 . Finally, the maximum differences between experiments and computations are within 3.6%, 2.4% and 3.3% for non-aerated chocolate and micro-aerated chocolate samples of 10% and 15% porosity respectively.

To assess the quality of the model predictions, a comparison between the rate of heat transfer, $\frac{dT}{dt}$, measured from the heat transfer experiments using the thermal rig and the macro-FEM for all chocolate samples for scenario 4 is performed, as shown in Fig. 9. The $\frac{dT}{dt}$ data obtained from the thermocouple measurements were averaged across twenty points, *i.e.* every 50 s, to minimise any random error from the thermocouple measurements and are shown Fig. 9 for all three materials. Based on the data of Fig. 9, a relatively good agree-

ment between experiments and computations is observed, especially in predicting the shapes of the curves for all cases. Most changes occur in the first 1000 seconds and then both the experimental and numerical results approached zero rates. The largest uncertainties in the heat transfer rate were found at the early stages of the experiment ($t = 250$ s), as indicated in Fig. 9, mainly due to the temperature measurement uncertainty. This uncertainty is caused by the limited temperature resolution of the thermocouple measurements.

So far, the micro-FEM, experiments and analytical models have shown that micro-aeration significantly affects the thermal conductivity. It is believed that the role of micro-pores is to act as barriers to heat transfer, as evidenced by the decrease of thermal conductivity with increasing porosity. The predictions from the micro-FEM and the EMT analytical model suggest a weak or no dependency of the specific heat capacity on the porosity. However, the both the thermal rig and the DSC experiments implied increasing trends of specific heat capacity with increasing porosity, which is unexpected for porous materials.^{47,53} This may be explained by the presence of a third phase or a thermal mechanism, which increases the averaged specific heat capacity of the micro-porous chocolate. This mechanism cannot be solely explained by the presence of N_2 in the pores due to the relatively low specific heat capacity and significantly lower density of N_2 which means that the



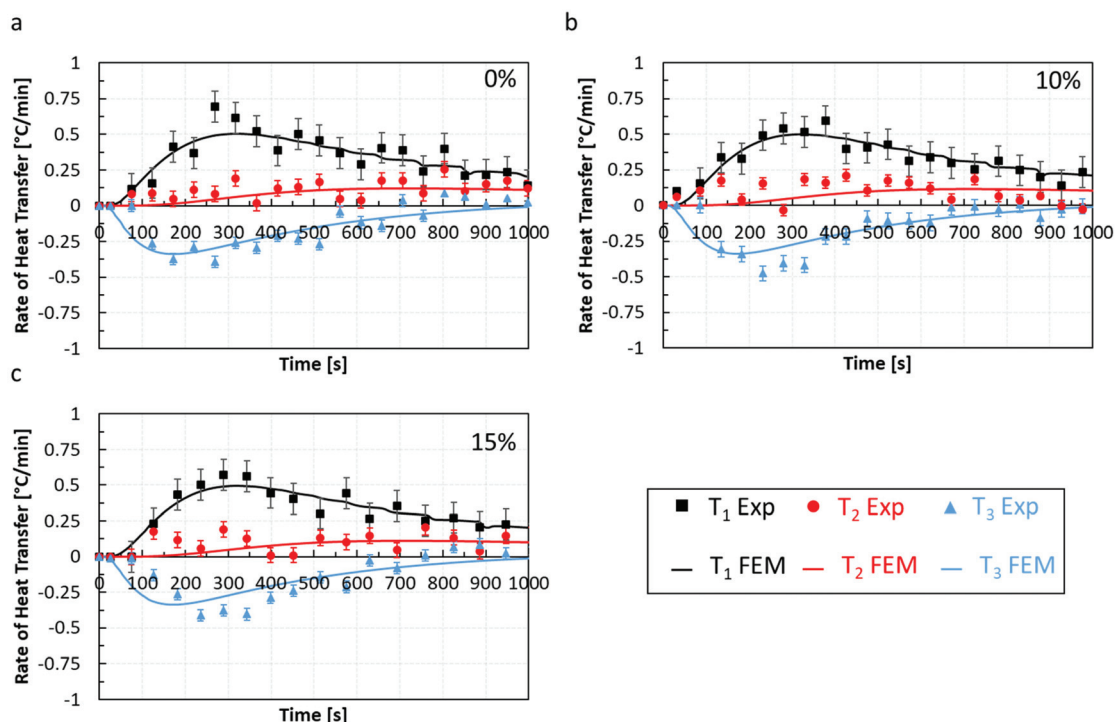


Fig. 9 Temporal evolution of the rate of heat transfer from the heat transfer experiments at locations T_1 , T_2 , and T_3 is compared against the macro-FEM predictions for the same locations for the (a) non aerated chocolate, (b) micro-aerated of 10% porosity, and (c) the micro-aerated of 15% porosity. The error bars depict the uncertainty of the temperature measurements.

pores are not capable to store substantial energy. However, the presence of a cocoa butter shell around the pores could be responsible for such effects. Cocoa butter stores a higher amount of energy, as evidenced by its C_p (see Table 5), whilst recent studies^{54,55} have shown concentration of fat crystals on the interface of the chocolate matrix and the pore.

All these phenomena will play a determinant role in oral food processing. According to literature,^{12,15,18} the melting behaviour of chocolate affects a plethora of sensory properties, such as melting time, moistness, grittiness and overall flavour intensity. Melting is an important step of chocolate's oral processing transforming the continuous fat system into an oil-water suspension in presence of saliva, which in turn is linked with taste and flavour perception. It has been reported that melting is one of the main mechanisms which drives masticatory parameters such as the required number of chews and the chewing time.¹² Our recent work has highlighted the need for coupling the mastication step with the thermal behaviour of chocolate in order to draw conclusive arguments regarding the chocolate's breakdown during oral processing.³⁵ It is believed these processes together guide the chocolate's transition from solid to liquid phase and perhaps the final properties of the bolus. Understanding the effect of micro-aeration on the melting behaviour of chocolate will assist the design of food products with optimum sensorial properties.³⁵

This work has presented a novel, and relatively inexpensive, experimental procedure, which can be used to accurately assess the effects of microscale structural changes in

materials, such as low levels of porosity, to the macroscopic heat transfer behaviour of the material. The advantage of this setup over commercially available instruments is the ability of the method to accurately record the temperature response of complex materials under small temperature differences, something which is not possible in most commercial instruments that usually operate at higher temperature ranges.

A multiscale finite element model is proposed, which is calibrated using results obtained on the thermal rig. Effective material properties were validated using analytical models from literature and compared to the heat transfer measurements. The specific heat capacity values were also compared to DSC experiments. Combined, the experimental setup and the new multiscale model could provide a powerful tool to investigate the effects of material microstructure on the thermal properties of foods and optimise their thermal behaviour for various applications. One potential application is to use this model to optimise the temporal evolution of the melting behaviour of chocolate, porous or otherwise, and, for example, modify sweetness perception, while reducing sugar content, as required when developing foods with improved health characteristics.³⁵ Future work will include the extension of the multiscale model to consider the presence of a third phase at the pore's interface between the chocolate matrix and the gas pore. The current suggestion that the third phase is in fact cocoa butter will be evaluated. The developed multiscale model can be expanded, in the future, to study the heat transfer response of porous chocolates at



higher porosity levels than 15% and pore sizes higher than 40 μm .

4. Conclusions

Heat transfer experiments were performed on chocolate specimens with varying levels of porosity, using four different thermal loading scenarios. These experiments were accompanied by a multiscale Finite Element Model (FEM) to estimate the thermal response of chocolate. The main findings are summarised below:

1. Micro-FEM steady-state predictions of the effective thermal conductivity of chocolate overlap with analytical models, such as Maxwell Eucken and effective medium theory.
2. The increasing trends of specific heat capacity with increasing porosity, as measured by DSC tests for porous materials, are opposed to the analytical and micro-FEM unsteady predictions. This observation suggests that a novel mechanism exists in the porous microstructure, which is able to store thermal energy. A concentration of material (as previously shown from XRT images⁸) at the pore's interface exists exhibiting higher specific heat capacity than chocolate. This material is believed to be cocoa butter, which exhibits approximately 1.3 times higher specific heat capacity than the non-aerated chocolate.
3. The developed multiscale model predictions are in good agreement (within 5%) with the measurements from the newly developed heat transfer experimental rig.
4. Compared to the non-aerated chocolate, the measurements showed that a 15% micro-aeration leads to an average reduction of 21% in thermal conductivity and increase of 21% in specific heat capacity. This results to a maximum 7% reduction in the rate of heat transfer.

This work assessed a method to quantify the change in melting behaviour of different types of chocolate with varying porosity, which in turn, can be used to design food structures with controlled sensorial attributes such as melting time, grittiness, flavour intensity, and sweetness.

Nomenclature

A	Cross sectional area (m^2)
α	Thermal diffusivity ($\text{m}^2 \text{s}^{-1}$)
α_{eff}	Effective thermal diffusivity ($\text{m}^2 \text{s}^{-1}$)
C_p	Specific heat capacity ($\text{J kg}^{-1} \text{K}^{-1}$)
$C_{\text{PCB-liquid}}$	Specific heat capacity of cocoa butter at liquid state ($\text{J kg}^{-1} \text{K}^{-1}$)
$C_{\text{PCB-solid}}$	Specific heat capacity of cocoa butter at solid state ($\text{J kg}^{-1} \text{K}^{-1}$)
C_{Peff}	Effective specific heat capacity ($\text{J kg}^{-1} \text{K}^{-1}$)
C_{Pg}	Specific heat capacity of pore ($\text{J kg}^{-1} \text{K}^{-1}$)
C_{Pliquid}	Specific heat capacity at liquid state ($\text{J kg}^{-1} \text{K}^{-1}$)
C_{Pm}	Specific heat capacity of matrix, chocolate, ($\text{J kg}^{-1} \text{K}^{-1}$)

C_{PN_2}	Specific heat capacity of N_2 ($\text{J kg}^{-1} \text{K}^{-1}$)
C_{Psolid}	Specific heat capacity at solid state ($\text{J kg}^{-1} \text{K}^{-1}$)
$\frac{dT}{dt}$	Temperature rate ($^{\circ}\text{C s}^{-1}$)
f	Porosity (-)
H	Height (m)
k	Thermal conductivity ($\text{W m}^{-1} \text{K}^{-1}$)
k_{eff}	Effective thermal conductivity ($\text{W m}^{-1} \text{K}^{-1}$)
k_g	Thermal conductivity of the pore ($\text{W m}^{-1} \text{K}^{-1}$)
k_{liquid}	Thermal conductivity at liquid state ($\text{W m}^{-1} \text{K}^{-1}$)
k_m	Thermal conductivity of the matrix (chocolate) ($\text{W m}^{-1} \text{K}^{-1}$)
k_{N_2}	Thermal conductivity of N_2 ($\text{W m}^{-1} \text{K}^{-1}$)
k_{solid}	Thermal conductivity at solid state ($\text{W m}^{-1} \text{K}^{-1}$)
L	Length (m)
m	Mass (kg)
\dot{Q}	Heat flow rate (W)
Q_i	Heat vector (j) $i = 1, 2, 3$
$Q _{x_i}$	Heat in direction x_i (J)
t	Thickness of the RVE (m)
T_i	Temperature ($^{\circ}\text{C}$) at location i (where $i = 1, 2, 3, \text{A, B}$)
T_{onset}	Temperature at the onset of phase change ($^{\circ}\text{C}$)
T_{peak}	Temperature at the peak of phase change ($^{\circ}\text{C}$)
T_{end}	Temperature at the end of phase change ($^{\circ}\text{C}$)
V	Volume of RVE (m^3)
W	Width (m)
w_g	Weight fraction of the pore
ΔH	Latent heat (J kg^{-1})
ΔT	Temperature difference ($^{\circ}\text{C}$)
ρ	Density (kg m^{-3})
ρ_{choco}	Density of chocolate (kg m^{-3})
ρ_{CB}	Density of cocoa butter (kg m^{-3})
ρ_{eff}	Effective density (kg m^{-3})
ρ_g	Density of the pore (kg m^{-3})
ρ_m	Density of the matrix, chocolate, (kg m^{-3})
ρ_{N_2}	Density of N_2 (kg m^{-3})

Abbreviations

DPDT	Double pole double throw
DSC	Differential scanning calorimetry
EMT	Effective medium theory
FEM	Finite element model
HWM	Hot wire method
LFM	Laser flash method
macro-FEM	Macromechanical/macroscale finite element model
MEM	Maxwell Eucken model
micro-FEM	Micromechanical/microscale finite element model
PID	Proportional integral derivative
PM	Parallel model
RoM	Rule of mixtures
RVE	Representative volume element
SM	Series model



Conflicts of interest

There are no conflicts to declare.

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