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ON 3D DYNAMIC CONTROL OF SECONDARY COOLING IN CONTINUOUS CASTING PROCESS

ABSTRACT. In this paper a 3D-model for simulation and dynamic control of the continuous casting process is presented. The diffusion convection equation with multiphase transition is used as a simulation model. The developed model is discretized by finite element method and the algebraic equations are solved using pointwise relaxation method. Two different type of methods are used to control the secondary cooling, namely PID and optimal control method. The numerical results are presented and analyzed.

1. INTRODUCTION

In the continuous casting process the molten steel is poured into a bottomless mold which is cooled with internal water flow. The cooling in the mold extracts heat from the molten steel and initiates the formation of the solid shell. The shell formation is essential for the support of the slab after mold exit. After the mold the slab enters into the secondary cooling area in which it is cooled by water mist sprays. The secondary cooling region is usually divided into cooling zones in which the amount of the cooling water can be controlled separately.

The control of cast cooling is of central importance in continuous casting process because it has a considerable influence on formation of cracks and other defects which can be formed in cast material. The cast should be cooled down according to a certain temperature field which depend on e.g. steel quality, cast shape, casting speed. Accurate knowledge of temperature field and liquid pool length is also important especially when using soft reduction or "near final shape" casting. Many numerical models for simulation of the casting process have been developed in recent years [6, 9, 10]. Some of them have been applied to control and optimize casting process [4, 5, 11]. To our knowledge, all real-time industrial control applications are based on two dimensional models. In many cases two dimensional models are sufficient for control purpose. However,

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nowadays the highly automated and instrumented casting machines in steel factories allows the use of more sophisticated simulation and control models.

Our aim is to use the developed 3D-model in dynamic process control. We have considered PID and optimal control methods to control the secondary cooling. The two methods are very different from each other. The PID algorithm is quite simple and computationally inexpensive. However, the use of PID method is limited to the control of the surface temperature. Moreover, the PID algorithm contains experimental tuning parameters. On the other hand the optimal control method can be very complicated. The optimal control method minimizes a cost function which is constructed by means of metallurgical cooling criteria. Several different cooling criteria can be used, e.g. the maximum length for the liquid pool, the maximum reheating and cooling rates on the slab surface, the minimum and maximum temperature at the unbending point and the maximum and minimum temperature on the surface along the casting machine [6]. The development work of our optimal control model is still ongoing. Therefore we use quite simple cost function in our numerical example.

2. MATHEMATICAL FORMULATION OF STATE PROBLEM

Let $\Omega \subset R^2$ be a rectangular domain $[0, L_X] \times [0, L_Y]$ and $\mathcal{V} = \Omega \times [0, L_Z]$ be a 3D domain. The boundary $\Gamma = \partial\mathcal{V}$ consists of the parts:

$$\begin{aligned}\Gamma_0 &= \Omega \times \{0\}, \\ \Gamma_N &= \{(x, y) \in \partial\Omega : x = 0 \vee y = 0\} \times [L_M, L_Z], \\ \Gamma_S &= \{(x, y) \in \partial\Omega : x \neq 0 \wedge y \neq 0\} \times [0, L_Z) \cup \Omega \times \{L_Z\}, \\ \Gamma_M &= \{(x, y) \in \partial\Omega : x = 0 \vee y = 0\} \times [0, L_M],\end{aligned}$$

where L_M is the length of the mould, L_Z is the length of the strand and L_X, L_Y are the width and thickness of the calculation domain.

We define the temperature $T = T(x, y, z, t)$, dependent enthalpy function $H(T)$ and the Kirchoff's temperature $K(T)$ by

$$H(T) = \rho \left(\int_0^T c(\xi) d\xi + L(1 - f_s(T)) \right), \quad K(T) = \int_0^T k(\xi) d\xi,$$

where $\rho, c(T), k(T)$, and L are density, specific heat, thermal conductivity and latent heat, $f_s(T)$ is the solid fraction.

The mathematical model of the continuous casting problem can be written as:

$$\left\{ \begin{array}{ll} \frac{\partial H(T)}{\partial t} + v \frac{\partial H(T)}{\partial z} - \Delta K(T) = 0 & \text{in } \mathcal{V} \times (0, t_f], \\ T = T_0 & \text{on } \Gamma_0 \times (0, t_f], \\ \frac{\partial K(T)}{\partial n} + h(T - T_w) + \sigma \epsilon (T^4 - T_{ext}^4) = 0 & \text{on } \Gamma_N \times (0, t_f], \\ \frac{\partial K(T)}{\partial n} = 0 & \text{on } \Gamma_S \times (0, t_f], \\ \frac{\partial K(T)}{\partial n} = Q & \text{on } \Gamma_M \times (0, t_f], \\ T(x, y, z, 0) = T^0 & \text{in } \mathcal{V}. \end{array} \right. \quad (1)$$

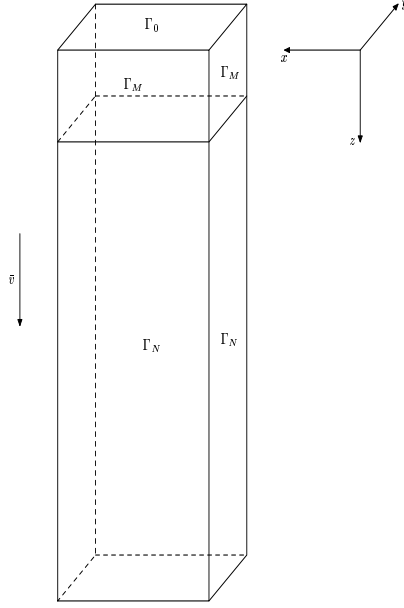
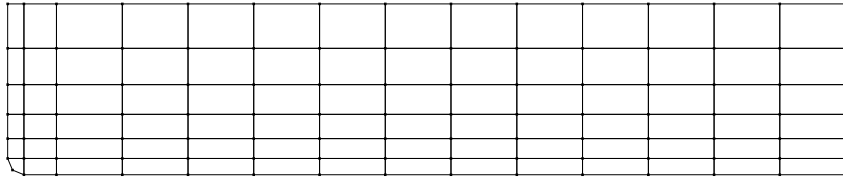


FIGURE 1. The calculation domain.

Here n is the unit vector of outward normal to $\partial\mathcal{V}$, h is the heat transfer coefficient, v is the casting speed, and T_w, T_{ext} are known temperatures. The σ is the Stefan-Boltzmann constant and ϵ is the emissivity. The cooling capacity Q in the mould is known constant and t_f is the simulation time.

3. MESH APPROXIMATION

We partition Ω into a set of quadrilateral finite elements as shown on figure 2. Step sizes are smaller near left and down boundaries of Ω , and all quadrilaterals

FIGURE 2. Partition of Ω on quadrilateral finite elements.

are rectangular except those elements near the round corner. The 1D-domain $[0, L_Z]$ is divided by n_z mesh points in z -direction. We decompose our 3D domain \mathcal{V} into a set of prismatic 3D finite elements with quadrilaterals in its cross section. Since the constructed mesh is Cartesian product of 2D mesh in Ω and 1D mesh on $[0, L_X]$, basis functions of finite element space will also be a product of 2D and 1D basis functions.

We discretize our problem in time by using semi-implicit approximation, more precisely, we approximate the term $\left(\frac{\partial}{\partial t} + v(t)\frac{\partial}{\partial z}\right)H$ by using the characteristics of the first order differential operator [2, 7, 8]. Namely, if (x, y, z, t)

is the mesh point on the time level t we choose $\tilde{z} = z - \int_{t-\tau}^t v(\xi) d\xi$ and approximate this term by:

$$\left(\frac{\partial}{\partial t} + v(t) \frac{\partial}{\partial z} \right) H \approx \frac{1}{\tau} (H(x, y, z, t) - H(x, y, \tilde{z}, t - \tau)). \quad (2)$$

We denote by $\tilde{H}(x, t - \tau) = H(x, y, \tilde{z}, t - \tau)$. If $\tilde{z} < 0$ then we set $\tilde{H}(x, t - \tau) = H(x, y, 0, t - \tau)$.

The weak solution of the semi-discrete problem is defined by integral identity

$$\begin{aligned} \int_{\mathcal{V}} (H(T) - \tilde{H}) / \tau \eta dx &+ \int_{\mathcal{V}} \nabla K(T) \cdot \nabla \eta dx + \int_{\Gamma_N} \{h(T - T_w) \\ &+ \sigma \epsilon (T^4 - T_{ext}^4)\} \eta d\Gamma_N + \int_{\Gamma_M} Q \eta d\Gamma_M = 0 \end{aligned} \quad (3)$$

for all trial functions $\eta \in V^0 = \{\eta \in H^1(\mathcal{V}) : \eta = 0 \text{ on } \Gamma_0\}$. Here $H^1(\mathcal{V})$ is the Sobolev space of first order and $T = T(x) = T(x, y, z, j\tau)$, $j \geq 1$.

Let V_h be a finite element approximation of the space $H^1(\mathcal{V})$, $V_h^{T_0} = \{v_h \in V_h(\mathcal{V}) : v_h = T_{0h} \text{ on } \Gamma_0\}$ with T_{0h} being the V_h -interpolant of T_0 and obviously defined subspace V_h^0 . Let further $\varphi_1(x), \dots, \varphi_n(x)$, be the basis functions in V_h . Thus, the function

$$T_h(x) = \sum_{i=1}^n T_i \varphi_i(x)$$

is the finite element approximation of $T(x)$, $T = (T_1, \dots, T_n)^t$ is the vector of the unknown nodal values of the function $T_h(x)$. Below the notations are: $H(T) = (H(T_1), \dots, H(T_n))^t$, $K(T) = (K(T_1), \dots, K(T_n))^t$ and

$$f_h(T_h) = \sum_{i=1}^n f(T_i) \varphi_i(x)$$

for every function $f(T)$, which depends on $T(x)$.

Approximation of the semi-discrete problem (3) by a finite element scheme is defined as

$$\begin{aligned} \int_{\mathcal{V}} \frac{H_h(T_h) - \tilde{H}_h}{\tau} \eta dx &+ \int_{\mathcal{V}} \nabla K_h(T_h) \cdot \nabla \eta dx + \int_{\Gamma_N} \{h(T_h - T_{wh}) \\ &+ \sigma \epsilon ((T^4)_h - (T_{ext}^4)_h)\} \eta d\Gamma_N + \int_{\Gamma_M} Q_h \eta d\Gamma_M = 0 \end{aligned} \quad (4)$$

for all trial functions $\eta \in V_h^0$. Discrete problem (4) is equivalent to the system of nonlinear algebraic equations:

$$M \frac{H(T) - \tilde{H}}{\tau} + AK(T) + B(h)(T - T_w) + B(T^4 - T_{ext}^4) + DQ = 0, \quad (5)$$

where M , A , $B(h)$, B and D are the $n \times n$ matrices with entries

$$\begin{aligned} m_{ij} &= \int_{\mathcal{V}} \varphi_j \varphi_i dx, & a_{ij} &= \int_{\mathcal{V}} \nabla \varphi_j \cdot \nabla \varphi_i dx, \\ b_{ij}(h) &= \int_{\Gamma_N} h \varphi_j \varphi_i d\Gamma_N, & b_{ij} &= \sigma \epsilon \int_{\Gamma_N} \varphi_j \varphi_i d\Gamma_N, \\ d_{ij} &= \int_{\Gamma_M} \varphi_j \varphi_i d\Gamma_M. \end{aligned}$$

More precisely, let $L(l, k)$ and $K(s, t)$ be the global indices of any two mesh points, where k and t indicate the node number in z-direction and l, s in xy-plane, respectively. By definition of mass matrix M we have

$$m_{LK} = \int_{\mathcal{V}} \varphi_L \varphi_K dx = \left(\int_0^{L_z} \varphi_k(z) \varphi_t(z) dz \right) \int_{\Omega_h} \varphi_l(x, y) \varphi_s(x, y) dx dy = m_{kt}^z m_{ls}^\Omega.$$

Therefore matrix M is equal to Kronecker product of standard 1D mass matrix M^z and 2D mass matrix M^Ω . Similarly we can define for the stiffness matrix A the entries

$$a_{LK} = m_{kt}^z a_{ls}^\Omega + a_{kt}^z m_{ls}^\Omega$$

and for matrices B , $B(h)$ and D

$$\begin{aligned} b_{LK} &= \sigma \epsilon m_{kt}^z m_{ls}^b, \\ b_{LK}(h) &= h_k m_{kt}^z m_{ls}^b, \\ d_{LK} &= m_{kt}^z m_{ls}^b. \end{aligned}$$

Temperature dependent enthalpy H and Kirchoff transform K are defined as piecewise linear and continuous functions. For the solution of system (5) the modified SOR-method [3] for points in \mathcal{V} is used and for nonlinear part on Γ_N Newton-Raphson method is used with one inner iteration.

We note that to implement our method it is sufficient to construct only 2D- and 1D-matrices. Therefore the computational efficiency of our model is very good. Also the memory allocation requirements are not so strong than in the case of ordinary 3D-brick elements.

4. CONTROL OF SECONDARY COOLING PROCESS

The secondary cooling region is divided into cooling zones (see figure 3) in which the values of heat transfer coefficients, h_i , can be controlled separately in each cooling zone. Two different control methods are used for optimizing the secondary cooling process on the boundary of the steel slab, namely PID and optimal control method.

4.1. PID control. In PID control the water cooling in each secondary cooling zone is done separately and indepentendly of each other. Heat transfer coefficient in each cooling zone i are calculated using the equation

$$h_i(t) = P \left[\Delta T_i(t) + I \int_0^t \Delta T_i(s) ds + D \frac{d\Delta T_i(t)}{dt} \right], \quad (6)$$

where $\Delta T_i(t) = T_i(t) - T_i^{tar}(t)$ is the difference between calculated temperature T and target temperature T^{tar} at time t and P, I, D are experimentally known

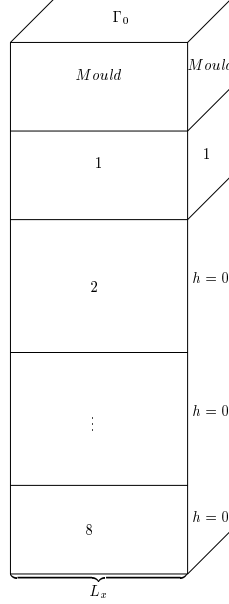


FIGURE 3. Schematic presentation of 3D-cooling zones for quarter of the slab

tuning parameters. After discretization of integral and derivative we obtain the recursive form [1]

$$h_i^k = h_i^{k-1} + P \left[\left(1 + \frac{D}{\tau} \right) \Delta T_i^k - \left(1 - \tau I + \frac{2D}{\tau} \right) \Delta T_i^{k-1} + \frac{D}{\tau} \Delta T_i^{k-2} \right], \quad (7)$$

with constant time step τ .

4.2. Optimal control. In optimal control method our aim is to minimize a cost function which is constructed by means of metallurgical cooling criteria. We can formulate our optimal control problem in the following way:

Find $h^* \in U_{ad}$ such that

$$J(T(h^*)) = \min_{h \in U_{ad}} J(T(h)), \quad (8)$$

where $U_{ad} = \{h(t) | h_{min} \leq h(t) \leq h_{max}\}$ and $T(h)$ is the solution of the state equation (5).

Our cost function, J , has the form

$$J(T) = \frac{1}{2} \int_{L_M}^{L_Z} (T - T^{tar})^2 dz.$$

To solve the optimal control problem (8) we use the following gradient method. For given initial guess $h^0 \in U_{ad}$ we find $h^{i+1} \in U_{ad}$ for $i = 0, 1, \dots$ by the formula

$$h^{i+1} = h^i + \rho_i p^i, \quad (9)$$

where p^i is the antigradient of the cost function $I(h) = J(T(h))$ at the point h^i and ρ_i is an iterative parameter. To find $I'(h)$ we use the method of Lagrange multipliers. Namely, we define Lagrange function

$$L(T, \lambda, h) = (\Psi(T, h), \lambda) + \tilde{J}(T), \quad (10)$$

where

$$\Psi(T, h) := M \frac{H(T) - \tilde{H}}{\tau} + AK(T) + B(h)(T - T_w) + B(T^4 - T_a^4) + DQ$$

and

$$\tilde{J}(T) = \frac{1}{2} \sum_{i=1}^{nz} \tilde{l}_i (T_i - T_i^{tar})^2$$

is the approximation of the cost function J and by (\cdot, \cdot) we mean inner product in R^n . We denote the length of the i -th element by \tilde{l}_i . Differentiation with respect to variable T gives

$$\left(\frac{1}{\tau} M(H'(T) \cdot \partial T) + A(K'(T) \cdot \partial T) + B(h) \cdot \partial T + B(4T^3 \cdot \partial T), \lambda \right) + \tilde{J}'(T) \cdot \partial T = 0$$

and we derive a system of linear algebraic equations

$$\frac{1}{\tau} H' M \lambda + K' A \lambda + B(h) \lambda + 4T^3 B \lambda = -\tilde{J}'. \quad (11)$$

After solving the adjoint state problem (11) the descent direction

$$p_j = \begin{cases} 0, & (h_j = 0 \wedge I'_j > 0) \vee (h_j = h_{max} \wedge I'_j < 0) \\ -I'_j, & \text{otherwise,} \end{cases}$$

$$I'(h) = \left(\frac{\partial \Psi}{\partial n}(T, h), \lambda \right) + \frac{\partial \tilde{J}}{\partial h}(T) = (B(1)(T - T_w), \lambda).$$

is found.

Choose a $h^0 \in U_{ad}$ and natural number n . Given $h^i \in U_{ad}$ we calculate the descent vector p^i and the upper bound for the iterative parameter ρ_i :

$$\rho_{max} = \max\{\rho \geq 0 : h^i + \rho p^i \in U_{ad}\}$$

The optimal step size in the equation (9) is still unknown. We use the following algorithm to determine ρ_{opt} :

Step 1: $\rho_0 = \frac{\rho_{max}}{2^{n+1}}$
 If $J(T(h(\rho_0))) < J(T(h(0)))$ GOTO Step 2;
 Else $\rho_{opt} = 0$ END;
Step 2: For $i = 1, 2, \dots, n$
 $\rho_i = (2^{i+1} - 1)\rho_0$
 $\rho_{opt} = \rho_i$
 If $J(T(h(\rho_i))) < J(T(h(\rho_{i-1})))$ CONTINUE;
 Else $\rho_{opt} = \rho_{i-1}$ END;

Our aim is to use the optimal control method in dynamic casting process. At each iteration the calculation of the solution of the discretized state problem (5) is needed, which is very time consuming. Therefore, we use $n = 7$ in our numerical calculations.

Half width of the slab	$L_x = 0.675m$
Half thickness of the slab	$L_y = 0.105m$
Length of the slab	$L_z = 32m$
Length of the mould	$L_M = 0.9m$
Average cooling in mould	$\hat{Q} = 1229kWm^{-2}$
Latent heat	$L = 261kJkg^{-1}$
Density	$\rho = 7312kgm^{-3}$
Solidus	$T_s = 1763K$
Liquidus	$T_l = 1793K$
Incoming steel temperature	$T_0 = 1808K$
Water temperature	$T_w = 293K$
External temperature	$T_{ext} = 293K$
Emissivity	$\epsilon = 0.8 \cdot 10^{-3}$
Stefan-Boltzmann constant	$\delta = 5.67 \times 10^{-8}Wm^{-2}K^{-4}$
Radius	$R = 0.005m$
Casting speed	$v = 1.4m/min$

TABLE 1. Input parameters for continuous casting problem.

5. NUMERICAL EXAMPLE

Here we present some numerical results for solving continuous casting problem. The secondary cooling region is divided into eight cooling zones. The essential input parameters are presented in the table 1.

In our first numerical test we compare our 3D-model with existing 2D continuous casting model used at Rautaruukki steel factory in Finland. The cooling is assumed to be constant along each cooling zone during the casting process. Thus, the values of heat transfer coefficients h_i are also constant. The values of h_i are presented in the table 2 and the calculated core and surface temperatures are shown in the figure 4.

zone	length [m]	$h[kWm^{-2}K^{-1}]$
1	0.47	0.4950
2	1.46	0.3708
3	1.62	0.3257
4	1.90	0.2783
5	3.84	0.1918
6	5.76	0.1175
7	5.18	0.1233
8	8.20	0.0995

TABLE 2. Zone lengths and values of h in each cooling zones.

We are also interested in calculated lengths of liquid pool (longitudinal length of the liquidus isotherm in the centre of the slab), 'stable' liquid pool ($T_l - 2K$ -isotherm) and mushy pool (T_s -isotherm). Calculated pool lengths for 2D- and 3D-methods are shown in the table 3.

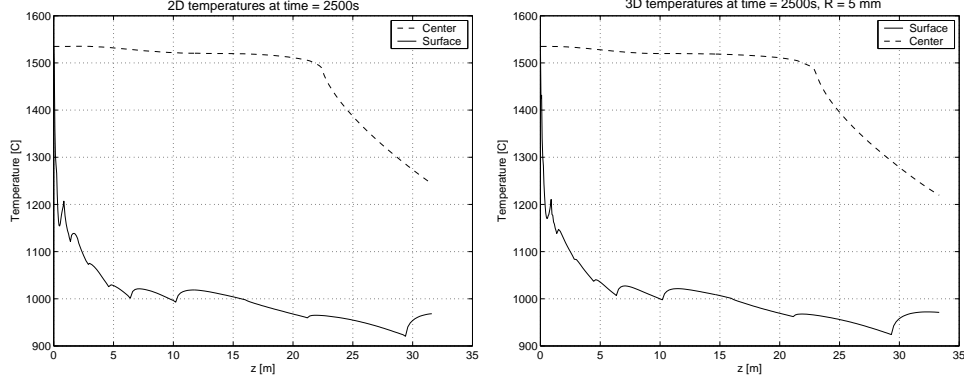


FIGURE 4. Calculated core and surface temperatures for 2D-model (on the left) and 3D-model (on the right) for piecewise constant h .

	2D-method	3D-method
Liquid pool	11.20m	10.55m
Stable liquid pool	16.88m	16.76m
Mushy pool	22.42m	22.63m

TABLE 3. Calculated pool lengths.

Our results show that calculated temperature fields of the two models are very similar but a difference in calculated lengths of liquid pools appear. From the control point of view the information about temperature field and the pool lengths is important because internal cracks of final product arise during solidification process. Moreover, soft reduction technology requires exact information about liquid and mushy pool lengths.

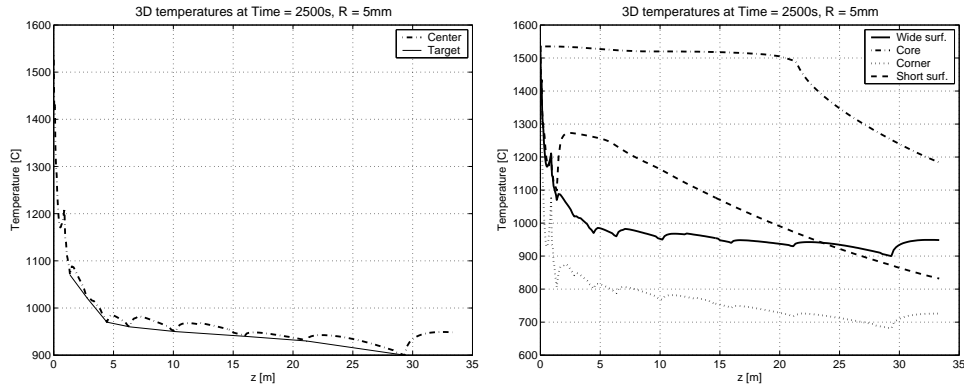


FIGURE 5. Calculated temperatures using 3D-model with PID-control. Left graph shows target and calculated surface temperatures on the centreline of wide face. Right graph shows calculated corner temperatures.

In our second numerical test the PID and optimal control algorithms were also tested to optimize the secondary cooling i.e. heat transfer coefficients.

The results are shown in figures 5 and 6. We can see from the figure 5 that PID control algorithm adjusts water cooling in the way that the difference between target and calculated temperatures at the end of each cooling zone is minimized. The optimal control method on the other hand minimizes the cost function J . From the figure 6 we see that the difference between target and calculated surface temperature is minimized on the whole length of the slab.

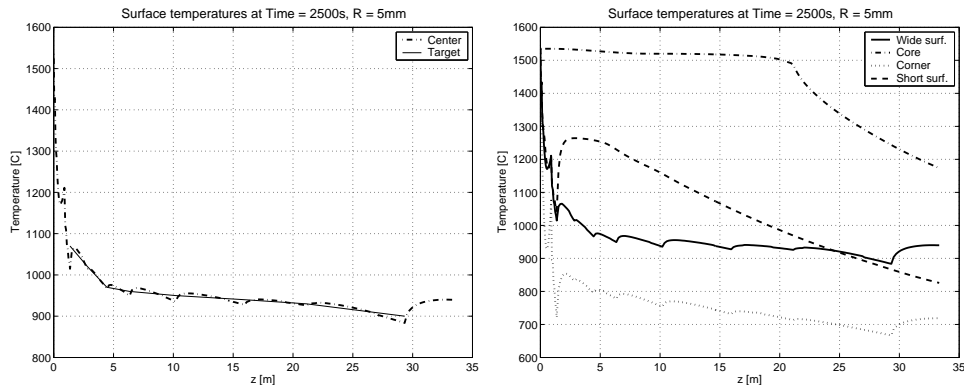


FIGURE 6. Calculated temperatures using 3D-model with optimal control. Left graph shows target and calculated surface temperatures on the centreline of wide face. Right graph shows calculated corner temperatures.

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