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A non-overlapping optimized Schwarz method for the heat equation with non linear boundary conditions and with applications to de-icing

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Abstract

When simulating complex physical phenomena such as aircraft icing or de-icing, several dedicated solvers often need to be strongly coupled. In this work, a non-overlapping Schwarz method is constructed with the unsteady simulation of de-icing as the targetted application. To do so, optimized coupling coefficients are first derived for the one dimensional unsteady heat equation with linear boundary conditions and for the steady heat equation with non-linear boundary conditions. The choice of these coefficients is shown to guarantee the convergence of the method. Using a linearization of the boundary conditions, the method is then extended to the case of a general unsteady heat conduction problem. The method is tested on simple cases and the convergence properties are assessed theoretically and numerically. Finally the method is applied to the simulation of an aircraft electrothermal de-icing problem in two dimensions.

Keywords: Numerical simulation, Schwarz method, Coupling, Aircraft icing, Electrothermal de-icing

1 Introduction

The simulation of complex multiphysical phenomena often requires the coupling of several dedicated solvers. As an example, the phenomenon of icing on aircrafts involves many physical aspects such as aerodynamics, droplet trajectories and heat and mass transfer [23]. A typical simulation thus requires the use of several solvers which pass information to each other. The unsteady simulation of thermal de-icing systems requires the coupling between an unsteady thermal model of the de-icing system and an unsteady thermal model of the ice accretion and melting process. As the physical properties of ice and the airfoil are different, the union of these domains is heterogenous. The heat equation is a central element of these models. It is therefore of great interest to conduct an analysis of coupling methods between two domains where the heat equation is to be solved in unsteady (parabolic) or steady (elliptic) form. Also, icing being a complex phenomenon involving for example evaporation or sublimation, the ability to deal with generic boundary conditions is also required. The general context of this work is therefore the coupled solution of the steady or unsteady heat equation on bounded non-overlapping domains with general non-linear boundary conditions.

The coupling of models and codes is an active area of research. Domain decomposition methods form a large family of coupling techniques[24, 22, 10]. Schwarz pioneered this area of research when, in 1870, he showed the existence of the solution of the heat equation on a domain consisting of the overlapping of a rectangle and a disk [21]. The idea is to first solve the heat equation in the disk. Then the resulting solution is used to impose a Dirichlet boundary condition on the rectangle. In turn, the heat equation on the rectangle is solved and provides Dirichlet boundary conditions for the disk. The process is repeated thus generating a sequence of solutions to the heat equation in each domain. Schwarz proved that the sequence

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21 converges which shows the existence of a solution of the heat equation on the union of the disk and the
 22 rectangle. This idea has since then be improved and revisited over time [13].

23 The Schwarz method has been applied to many problems, with and without overlapping [16, 17, 9, 1].
 24 The issue of obtaining optimal coefficients has also been addressed in many different contexts [15, 12, 5].
 25 The general approach is to study the problem on an unbounded domain and use Fourier analysis to derive
 26 the conditions and coefficients that guarantee optimal convergence.

27 For elliptic problems on bounded non-overlapping domains, Lions proposed to use Robin transmission
 28 conditions instead of Dirichlet transmission conditions and proved general convergence results[18]. On the
 29 outer boundaries Lions considered homogeneous Dirichlet boundary conditions. He noted that the effective
 30 choice of the coupling coefficient is not trivial and provided some example for simple geometries. Errera *et al.*
 31 derived schemes for the numerical solution of two coupled heat conduction problems[11]. In particular, they
 32 studied non-overlapping Robin-Robin transmission conditions and analysed the stability and convergence
 33 properties of their approach. Later, Gander *et al.* studied an optimized Schwarz method for the diffusion
 34 problem on non-overlapping heterogeneous media[14]. After deriving optimal coefficients using Fourier
 35 analysis, they performed numerical tests to demonstrate the efficiency of the method. More recently, Meng
 36 *et al.* developed the Conjugate Heat transfer Advanced Multi-domain Partitioned scheme[20]. To do so
 37 they use elements of optimal overlapping Schwarz methods with Robin transmission conditions combined
 38 with interface jump conditions. The method is demonstrated to be stable for a large range of heat transfer
 39 problems. Concerning the external boundary, the previous studies all used linear Dirichlet, Neumann or
 40 mixed boundary conditions.

41 In typical of icing simulations, the evaporative term induces non-linear boundary conditions. Moreover,
 42 when simulating the formation or melting of ice, the shape of the different layers, and hence the mesh,
 43 changes with time. Moreover, in an industrial context, it is highly desirable that the code simulating the
 44 ice protection system and the one simulating the ice accretion and melting be separate. Therefore the goal
 45 of this work is to construct a non-overlapping Robin-Robin Schwarz method suited for icing simulation,
 46 building upon previous research and the authors' previous work [8].

47 The article is organized as follows. The idea is to first study and derive coupling coefficients for a one
 48 dimensional unsteady heat conduction problem with linear boundary conditions. The next step is to consider
 49 a steady state problem with non-linear boundary conditions. The convergence of the method is proved in
 50 both of these cases. The two previous steps are then used to propose an extension for the unsteady case with
 51 non-linear boundary conditions. Numerical tests are performed to assess the performance of the method.
 52 Finally an application to the simulation of electro-thermal de-icing is performed to illustrate the method in
 53 more complex situations.

54 The general setting for the derivation of the method is given by problem (1). It consists of an unsteady
 55 coupled heat conduction problem in two domains of different physical properties (see Figure 1) and generic
 56 boundary conditions.

$$\left\{ \begin{array}{ll} \rho_1(x)c_1(x)\frac{\partial T_1}{\partial t}(t,x) = \frac{\partial}{\partial x} \left(\lambda_1(x)\frac{\partial T_1}{\partial x}(t,x) \right) & \forall (t,x) \in \mathbb{R}^+ \times \Omega_1 \quad (1a) \\ \rho_2(x)c_2(x)\frac{\partial T_2}{\partial t}(t,x) = \frac{\partial}{\partial x} \left(\lambda_2(x)\frac{\partial T_2}{\partial x}(t,x) \right) & \forall (t,x) \in \mathbb{R}^+ \times \Omega_2 \quad (1b) \\ \lambda_1(-l_1)\frac{\partial T_1}{\partial x}(t,-l_1) = f_1(T_1(t,-l_1)) & \forall t \in \mathbb{R}^+ \quad (1c) \\ \lambda_2(l_2)\frac{\partial T_2}{\partial x}(t,l_2) = -f_2(T_2(t,l_2)) & \forall t \in \mathbb{R}^+ \quad (1d) \\ T_1(t,0) = T_2(t,0) & \forall t \in \mathbb{R}^+ \quad (1e) \\ \lambda_1(0)\frac{\partial T_1}{\partial x}(t,0) = \lambda_2(0)\frac{\partial T_2}{\partial x}(t,0) & \forall t \in \mathbb{R}^+ \quad (1f) \end{array} \right.$$

57 where t , x and T stand respectively for time, position and temperature. For each domain $i = 1, 2$ ρ_i
 58 is the density, c_i the specific heat, λ_i the thermal conductivity, and f_1 and f_2 are increasing and regular
 59 enough functions.

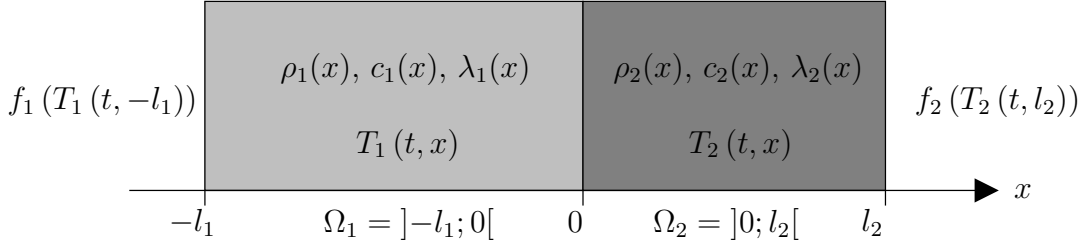


Figure 1: Illustration of the two domains Ω_1 and Ω_2

The goal is to solve each heat conduction problem in its respective domain Ω_i . The domains are coupled through the relations (1e) (continuity of temperature) and (1f) (continuity of heat flux). In the following, this problem is studied by considering two specific problems for which an optimized coupling procedure is derived:

- Unsteady heat conduction with linear boundary conditions.
- Steady heat conduction with general boundary conditions.

These analyses are then used to provide a coupling procedure for the general problem of unsteady heat conduction with general boundary conditions.

1. Unsteady case with linear boundary conditions

In this case, the problem (1) is considered with constant physical properties over each domain and with linear external boundary conditions. More precisely, the functions f_i are assumed to be linear functions of T_i . In order to proceed the time derivative is discretized using an implicit Euler scheme. Discretization is performed using a time step Δt . At the n -th step (where $n \in (N)$) the time is given by $t^n = n\Delta t$. The set of equations hence becomes:

Semi-discrete heat equation in each domain:

$$\rho_1 c_1 \frac{T_1^{n+1} - T_1^n}{\Delta t} = \lambda_1 \frac{d^2 T_1^{n+1}}{dx^2} \quad \forall x \in \Omega_1 \quad (2a)$$

$$\rho_2 c_2 \frac{T_2^{n+1} - T_2^n}{\Delta t} = \lambda_2 \frac{d^2 T_2^{n+1}}{dx^2} \quad \forall x \in \Omega_2 \quad (2b)$$

where T_j^n is the temperature field at time step n . Note that the unknown of each equation is T_j^{n+1} .

External boundary conditions:

$$-\lambda_1 \frac{dT_1^{n+1}}{dx}(-l_1) = h_1 (T_{r,1} - T_1^{n+1}(-l_1)) \quad (2c)$$

$$\lambda_2 \frac{dT_2^{n+1}}{dx}(l_2) = h_2 (T_{r,2} - T_2^{n+1}(l_2)) \quad (2d)$$

where the constants h_i are heat transfer coefficients and the $T_{r,i}$ represent cooling/heating temperatures.

Interfacial conditions:

$$T_1^{n+1}(0) = T_2^{n+1}(0) \quad (2e)$$

$$\lambda_1 \frac{dT_1^{n+1}}{dx}(0) = \lambda_2 \frac{dT_2^{n+1}}{dx}(0) \quad (2f)$$

80 where $n_{e,j}$ is the unit normal vector of the external boundary.

81 By linearity it suffices to consider only the associated homogeneous problems in each domain. To avoid
82 cumbersome notations, T_j^{n+1} is now written simply T_j . The homogeneous problems are defined by:

Semi-discrete homogeneous heat equation in each domain:

$$-\frac{d^2 T_1}{dx^2} + \mu_1^2 T_1 = 0 \quad \forall x \in \Omega_1 \quad (3a)$$

$$-\frac{d^2 T_2}{dx^2} + \mu_2^2 T_2 = 0 \quad \forall x \in \Omega_2 \quad (3b)$$

External boundary conditions:

$$-\lambda_1 \frac{dT_1}{dx}(-l_1) = -h_1 T_1(-l_1) \quad (3c)$$

$$\lambda_2 \frac{dT_2}{dx}(l_2) = -h_2 T_2(l_2) \quad (3d)$$

Interfacial conditions:

$$T_1(0) = T_2(0) \quad (3e)$$

$$\lambda_1 \frac{dT_1}{dx}(0) = \lambda_2 \frac{dT_2}{dx}(0) \quad (3f)$$

83 where $\mu_j^2 = \frac{\rho_j c_j}{\Delta t \lambda_j}$ ($j = 1, 2$).

84 This coupled problem can be solved using a Schwarz algorithm. It defines the sequences $(T_1^{(k)})_{k \in \mathbb{N}}$ and
85 $(T_2^{(k)})_{k \in \mathbb{N}}$ as solutions (in respectively $H^1(\Omega_1)$ and $H^1(\Omega_2)$) to the following problems (see for example [18]):

$$\left\{ \begin{array}{l} -\frac{d^2 T_1^{(k+1)}}{dx^2} + \mu_1^2 T_1^{(k+1)} = 0 \quad \forall x \in \Omega_1 \quad (4a) \\ -\lambda_1 \frac{dT_1^{(k+1)}}{dx}(-l_1) = -h_1 T_1^{(k+1)}(-l_1) \quad (4b) \\ \lambda_1 \frac{dT_1^{(k+1)}}{dx}(0) = \lambda_2 \frac{dT_2^{(k)}}{dx}(0) + \omega_1 \left(T_2^{(k)}(0) - T_1^{(k+1)}(0) \right) \quad (4c) \end{array} \right.$$

$$\left\{ \begin{array}{l} -\frac{d^2 T_2^{(k+1)}}{dx^2} + \mu_2^2 T_2^{(k+1)} = 0 \quad \forall x \in \Omega_2 \quad (5a) \\ \lambda_2 \frac{dT_2^{(k+1)}}{dx}(l_2) = -h_2 T_2^{(k+1)}(l_2) \quad (5b) \\ -\lambda_2 \frac{dT_2^{(k+1)}}{dx}(0) = -\lambda_1 \frac{dT_1^{(k)}}{dx}(0) + \omega_2 \left(T_1^{(k+1)}(0) - T_2^{(k+1)}(0) \right) \quad (5c) \end{array} \right.$$

86 where ω_1 and ω_2 are two strictly positive real numbers (the coupling coefficients). The algorithm is initialized
87 with an arbitrary guess of the temperature fields given by T_1^0 and T_2^0 . In practice, the temperature field
88 at time t^n can be used as an initializing guess of the solution. One now has to determine whether this
89 algorithm converges or not.

However, by applying the Lax-Milgram theorem [6], it can be shown that in $H^1(] - l_1, l_2[)$ problem (3) has a unique solution given by:

$$T_1(x) = 0 \quad \forall x \in \Omega_1 \quad (6a)$$

$$T_2(x) = 0 \quad \forall x \in \Omega_2 \quad (6b)$$

90 Given the solution (6), the goal is now to determine for which conditions the sequences $(T_1^{(k)})_{k \in \mathbb{N}}$ and
 91 $(T_2^{(k)})_{k \in \mathbb{N}}$ converge to 0. Before doing so, let us first introduce some notations. For $i \in 1, 2$, let

$$\chi_i = \exp(-2\mu_i l_i) \frac{\frac{\lambda_i \mu_i}{h_i} - 1}{\frac{\lambda_i \mu_i}{h_i} + 1} \quad (7)$$

$$\xi_i = \frac{\lambda_i \mu_i (1 - \chi_i)}{1 + \chi_i} \quad (8)$$

93 Note that as $\lambda_i, \mu_i, h_i, l_i > 0$:

$$-1 < \chi_i < 1 \quad i = 1, 2 \quad (9)$$

$$\xi_i > 0 \quad i = 1, 2 \quad (10)$$

95 The following proposition now provides a criterion to chose the coupling coefficients ω_1 and ω_2 .

96 **Proposition 1.** *If:*

$$\left| \frac{(\omega_1 - \xi_2)(\omega_2 - \xi_1)}{(\omega_1 + \xi_1)(\omega_2 + \xi_2)} \right| < 1 \quad (11)$$

97 and $T_1^{(0)}$ and $T_2^{(0)}$ are arbitrary initial guesses for the temperature field then:

$$\begin{cases} \lim_{k \rightarrow \infty} T_1^{(k)} = 0 \\ \lim_{k \rightarrow \infty} T_2^{(k)} = 0 \end{cases} \quad (12)$$

98 *Proof.* The solutions $T_1^{(k)}$ and $T_2^{(k)}$, for $k \geq 1$, to the constant coefficient homogeneous ODEs (4a) and (5a)
 99 are:

$$T_1^{(k)} = A_1^{(k)} e^{\mu_1 x} + B_1^{(k)} e^{-\mu_1 x} \quad (13)$$

$$T_2^{(k)} = A_2^{(k)} e^{\mu_2 x} + B_2^{(k)} e^{-\mu_2 x} \quad (14)$$

100 Inserting (13) into boundary condition (4b) yields:

$$B_1^{(k)} = \chi_1 A_1^{(k)} \quad (15)$$

101 Moreover, inserting (14) into boundary condition (5b) yields:

$$A_2^{(k)} = \chi_2 B_2^{(k)} \quad (16)$$

First, if $k = 1$ then the coupling boundary conditions (4c) and (5c) give the relations:

$$\begin{aligned} \lambda_1 \mu_1 [A_1^{(1)} - B_1^{(1)}] &= \lambda_2 \frac{dT_2^{(0)}}{dx}(0) \\ &+ \omega_1 T_2^{(0)} - \omega_1 [A_1^{(1)} + B_1^{(1)}] \end{aligned} \quad (17)$$

$$\begin{aligned} -\lambda_2 \mu_2 [A_2^{(1)} - B_2^{(1)}] &= -\lambda_1 \frac{dT_1^{(0)}}{dx}(0) \\ &+ \omega_2 ([A_1^{(1)} + B_1^{(1)}] - [A_2^{(1)} + B_2^{(1)}]) \end{aligned} \quad (18)$$

102 In combination with (15) and (16), relations (17) and (18) completely determine $A_1^{(1)}$, $B_1^{(1)}$, $A_2^{(1)}$ and
 103 $B_2^{(1)}$.

On the other hand for $k \geq 1$, combining (13) with the Fourier-Robin coupling boundary condition (4c), one obtains:

$$\begin{aligned} \lambda_1 \mu_1 \left[A_1^{(k+1)} - B_1^{(k+1)} \right] &= \lambda_2 \mu_2 \left[A_2^{(k)} - B_2^{(k)} \right] \\ &+ \omega_1 \left(\left[A_2^{(k)} + B_2^{(k)} \right] - \left[A_1^{(k+1)} + B_1^{(k+1)} \right] \right) \end{aligned} \quad (19)$$

104 Using relations (15) and (16) the previous equation yields:

$$A_1^{(k+1)} = \frac{\omega_1 - \xi_2}{\omega_1 \frac{1+\chi_1}{1+\chi_2} + \lambda_1 \mu_1 \frac{1-\chi_1}{1+\chi_2}} B_2^{(k)} \quad (20)$$

105 Proceeding the same way with solution (14), the Fourier-Robin coupling boundary condition (5c) and
106 relations (15) and (16) one also obtains:

$$B_2^{(k)} = \frac{\omega_2 - \xi_1}{\omega_2 \frac{1+\chi_2}{1+\chi_1} + \lambda_2 \mu_2 \frac{1-\chi_2}{1+\chi_1}} A_1^{(k)} \quad (21)$$

107 Combining the two previous relations (20) and (21) yields:

$$A_1^{(k+1)} = \frac{(\omega_1 - \xi_2)(\omega_2 - \xi_1)}{(\omega_1 + \xi_1)(\omega_2 + \xi_2)} A_1^{(k)} \quad (22)$$

108 Hypothesis (11) implies that:

$$\left| \frac{A_1^{(k+1)}}{A_1^{(k)}} \right| = \left| \frac{B_1^{(k+1)}}{B_1^{(k)}} \right| < 1 \quad (23)$$

109 Therefore the sequences $\left(A_1^{(k)} \right)_{k \in \mathbb{N}}$ and $\left(B_1^{(k)} \right)_{k \in \mathbb{N}}$ tend to zero as k tends to infinity. Hence:

$$\lim_{k \rightarrow \infty} T_1^{(k)} = 0 \quad (24)$$

110 Moreover, given that for all $k \in \mathbb{N}$, $A_2^{(k)}$ and $B_2^{(k)}$ are proportional to $A_1^{(k)}$ and $B_1^{(k)}$, the sequences
111 $\left(A_2^{(k)} \right)_{k \in \mathbb{N}}$ and $\left(B_2^{(k)} \right)_{k \in \mathbb{N}}$ also tend to zero as k tends to infinity. Therefore:

$$\lim_{k \rightarrow \infty} T_2^{(k)} = 0 \quad (25)$$

112 □

An immediate corollary of the previous proof is that if one choses $\omega_1 = \xi_2$ and/or $\omega_2 = \xi_1$ then the algorithm converges in one iteration. This hence provides optimal coupling coefficients for the linear unsteady case:

$$\omega_1^\dagger = \xi_2 \quad (26a)$$

$$\omega_2^\dagger = \xi_1 \quad (26b)$$

113 2. The steady non-linear problem

114 The steady version of Eq. (1) is now studied. Contrary to the previous section, no simplifying assump-
115 tions are made about the boundary conditions and about the distribution of the thermal conductivity λ . T_1
116 and T_2 are supposed to be functions of the single variable x .

117 In each domain Ω_i ($i \in \{1, 2\}$), Eq. (1a) can be rewritten:

$$\frac{\partial}{\partial x} \left(\lambda_i(x) \frac{\partial T_i}{\partial x}(x) \right) = 0 \quad \forall x \in \Omega_i \quad (27)$$

For the sake of clarity, the non-linear boundary conditions computed at the outer boundary conditions (Eqs. (1c) and (1d)) are reminded:

$$\begin{cases} \lambda_1(-l_1) \frac{\partial T_1}{\partial x}(-l_1) = f_1(T_1(-l_1)) & (28a) \\ \lambda_2(l_2) \frac{\partial T_2}{\partial x}(l_2) = -f_2(T_2(l_2)) & (28b) \end{cases}$$

The same applies to the boundary conditions between the domains Ω_1 and Ω_2 (Eqs. (1e) and (1f)):

$$\begin{cases} T_1(0) = T_2(0) & (29a) \\ \lambda_1(0) \frac{\partial T_1}{\partial x}(0) = \lambda_2(0) \frac{\partial T_2}{\partial x}(0) & (29b) \end{cases}$$

118 To begin with, a useful result for the remainder of the paper is established:

Proposition 2.

$$\begin{cases} f_1(T_1(-l_1)) = \lambda_1(-l_1) \frac{\partial T_1}{\partial x}(-l_1) = \lambda_1(0) \frac{\partial T_1}{\partial x}(0) & (30a) \\ -f_2(T_2(l_2)) = \lambda_2(l_2) \frac{\partial T_2}{\partial x}(l_2) = \lambda_2(0) \frac{\partial T_2}{\partial x}(0) & (30b) \end{cases}$$

119
120 *Proof.* The first equality in Eqs. (30a) and (30b) is derived from Eq. (28). Regarding the second equality,
121 $\lambda_i(x) \frac{\partial T_i}{\partial x}(x)$ is constant $\forall x \in \Omega_i$ (Eq. (27)). Therefore, $\lambda_1(-l_1) \frac{\partial T_1}{\partial x}(-l_1) = \lambda_1(0) \frac{\partial T_1}{\partial x}(0)$ and $\lambda_2(l_2) \frac{\partial T_2}{\partial x}(l_2) =$
122 $\lambda_2(0) \frac{\partial T_2}{\partial x}(0)$ which justifies the second equality. \square

123 Using Eq. (29b), a direct consequence of Eqs. (30) from Prop. (2) is:

$$f_1(T_1(-l_1)) = -f_2(T_2(l_2)) \quad (31)$$

124 The following proposition is a key step in the demonstration process:

125 **Proposition 3.** *The two following systems of equations i) and ii) are equivalent:*

126 i) *Initial system composed of Eqs. (27) + (28) + (29)*

ii) *Eqs. (27) + (28) with the following additional condition (Eq. (32)):*

$T_1(-l_1)$ and $T_2(l_2)$ are the solutions of the system made of the following non linear equations:

$$\begin{cases} T_1(-l_1) + (l_1 r_1 + l_2 r_2) f_1(T_1(-l_1)) = T_2(l_2) & (32a) \\ T_2(l_2) + (l_1 r_1 + l_2 r_2) f_2(T_2(l_2)) = T_1(-l_1) & (32b) \end{cases}$$

127

128 where the thermal resistances r_i ($i \in \{1, 2\}$) for the domain Ω_i are defined by:

$$r_i \triangleq \frac{1}{l_i} \int_{\Omega_i} \frac{dx}{\lambda_i(x)} \quad (33)$$

Proof. The solution of the problem (27) and (28) is obviously given by:

$$\begin{cases} T_1(x) = T_1(-l_1) + f_1(T_1(-l_1)) \int_{\widehat{\Omega}_1(x)} \frac{dx'}{\lambda_1(x')} & \forall x \in \widehat{\Omega}_1 \\ T_2(x) = T_2(l_2) + f_2(T_2(l_2)) \int_{\widehat{\Omega}_2(x)} \frac{dx'}{\lambda_2(x')} & \forall x \in \widehat{\Omega}_2 \end{cases} \quad (34a)$$

$$\quad (34b)$$

where $\widehat{\Omega}_1(x) =]-l_1; x[$ and $\widehat{\Omega}_2(x) =]x; l_2[$. Note that $\widehat{\Omega}_1(0) = \Omega_1$ and $\widehat{\Omega}_2(0) = \Omega_2$. Therefore, from Eq. (34), the only unknowns to be computed to determine the fields T_1 and T_2 in the whole computational domain Ω_1 and Ω_2 respectively are $T_1(-l_1)$ and $T_2(l_2)$. If Eq. (34) is written for $x = 0$, it provides:

$$\begin{cases} T_1(0) = T_1(-l_1) + l_1 r_1 f_1(T_1(-l_1)) \\ T_2(0) = T_2(l_2) + l_2 r_2 f_2(T_2(l_2)) \end{cases} \quad (35a)$$

$$\quad (35b)$$

i) \Rightarrow ii) Since $T_1(0) = T_2(0)$ (Eq. (29a) and $-f_1(T_1(-l_1)) = f_2(T_2(l_2))$ (Eq. (31), Eq. (32) is derived from Eq. (35).

129 ii) \Rightarrow i) Using Eq. (31), Eq. (35) may be rewritten:

$$T_1(0) - T_2(0) = T_1(-l_1) - T_2(l_2) + (l_1 r_1 + l_2 r_2) f_1(T_1(-l_1)) \quad (36)$$

130 Therefore, from Eq. (32), it provides $T_1(0) = T_2(0)$. A direct consequence of Eqs. (30) and (31) is
131 $\lambda_1(0) \frac{\partial T_1}{\partial x}(0) = \lambda_2(0) \frac{\partial T_2}{\partial x}(0)$. \square

The following Schwarz algorithm is proposed between the domains Ω_1 and Ω_2 to find a solution of the system made of Eqs. (27), (28) and (29). Let $T_1^{(k)}$ and $T_2^{(k)}$ two sequences of functions defined iteratively by Eq. (37) and (38).

$$\begin{cases} \frac{\partial}{\partial x} \left(\lambda_1(x) \frac{\partial T_1^{(k+1)}}{\partial x}(x) \right) = 0 & \forall x \in \Omega_1 \\ \lambda_1(-l_1) \frac{\partial T_1^{(k+1)}}{\partial x}(-l_1) = f_1(T_1^{(k+1)}(-l_1)) \\ \lambda_1(0) \frac{\partial T_1^{(k+1)}}{\partial x}(0) = \lambda_2(0) \frac{\partial T_2^{(k)}}{\partial x}(0) + \omega_1 (T_2^{(k)}(0) - T_1^{(k+1)}(0)) \end{cases} \quad (37a)$$

$$\quad (37b)$$

$$\quad (37c)$$

132

$$\begin{cases} \frac{\partial}{\partial x} \left(\lambda_2(x) \frac{\partial T_2^{(k+1)}}{\partial x}(x) \right) = 0 & \forall x \in \Omega_2 \\ \lambda_2(l_2) \frac{\partial T_2^{(k+1)}}{\partial x}(l_2) = -f_2(T_2^{(k+1)}(l_2)) \\ \lambda_2(0) \frac{\partial T_2^{(k+1)}}{\partial x}(0) = \lambda_1(0) \frac{\partial T_1^{(k+1)}}{\partial x}(0) - \omega_2 (T_1^{(k+1)}(0) - T_2^{(k+1)}(0)) \end{cases} \quad (38a)$$

$$\quad (38b)$$

$$\quad (38c)$$

133 with ω_1 and ω_2 in \mathbb{R}^+ .

Proposition 4. *Given the following hypothesis:*

$$\begin{cases} \omega_1 = 1/(r_2 l_2) \\ \omega_2 = 1/(r_1 l_1) \end{cases} \quad (39a)$$

$$\quad (39b)$$

$$\begin{cases} \exists \kappa_1 \in \mathbb{R}^{+*}, \quad \forall x \in \mathbb{R}, \quad f_1'(x) \geq \kappa_1 \\ \exists \kappa_2 \in \mathbb{R}^{+*}, \quad \forall x \in \mathbb{R}, \quad f_2'(x) \geq \kappa_2 \end{cases} \quad (39c)$$

$$\quad (39d)$$

134 the sequences $T_1^{(k)}$ and $T_2^{(k)}$ are convergent for every initial solution. More precisely: for every initial
 135 solution $T_1^{(0)}$ and $T_2^{(0)}$ which satisfy Eqs. (27) and (28), we have $T_1^{(k)} \xrightarrow[k \rightarrow \infty]{} T_1$ and $T_2^{(k)} \xrightarrow[k \rightarrow \infty]{} T_2$ where T_1
 136 and T_2 are the solutions of Eqs. (27), (28) and (29).

Proof. From Prop. (2) the previous algorithm can be rewritten:

$$\left\{ \begin{array}{l} \frac{\partial}{\partial x} \left(\lambda_1(x) \frac{\partial T_1^{(k+1)}}{\partial x}(x) \right) = 0 \quad \forall x \in \Omega_1 \end{array} \right. \quad (40a)$$

$$\left\{ \begin{array}{l} \lambda_1(-l_1) \frac{\partial T_1^{(k+1)}}{\partial x}(-l_1) = f_1(T_1^{(k+1)}(-l_1)) \end{array} \right. \quad (40b)$$

$$\left\{ \begin{array}{l} f_1(T_1^{(k+1)}(-l_1)) = -f_2(T_2^{(k)}(l_2)) + \omega_1(T_2^{(k)}(0) - T_1^{(k+1)}(0)) \end{array} \right. \quad (40c)$$

137

$$\left\{ \begin{array}{l} \frac{\partial}{\partial x} \left(\lambda_2(x) \frac{\partial T_2^{(k+1)}}{\partial x}(x) \right) = 0 \quad \forall x \in \Omega_2 \end{array} \right. \quad (41a)$$

$$\left\{ \begin{array}{l} \lambda_2(l_2) \frac{\partial T_2^{(k+1)}}{\partial x}(l_2) = -f_2(T_2^{(k+1)}(l_2)) \end{array} \right. \quad (41b)$$

$$\left\{ \begin{array}{l} -f_2(T_2^{(k+1)}(l_2)) = f_1(T_1^{(k+1)}(-l_1)) - \omega_2(T_1^{(k+1)}(0) - T_2^{(k+1)}(0)) \end{array} \right. \quad (41c)$$

138 We focus now on the sequences $T_1^{(k)}(-l_1)$ and $T_2^{(k)}(l_2)$. Using Eq. (35), the equations (40c) and (41c)
 139 may be written:

$$\left\{ \begin{array}{l} f_1(T_1^{(k+1)}(-l_1)) = -f_2(T_2^{(k)}(l_2)) \\ \quad + \omega_1(T_2^{(k)}(l_2) + l_2 r_2 f_2(T_2^{(k)}(l_2)) - T_1^{(k+1)}(-l_1) - l_1 r_1 f_1(T_1^{(k+1)}(-l_1))) \\ -f_2(T_2^{(k+1)}(l_2)) = f_1(T_1^{(k+1)}(-l_1)) \\ \quad - \omega_2(T_1^{(k+1)}(-l_1) + l_1 r_1 f_1(T_1^{(k+1)}(-l_1)) - T_2^{(k+1)}(l_2) - l_2 r_2 f_2(T_2^{(k+1)}(l_2))) \end{array} \right. \quad (42)$$

which may be rewritten using hypothesis (39a) and (39b):

$$\left\{ \begin{array}{l} G_1(T_1^{(k+1)}(-l_1)) = T_2^{(k)}(l_2) + (l_2 r_2 - 1/\omega_1) f_2(T_2^{(k)}(l_2)) \end{array} \right. \quad (43a)$$

$$\left\{ \begin{array}{l} G_2(T_2^{(k+1)}(l_2)) = T_1^{(k+1)}(-l_1) + (l_1 r_1 - 1/\omega_2) f_1(T_1^{(k+1)}(-l_1)) \end{array} \right. \quad (43b)$$

where G_1 and G_2 are defined as:

$$\left\{ \begin{array}{l} G_1(x) \triangleq x + \left(l_1 r_1 + \frac{1}{\omega_1} \right) f_1(x) \end{array} \right. \quad (44a)$$

$$\left\{ \begin{array}{l} G_2(x) \triangleq x + \left(l_2 r_2 + \frac{1}{\omega_2} \right) f_2(x) \end{array} \right. \quad (44b)$$

From hypothesis (39c) and (39d), it can be written:

$$\left\{ \begin{array}{l} G_1'(x) > 1 + \left(l_1 r_1 + \frac{1}{\omega_1} \right) \kappa_1 \quad \forall x \in \mathbb{R} \end{array} \right. \quad (45a)$$

$$\left\{ \begin{array}{l} G_2'(x) > 1 + \left(l_2 r_2 + \frac{1}{\omega_2} \right) \kappa_2 \quad \forall x \in \mathbb{R} \end{array} \right. \quad (45b)$$

which means that G_1 and G_2 are invertible functions. Equation (45) allows to limit the derivatives of the inverse functions $g_1 = G_1^{-1}$ and $g_2 = G_2^{-1}$:

$$\begin{cases} 0 < g_1'(x) \leq \frac{1}{1 + (l_1 r_1 + 1/\omega_1) \kappa_1} < 1 & \forall x \in \mathbb{R} \\ 0 < g_2'(x) \leq \frac{1}{1 + (l_2 r_2 + 1/\omega_2) \kappa_2} < 1 & \forall x \in \mathbb{R} \end{cases} \quad (46a)$$

$$\begin{cases} 0 < g_1'(x) \leq \frac{1}{1 + (l_1 r_1 + 1/\omega_1) \kappa_1} < 1 & \forall x \in \mathbb{R} \\ 0 < g_2'(x) \leq \frac{1}{1 + (l_2 r_2 + 1/\omega_2) \kappa_2} < 1 & \forall x \in \mathbb{R} \end{cases} \quad (46b)$$

Equation (43) may be rewritten:

$$\begin{cases} T_1^{(k+1)}(-l_1) = g_1 \circ p_1(T_2^{(k)}(l_2)) \\ T_2^{(k+1)}(l_2) = g_2 \circ p_2(T_1^{(k+1)}(-l_1)) \end{cases} \quad (47a)$$

$$\begin{cases} T_1^{(k+1)}(-l_1) = g_1 \circ p_1(T_2^{(k)}(l_2)) \\ T_2^{(k+1)}(l_2) = g_2 \circ p_2(T_1^{(k+1)}(-l_1)) \end{cases} \quad (47b)$$

where the functions p_1 and p_2 are given by:

$$\begin{cases} p_1(x) \triangleq x + (l_2 r_2 - 1/\omega_1) f_2(x) & \forall x \\ p_2(x) \triangleq x + (l_1 r_1 - 1/\omega_2) f_1(x) & \forall x \end{cases} \quad (48a)$$

$$\begin{cases} p_1(x) \triangleq x + (l_2 r_2 - 1/\omega_1) f_2(x) & \forall x \\ p_2(x) \triangleq x + (l_1 r_1 - 1/\omega_2) f_1(x) & \forall x \end{cases} \quad (48b)$$

Therefore Eq. (47) can be written:

$$T_2^{(k+1)}(l_2) = g_2 \circ p_2 \circ g_1 \circ p_1(T_2^{(k)}(l_2)) \quad (49)$$

From hypothesis (39a) and (39b), the functions p_1 and p_2 are the function identity and $T_2^{(k+1)}(l_2) = g_2 \circ g_1(T_2^{(k)}(l_2))$. Let $A \in \mathbb{R}^{+*}$ and $(x, y) \in [-A; A]$. According to the mean value equality on $[-A; A]$:

$$|g_2 \circ g_1(x) - g_2 \circ g_1(y)| \leq \sup_{z \in [-A; A]} |(g_2 \circ g_1)'(z)| \cdot |x - y| \quad (50)$$

where $\sup_{z \in [-A; A]} |(g_2 \circ g_1)'(z)| < \sup_{z \in \mathbb{R}^+} |(g_2 \circ g_1)'(z)| < 1$ from Eq. (46). Therefore, the function $g_2 \circ g_1$ is a contractant function, has a single fixed point $T_1(-l_1)$ and $T_1^{(k)}(-l_1) \xrightarrow[k \rightarrow +\infty]{} T_1(-l_1)$. By analogous reasoning, it can be shown that $g_1 \circ g_2$ is a contractant function, has a single fixed point $T_2(l_2)$ and $T_2^{(k)}(l_2) \xrightarrow[k \rightarrow +\infty]{} T_2(l_2)$. After a passage to the limit in Eq. (43), $T_1(-l_1)$ and $T_2(l_2)$ are the solution of Eqs. (32). Therefore, from Prop. (3), the limit functions T_1 and T_2 of the sequences $T_1^{(k)}$ and $T_2^{(k)}$ are solutions of the problem (27) + (28) + (29). \square

An optimized version of the presented Schwarz algorithm is now presented. To do this, another definition for the coefficients ω_1 and ω_2 is proposed so as to define a “more contractant” function $g_2 \circ p_2 \circ g_1 \circ p_1$ (Eq. (49)). One way to proceed is to cancel the derivative $p_1'(T_2^{(k)}(l_2))$:

$$p_1'(T_2^{(k)}(l_2)) = 0 \iff (\omega_1^{(k+1)})^{\dagger\dagger} = \left(l_2 r_2 + \frac{1}{f_2'(x_2^{(k)})} \right)^{-1} \quad (51)$$

ω_1 changes over iterations k . Similarly, it is shown that:

$$(\omega_2^{(k+1)})^{\dagger\dagger} = \left(l_1 r_1 + \frac{1}{f_1'(x_1^{(k+1)})} \right)^{-1} \quad (52)$$

Note that choosing the coefficients $(\omega_1^{(k+1)})^{\dagger\dagger}$ and $(\omega_2^{(k+1)})^{\dagger\dagger}$ allows to increase the convergence speed. However, the convergence of the Schwarz algorithm is not ensured since the condition (32) is not verified.

155 **3. Unsteady case with general boundary conditions**

156 The two previous sections dealt with:

- 157 • the coupling between two unsteady problems with linear boundary conditions and constant physical
- 158 properties in each domain.
- 159 • the coupling between two steady problems with generic boundary conditions and non constant physical
- 160 properties in each domain.

161 The goal is now to combine the results of these two previous sections to construct a coupling algorithm
 162 for the generic unsteady problem (1). To do so, the coefficients (26a) and (26b) are extended using mean
 163 values of physical properties over each domain. As in section (1), the starting is the time discrete heat
 164 equation:

Semi-discrete heat equation in each domain:

$$\mathcal{L}_1 T_1^{n+1} = \mu_1^2 T_1^n \quad \forall x \in \Omega_1 \quad (53a)$$

$$\mathcal{L}_2 T_2^{n+1} = \mu_2^2 T_2^n \quad \forall x \in \Omega_2 \quad (53b)$$

External boundary conditions:

$$-\lambda_1(-l_1) \frac{dT_1^{n+1}}{dx}(-l_1) = -f_1(T_1^{n+1}(-l_1)) \quad (53c)$$

$$\lambda_2(l_2) \frac{dT_2^{n+1}}{dx}(l_2) = -f_2(T_2^{n+1}(l_2)) \quad (53d)$$

Interfacial conditions:

$$T_1^{n+1}(0) = T_2^{n+1}(0) \quad (53e)$$

$$\lambda_1(0) \frac{dT_1^{n+1}}{dx}(0) = \lambda_2(0) \frac{dT_2^{n+1}}{dx}(0) \quad (53f)$$

165 where $\mu_j(x) = \frac{\rho_j(x)c_j(x)}{\Delta t \lambda_j(x)}$ is now not necessarily uniform and the operator \mathcal{L}_j is defined as:

$$\mathcal{L}_j T = -\frac{1}{\lambda_j} \frac{d}{dx} \left(\lambda_j \frac{dT}{dx} \right) + \mu_j^2 T \quad (54)$$

In addition, the mean values of the parameters are introduced:

$$\widehat{\rho_j c_j} = \frac{1}{l_j} \int_{\Omega_j} \rho_j c_j dx \quad (55a)$$

$$r_j = \frac{1}{l_j} \int_{\Omega_j} \frac{dx}{\lambda_j} \quad (55b)$$

$$\widehat{\mu_j} = \sqrt{\frac{\widehat{\rho_j c_j} r_j}{\Delta t}} \quad (55c)$$

The Schwarz coupling algorithm to solve this problem is given by:

$$\left\{ \begin{array}{l} \mathcal{L}_1 T_1^{(k+1)} = \mu_1^2 T_1^n \quad \forall x \in]-l_1, 0[\quad (56a) \\ -\lambda_1 \frac{dT_1^{(k+1)}}{dx}(-l_1) = -f_1 \left[T_1^{(k+1)}(-l_1) \right] \quad (56b) \\ \lambda_1 \frac{dT_1^{(k+1)}}{dx}(0) = \lambda_2 \frac{dT_2^{(k)}}{dx}(0) + \omega_1 \left(T_2^{(k)}(0) - T_1^{(k+1)}(0) \right) \quad (56c) \end{array} \right.$$

$$\begin{cases} \mathcal{L}_2 T_2^{(k+1)} = \mu_2^2 T_2^n \quad \forall x \in]0, l_2[& (57a) \\ \lambda_2 \frac{dT_2^{(k+1)}}{dx}(l_2) = -f_2 \left(T_1^{(k+1)}(l_2) \right) & (57b) \\ -\lambda_2 \frac{dT_2^{(k+1)}}{dx}(0) = -\lambda_1 \frac{dT_1^{(k)}}{dx}(0) + \omega_2 \left(T_1^{(k+1)}(0) - T_2^{(k+1)}(0) \right) & (57c) \end{cases}$$

166 In order to define the coupling coefficients at iteration $k+1$, the idea is to linearise the boundary conditions
 167 (56b) and (57b) around the solution at iteration k . For example, noting $\Delta_{T,k} = T_1^{(k+1)}(-l_1) - T_1^{(k)}(-l_1)$,
 168 the right hand side of (56b) may be written:

$$-f_1 \left(T_1^{(k+1)}(-l_1) \right) = - \left[f_1 \left(T_1^{(k)}(-l_1) \right) + \Delta_{T,k} f_1' \left(T_1^{(k)}(-l_1) \right) + O(\Delta_{T,k}^2) \right] \quad (58)$$

By analogy with section 1 the coefficients $\eta_j^{(k+1)}$, χ_j and $\omega_j^{(k+1)}$ are defined by:

$$\begin{cases} \eta_1^{(k+1)} = \frac{\widehat{\mu}_1}{r_1 f_1' \left(T_1^{(k)}(-l_1) \right)} & (59a) \\ \eta_2^{(k+1)} = \frac{\widehat{\mu}_2}{r_2 f_2' \left(T_2^{(k)}(l_2) \right)} & (59b) \end{cases}$$

$$\chi_j = \exp(-2\widehat{\mu}_j l_j) \frac{\eta_j^{(k+1)} - 1}{\eta_j^{(k+1)} + 1} \quad (60)$$

$$\begin{cases} \omega_1^{(k+1)} = \frac{\widehat{\mu}_2}{r_2} \frac{1 - \chi_2^{(k+1)}}{1 + \chi_2^{(k+1)}} & (61a) \\ \omega_2^{(k+1)} = \frac{\widehat{\mu}_1}{r_1} \frac{1 - \chi_1^{(k+1)}}{1 + \chi_1^{(k+1)}} & (61b) \end{cases}$$

169 As stated previously, this definition is made by analogy with the unsteady linear problem and there is
 170 no apriori guarantee that the algorithm will converge with this choice of coupling coefficients. However,
 171 as will be shown in the following section, the algorithm performs well in generic conditions such as those
 172 encountered in icing applications. Moreover, one has the following interesting property:

173 **Proposition 5.** *In the limit $\Delta t \rightarrow \infty$, the coupling coefficients (61) are given by (51) and (52), that is to*
 174 *say the optimized values derived for the steady state case.*

175 *Proof.* An asymptotic analysis is required in order to obtain the behaviour of the coupling coefficients when
 176 $\Delta t \rightarrow \infty$. First note that $\lim_{\Delta t \rightarrow \infty} \widehat{\mu}_j = 0$ ($j = 1, 2$). Also, the coupling coefficients are of the form:

$$\omega = \frac{X}{r} \frac{1 - e^{-2lX} \frac{aX-1}{aX+1}}{1 + e^{-2lX} \frac{aX-1}{aX+1}} \quad (62)$$

177 with the following generic notations:

- 178 • X : $\widehat{\mu}_1$ or $\widehat{\mu}_2$
- 179 • l : l_1 or l_2
- 180 • r : r_1 or r_2
- 181 • a : $\left(r_1 f_1' \left(T_1^{(k)}(-l_1) \right) \right)^{-1}$ or $\left(r_2 f_2' \left(T_2^{(k)}(l_2) \right) \right)^{-1}$

182 Given the following Taylor expansions:

$$\frac{aX - 1}{aX + 1} = -1 + 2aX + O(X^2) \quad (63)$$

$$\exp(-2lX) = 1 - 2lX + O(X^2) \quad (64)$$

183 one obtains:

$$\exp(-2lX) \frac{aX - 1}{aX + 1} = -1 + 2(a + l)X + O(X^2) \quad (65)$$

Hence:

$$1 - \exp(-2lX) \frac{aX - 1}{aX + 1} = 2 - 2(a + l)X + O(X^2) \quad (66a)$$

$$1 + \exp(-2lX) \frac{aX - 1}{aX + 1} = 2(a + l)X + O(X^2) \quad (66b)$$

Therefore:

$$\omega = \frac{X}{r} \frac{1 - e^{-2lX} \frac{aX - 1}{aX + 1}}{1 + e^{-2lX} \frac{aX - 1}{aX + 1}} \quad (67a)$$

$$= \frac{X}{r} \frac{2 - 2(a + l)X + O(X^2)}{2(a + l)X + O(X^2)} \quad (67b)$$

$$= \frac{2X + O(X^2)}{2(ra + rl)X + O(X^2)} \quad (67c)$$

Therefore, in the limit $\Delta t \rightarrow \infty$, hence $X \rightarrow 0$, the coupling coefficients reduce to:

$$\omega_1 = \left(l_2 r_2 + \frac{1}{f_2' \left(T_2^{(k)}(l_2) \right)} \right)^{-1} \quad (68a)$$

$$\omega_2 = \left(l_1 r_1 + \frac{1}{f_1' \left(T_1^{(k)}(-l_1) \right)} \right)^{-1} \quad (68b)$$

184 where the values of a , l and r have been replace by their respective counterparts. Therefore, when $\Delta t \rightarrow \infty$,
 185 the coupling coefficients reduce to the steady-state values given by (51) and (52).
 186 □

187 4. Numerical examples

188 In this section, the algorithm is illustrated by considering two test cases. Each test case consists in solving
 189 problem (1), illustrated in Figure 1, with a specific set of material parameters and boundary conditions.
 190 The numerical implementation is based on an implicit finite difference scheme using classical Euler time
 191 discretization and second order spatial finite difference scheme for the diffusion term.

192 4.1. Linear case

193 Consider the problem as illustrated in Figure 2. A block of a given material of length $2l = 0.02m$ is
 194 initially at $T(x, 0) \triangleq T_i = 300K$. Here the block is homogeneous and split only for the purpose of the test
 195 case into two subdomains of length $l_1 = 0.005m$ and $l_2 = 0.015m$.

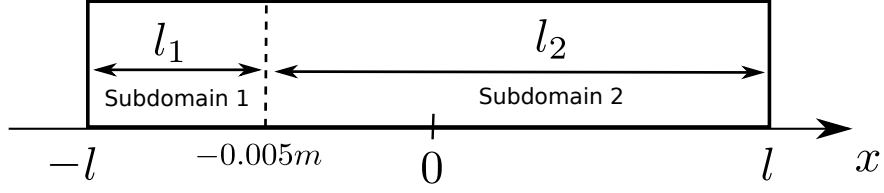


Figure 2: Illustration of the unsteady linear test case

196 The physical properties are uniform and given in Table 1. The block is subjected to convective heat
 197 transfer on both boundaries. The boundary heat transfer characteristics are given in Table 1. In this case
 198 the functions f_1 and f_2 are given by:

$$f_1(T_1(-l)) = h_{tc}(T_1(-l) - T_r) \quad (69a)$$

$$f_2(T_2(l)) = h_{tc}(T_2(l) - T_r) \quad (69b)$$

ρ ($kg.m^{-3}$)	c_p ($J.kg^{-1}.K^{-1}$)	λ ($W.m^{-1}.K^{-1}$)	h_{tc} ($W.m^{-2}.K^{-1}$)	T_r (K)
2700	900	100	400	340

Table 1: Material properties and heat transfer characteristics for the first test case

199 This problem has an analytical solution given by [25, 19]:

$$\frac{T - T_r}{T_i - T_r} = \sum_{n=1}^{\infty} A_n \cos(k_n \frac{x}{l}) e^{-k_n^2 F_o} \quad (70)$$

where

$$F_o = \frac{\alpha t}{l^2} \quad (71a)$$

$$\alpha = \frac{\lambda}{\rho c_p} \quad (71b)$$

200 and

$$A_n = \frac{\sin(k_n)}{k_n + \sin(k_n) \cos(k_n)} \quad (72)$$

201 and the k_n 's are solutions of the equation:

$$k_n \tan(k_n) = B_i \quad (73)$$

202 where B_i is the Biot number defined by: $B_i = \frac{h_{tc} l}{\lambda}$.

203 In order to test the performance of the algorithm, the domain is split at $x = -0.005m$. The solution
 204 can then be obtained by solving the heat equation on each sub-domain and using the unsteady coupling
 205 procedure previously defined (56 and 57). The computation is run with a time step of 0.1s and optimized
 206 coupling coefficients (given by equations (61a) and (61b)). At every time step, the coupling procedure is
 207 performed until the relative temperature difference and relative difference of heat flux at the interface both
 208 have values below 10^{-6} .

209 As shown in Figure 3, the numerical solution is in excellent agreement with the analytical one (given by
 210 equation (70)). The convergence criterion was met within two coupling iterations over the whole simulation.

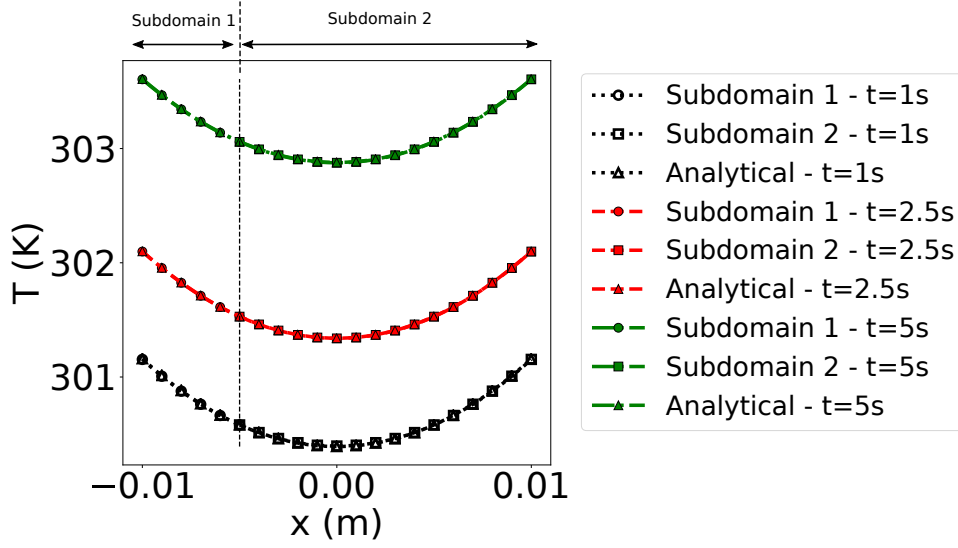
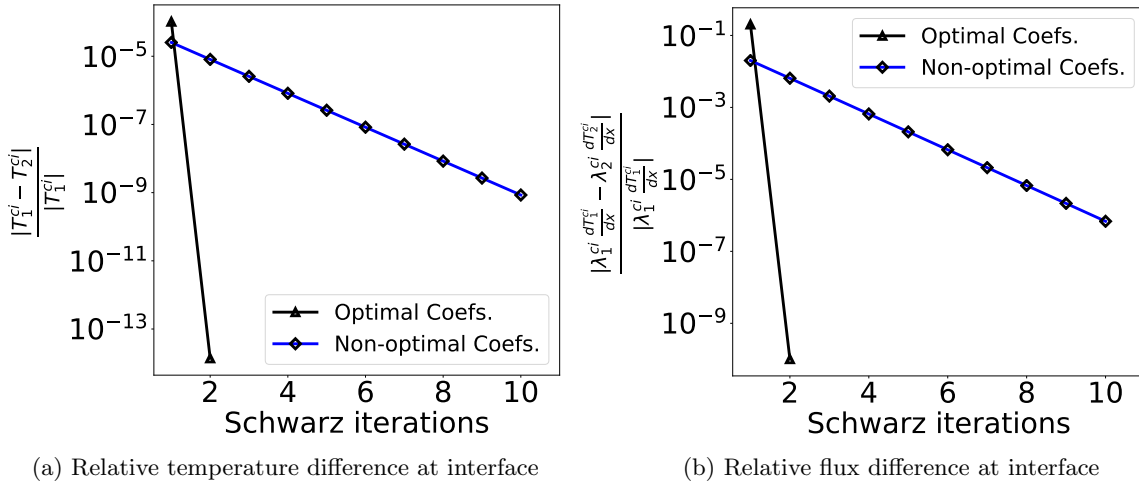


Figure 3: Comparison of numerical and analytical solutions (using optimized coupling coefficients)

211 In addition, one may also evaluate the numerical convergence properties of the algorithm over one time
 212 step. Figure 4 shows the relative differences in temperature and heat flux at the interface, at each iteration
 213 of the Schwarz algorithm and at $t = 1s$. In order to perform a comparative study the simulation is also
 214 run with non optimal coefficients. The coupling coefficients given by (39a) and (39b) are chosen as the non-
 215 optimal ones. When using optimal coefficients, the rate of convergence is much higher. After two iterations
 216 the differences is well below the acceptable tolerance for practical applications. This is coherent with the fact
 217 that in the semi-discrete case, these optimal coefficients guarantee convergence in one iteration (as discussed
 218 at the end of section 1). On the other hand, the use of non-optimal coupling coefficients induces a lagrer
 219 amount of iterations to satisfy the convergence criterion.



(a) Relative temperature difference at interface

(b) Relative flux difference at interface

Figure 4: Relative differences in temperature and flux and the interface using optimal and non-optimal coupling coefficients (at $t = 1s$). The superscript ci means that the values are taken at the coupling interface.

220 4.2. Non-linear case

221 The case considered in this section, illustrated by Figure 1, consists of two different materials in contact
 222 each with uniform physical properties given in Table 2. One of them is only subjected to convective heat
 223 transfer while the other is also subject to evaporation. In this case, the function f_1 and f_2 are given by:

$$f_1(T_1(-l_1)) = \dot{m}_{ev}(T_1(-l_1))(c_{p,1}T_1(-l_1) + L_v(T_1(-l_1))) + h_{tc,1}(T_1(-l_1) - T_{r,1}) \quad (74a)$$

$$f_2(T_2(l_2)) = h_{tc,2}(T_2(l_2) - T_{r,2}) \quad (74b)$$

224 where \dot{m}_{ev} is the evaporation rate and L_v is the latent heat of vaporisation (see Appendix A for further
 225 details. For this case $T_\infty = 280K$, $p_\infty = 98000Pa$ and $p_e = 98000Pa$). Note that the evaporation rate is a
 226 non linear function of temperature.

Material	ρ_i ($kg.m^{-3}$)	$c_{p,i}$ ($J.kg^{-1}.K^{-1}$)	λ_i ($W.m^{-1}.K^{-1}$)	$h_{tc,i}$ ($W.m^{-2}.K^{-1}$)	$T_{r,i}$ (K)
1	1000	4181	0.6	200	313.15
2	2700	900	167	500	300

Table 2: Material properties and heat transfer characteristics for the second test case ($i = 1, 2$)

227 In this case, f_1 is non-linear. Figure 5 shows the numerical solution at $t = 1s$, $t = 5s$ and $t = 10s$.
 228 Due to the evaporative cooling, the temperature at the left boundary decreases rapidly. At every time step,
 229 the coupling procedure is performed until the relative temperature difference and relative difference of heat
 230 flux at the interface both have values below 10^{-6} . The convergence criterion was met within two coupling
 231 iterations over the whole simulation.

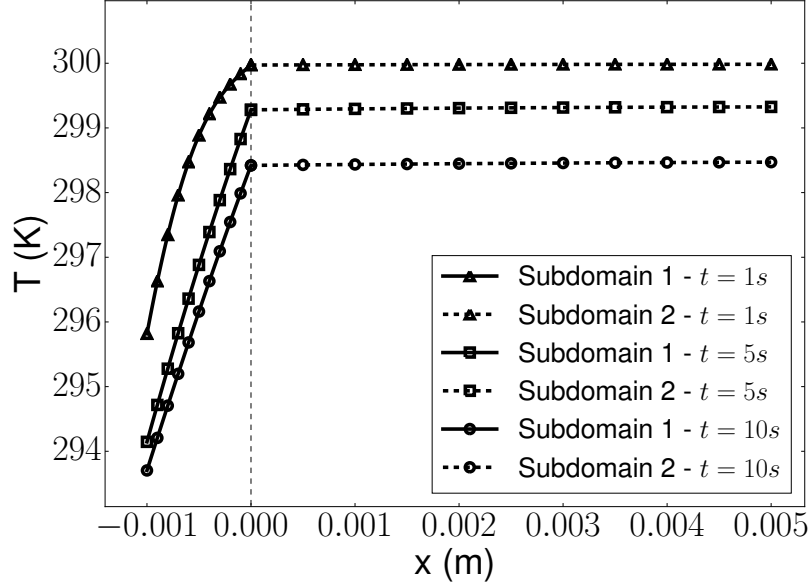


Figure 5: Temperature field at $t = 1s$, $t = 5s$ and $t = 10s$

232 In order to evaluate the convergence properties in this case, the procedure of the previous section is
 233 repeated (for $t = 5s$). The results are shown in Figure 6. As in the previous case, the optimized set of
 234 coefficients yields a much higher rate of convergence. Note that in this case, the theoretical background
 235 regarding the convergence properties is much weaker than in the linear case. Therefore, this test case also
 236 serves as a numerical investigation of the convergence properties of the methodology in a general setting.

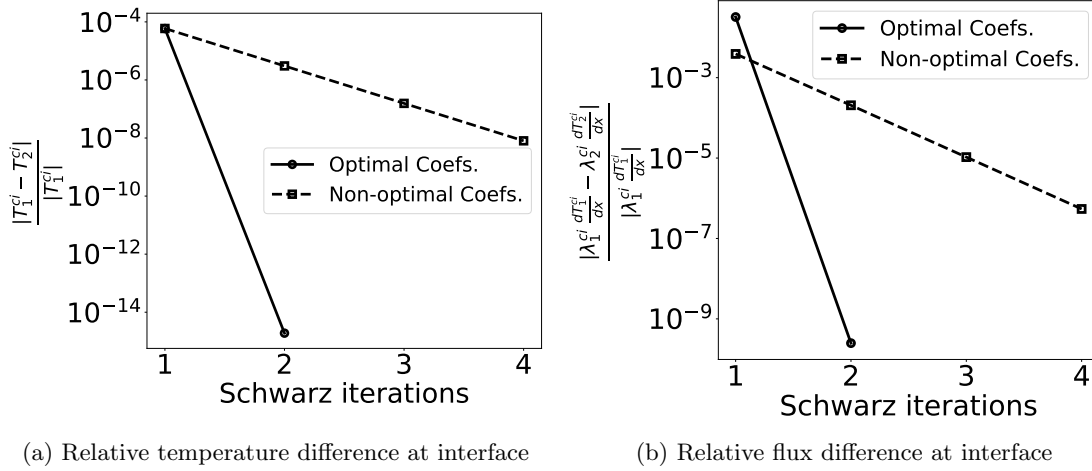


Figure 6: Relative differences in temperature and flux and the interface using optimal and non-optimal coupling coefficients (at $t = 5s$). The superscript ci means that the values are taken at the coupling interface.

237 5. Application to Electrothermal De-Icing

238 An application to the phenomenon of aircraft icing and ice protection is now considered. Icing is due to
 239 the presence in clouds of supercooled water droplets. Upon impacting an aircrafts surface, the metastable
 240 supercooled state of the droplets is broken. At that point, the droplets undergo a liquid-solid phase change,
 241 leading to ice build-up on the impinging surface. This phenomenon has many undesirable consequences and
 242 manufacturers must therefore equip their aircraft with ice protection systems.

243 This case deals with the simulation of electrothermal de-icing. First, let's recall the typical operating of
 244 an electro-thermal de-icing system. Consider the case where an airfoil has to be protected from the build
 245 up of ice on its surface. To do so, one may place several heater mats within the thickness of the airfoil and
 246 activate them according to a given power cycle. During this power cycle ice may build up or melt and liquid
 247 water may run downstream under the effect of the aerodynamic forces. This may lead to several possible
 248 states as illustrated in Figure 7b [7].

249 The simulation of the thermal behaviour of an electro-thermal de-icing system therefore requires two
 250 solvers. One that solves the heat conduction problem in the skin of the airfoil, composed of several layers of
 251 materials and electrical heaters. Another that solves the unsteady phase change problems of ice accretion
 252 and melting. In this work, a finite volume solver called ETIPS2D is used to simulate the electro-thermal
 253 system (see [2] for a description). Concerning the unsteady ice accretion and melting problem, an unsteady
 254 mixed finite volume-Galerkin solver called MiLeS2D is used (see [7] for a detailed description). A global
 255 illustration is shown in Figure 7b.

256 In order to perform the simulation of the electro-thermal de-icer in icing conditions, the two previously
 257 mentionned solvers are coupled using the Schwarz method presented in section 3. Here the simulation is
 258 two dimensional and both meshes are coincident. The coupling method is extended by performing a locally
 259 one dimensional coupling. More precisely, for each interfacial edge only the normal cells to that edge are
 260 used to compute the coupling coefficients (see Figure 7a).

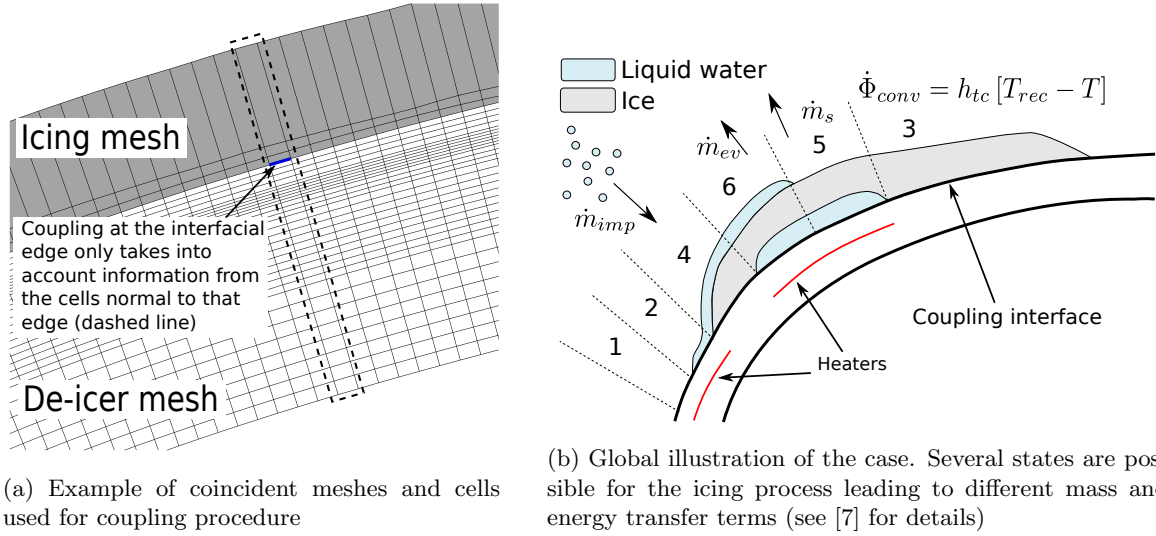


Figure 7: Differences in temperature and flux and the interface at $t = 39.1s$ showing convergence with a low number of iterations.

Before moving to the test case, note that MiLeS2D has a complex algorithmic structure, the explicit tracking of phase change fronts requiring to take into account several possible states and possible shifts between these states. The six possible states, illustrated in Figure 7b, are defined as:

1. Full evaporative: the whole mass of impacting droplets is evaporated (for example, due to heat provided by an ice protection system).
2. Running wet: only a liquid water film is present. Under the action of the aerodynamic forces, the liquid film runs back along the surface.
3. Rime accretion: the droplets freeze almost instantaneously leading to ice build up with no liquid water.
4. Glaze accretion: the droplets freeze, but at a slower rate than in the rime case. Therefore, a running liquid water film is present on top of the ice layer.
5. Rime accretion with melting at the surface (due to heat provided by an ice protection system for example).
6. Glaze accretion with melting at the surface (due to heat provided by an ice protection system for example).

So as to illustrate the complexity of the algorithm, suppose for example that an ice layer is present (rime ice state). If the heater mat is activated, this will lead to the melting of the ice. This means that at that point in time, the state must be switched from rime ice to rime ice with static film. More precisely, at each time step, the algorithm will first start by assuming the same state as in the previous time step. If this yields a result compatible with the current state, the solution is conserved and the algorithm proceeds. However, if an incompatibility is detected (for example the temperature of the ice is greater than the melting temperature), the algorithm switches to another mode. As will be discussed later, this can lead to jumps in the Schwarz coupling procedure. Indeed, switching from one state to another changes the flux at the interface and hence directly impacts the coupling algorithm. Figure 8 illustrates the general architecture for the simulation.

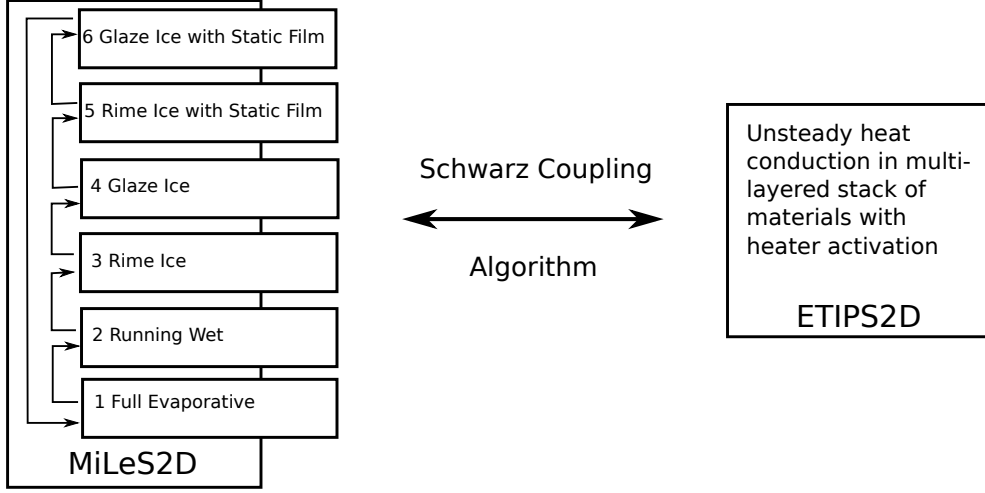


Figure 8: General architecture of the numerical simulation procedure involving MiLeS2D and ETIPS2D.

285 *5.1. Description of the case*

286 This illustrative test case consists of a de-iced NACA0012 airfoil in icing conditions. The droplets start
 287 impacting after 5s of activation of the system. The aerodynamic and icing conditions are given in Table 3.

Chord (m)	α ($^\circ$)	T_∞ (K)	P_∞ (Pa)	M_∞	MVD (μm)	LWC ($g.m^{-3}$)
0.6	0	253.15	79470	0.5	20	0.34

Table 3: Aerodynamic and icing conditions for the de-iced NACA0012 case

288 The airfoil is made out of a multi-layered material whose properties are provided in Table 4. The layers
 289 are numbered from outer to inner as shown in Figure 9.

Layer no	c_p ($J.kg^{-1}.K^{-1}$)	ρ ($kg.m^{-3}$)	λ ($W.m^{-1}.K^{-1}$)	thickness (m)
1	2350	1000	17.03	8.e-04
2	2009	1000	0.293	3.e-04
3	2381	1000	0.313	1.e-04
4	2009	1250	0.293	2.e-04
5	2009	1250	0.293	5.e-04
6	1717	1000	0.25	2.52e-03
7	1717	1000	0.25	4.e-03

Table 4: Layer thicknesses and material properties

290 Heater mats are embedded into the stack of materials between layers 4 and 5. They are shown in Figure
 291 9 and are labelled PS, HM1 and HM2. The heater mat located at the leading edge is comonly refered to
 292 as the parting strip (hence the label PS). The locations of the heater mats (in terms of curvilinear abscissa,
 293 where $s = 0$ at the leading edge) are given by:

- 294 • PS located between $s_1 = -0.015625 m$ and $s_2 = 0.015625 m$.
- 295 • HM1 located between $s_1 = -0.048875 m$ and $s_2 = -0.017625 m$.
- 296 • HM2 located between $s_1 = 0.017625 m$ and $s_2 = 0.048875 m$.

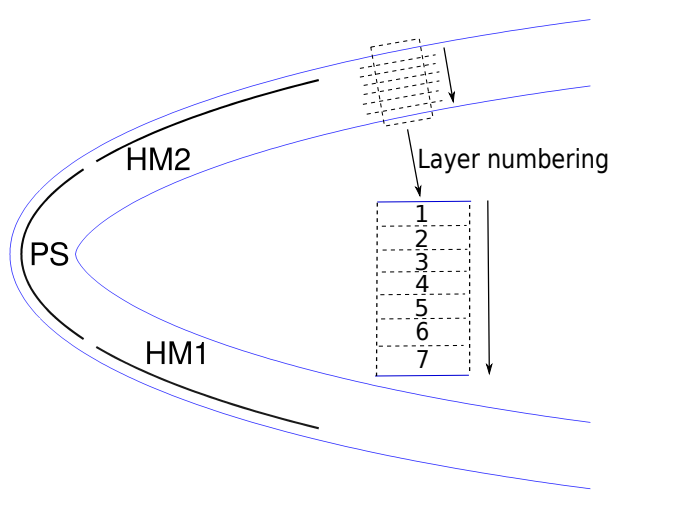


Figure 9: De-icing heater layout

297 Numerically, the coupling method requires a convergence criterion. To do so, the following quantities
 298 are considered:

$$\epsilon_T = \max_{\Gamma_i} (|T_{\Gamma_i}^{MiLeS} - T_{\Gamma_i}^{ETIPS}|) \quad (75)$$

299 where $T_{\Gamma_i}^{MiLeS}$ and $T_{\Gamma_i}^{ETIPS}$ are the temperatures at the interfacial edge Γ_i computed by MiLeS2D and
 300 ETIPS2D respectively. The maximum is taken over all edges of the coupling interface.

$$\epsilon_\phi = \max_{\Gamma_i} \left(\frac{|\phi_{n_i}^{MiLeS} - \phi_{n_i}^{ETIPS}|}{|\phi_{n_i}^{ETIPS}|} \right) \quad (76)$$

301 where $\phi_{n_i}^{MiLeS}$ and $\phi_{n_i}^{ETIPS}$ are the heat fluxes normal to the i -th coupling interface Γ_i , computed by MiLeS2D
 302 and ETIPS2D respectively. In this case, the coupling loops is considered converged if ϵ_T and ϵ_ϕ fall below
 303 0.001.

304 This type of simulation is embedded within a larger simulation chain called IGLOO2D [23]. Computations
 305 of the inviscid flow around the airfoil, boundary layer flow and droplet trajectories are performed prior to
 306 the presented coupled simulation. Relevant data are given in Appendix B.

307 Two variants of this configuration are simulated. The differences are in the heater mat power cycles.
 308 They are therefore referred to as "activation cycle 1" and "activation cycle 2". In both cases, during the first
 309 five seconds of simulation the mat PS is activated with no droplet impact. The cases are symmetric with
 310 respect to the x -axis and therefore the results are presented only at the upper part of the airfoil.

311 5.2. Activation cycle 1

312 This activation power cycle ($t = 0s$ corresponds to the beginning of the cycle and the whole cycle lasts
 313 for 80s) is given by:

- 314 • PS activated at $t = 0s$ with a power density of $50kW.m^{-2}$. Activation lasts 80s.
- 315 • HM1 activated at $t = 40s$ with a power density of $25kW.m^{-2}$. Activation lasts 40s.
- 316 • HM2 activated at $t = 40s$ with a power density of $25kW.m^{-2}$. Activation lasts 40s.

317 Figure 10 shows the resulting temperature field at $t = 5s$, just before droplets start to impact. The
 318 region where the heater mat's power is applied is clearly visible.

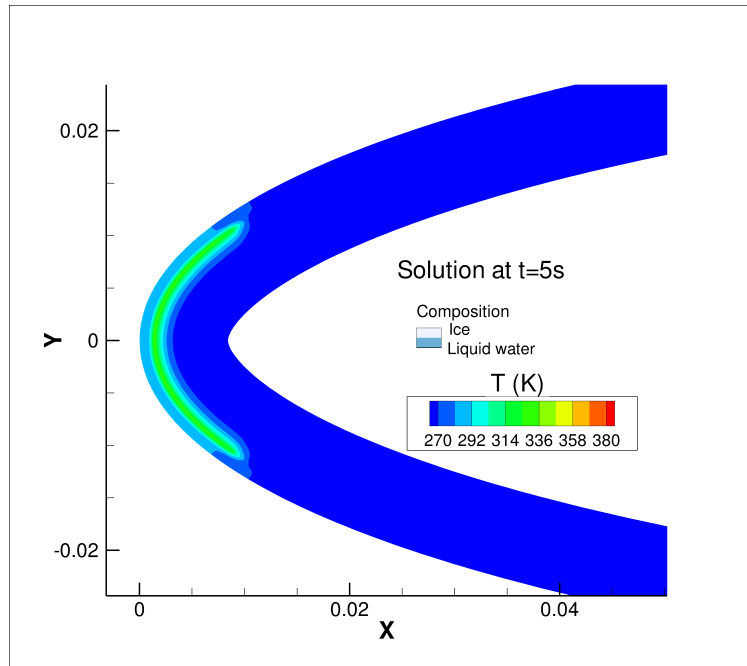


Figure 10: Solution at $t = 5s$

319 At $t = 5s$, droplets start to impact the airfoil and ice starts to build-up beyond the region heated by PS.
 320 This can be observed in Figure 11 which shows the temperature field in the airfoil, the ice shape and the
 321 dynamic liquid film (height scaled by a factor 500 for visualisation purposes) at $t = 40s$.

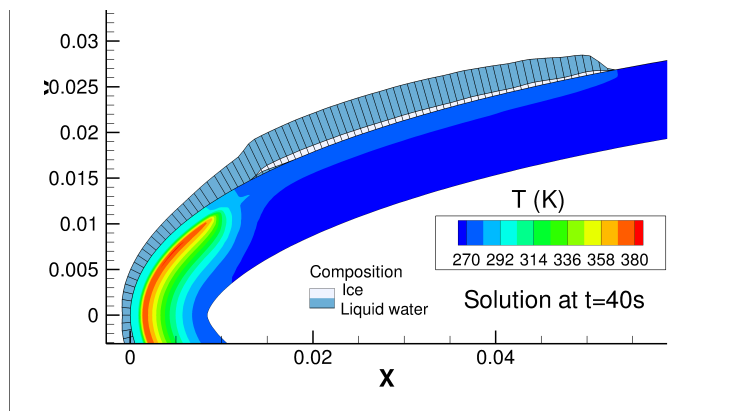


Figure 11: Solution at $t = 40s$ (the running film height is scaled by a factor 500 for visualisation purposes)

322 At $t = 40s$ the heaters HM1 and HM2 are activated. The heat melts the ice layer above those mats.
 323 Therefore, a static film layer is created underneath the ice, as shown in Figure 12.

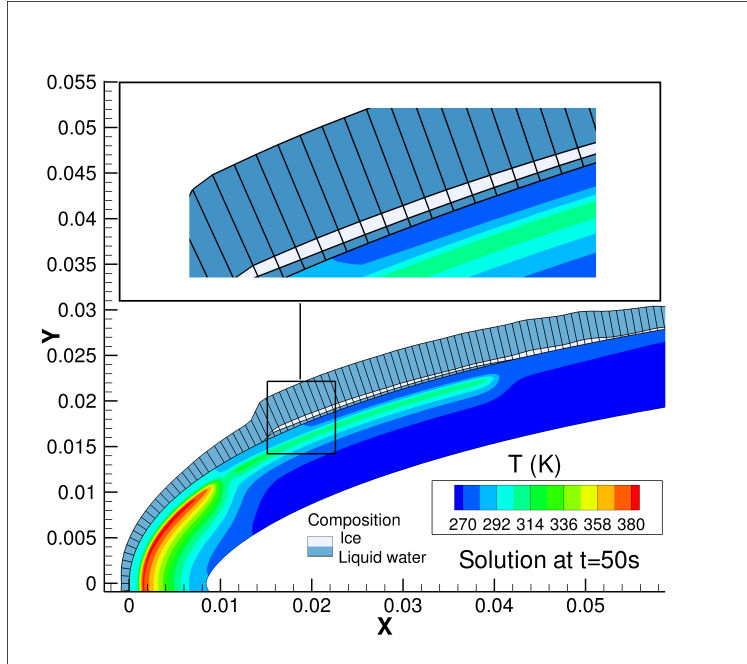


Figure 12: Solution at $t = 50s$ (the running film height is scaled by a factor 500 for visualisation purposes)

324 The activation of HM1 and HM2 eventually melts all the ice layer in their respective heated regions. At
 325 the end of the simulation ($t = 80s$) water has now runback and frozen beyond the protected region (see
 326 Figure 13).

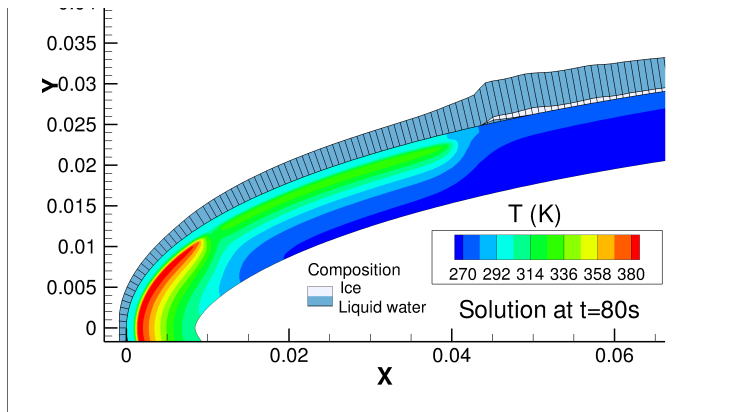


Figure 13: Solution at $t = 80s$ (the running film height is scaled by a factor 500 for visualisation purposes)

327 As stated earlier, the ice accretion model is complex and non-linear. Moreover, the heater mats are
 328 taken into account thanks to a source term which was not accounted for in the theoretical derivation of the
 329 Schwarz coupling method. There is no theoretical guarantee of the convergence of the coupling procedure.
 330 Nevertheless, the situation is close enough to those of sections 1 and 2 for the method to exhibit globally
 331 good convergence properties in practice. To illustrate this point let's consider two extreme convergence
 332 cases.

333 Figure 14 shows the convergence curve for the absolute difference in temperature and relative difference
 334 in flux at the interface at $t = 39.1s$. At that time step, the coupling procedure converges rapidly. The values
 335 fall below the given tolerance of 0.001 after five iterations.

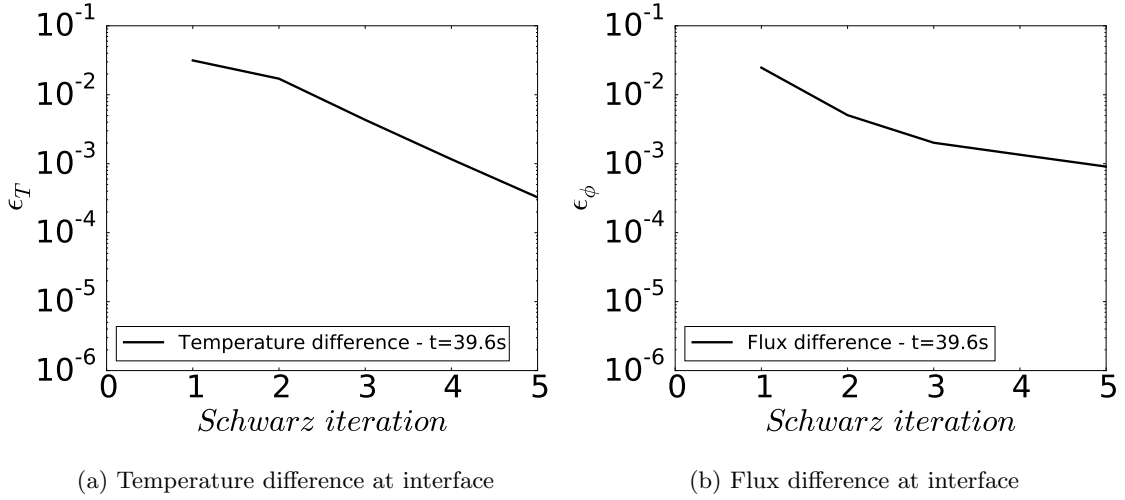


Figure 14: Differences in temperature and flux and the interface at $t = 39.1s$ showing convergence with a low number of iterations.

336 On the other hand, certain time steps may require more iterations to converge. As discussed previously
 337 (see Fig. 8), depending on the thermal state of the system the ice accretion solver may be required to
 338 switch modes in order to obtain the solution at a given time step. As this changes the fluxes at the coupling
 339 interface, this can induce convergence issues as shown in Figure 15. Nevertheless, convergence is still reached.

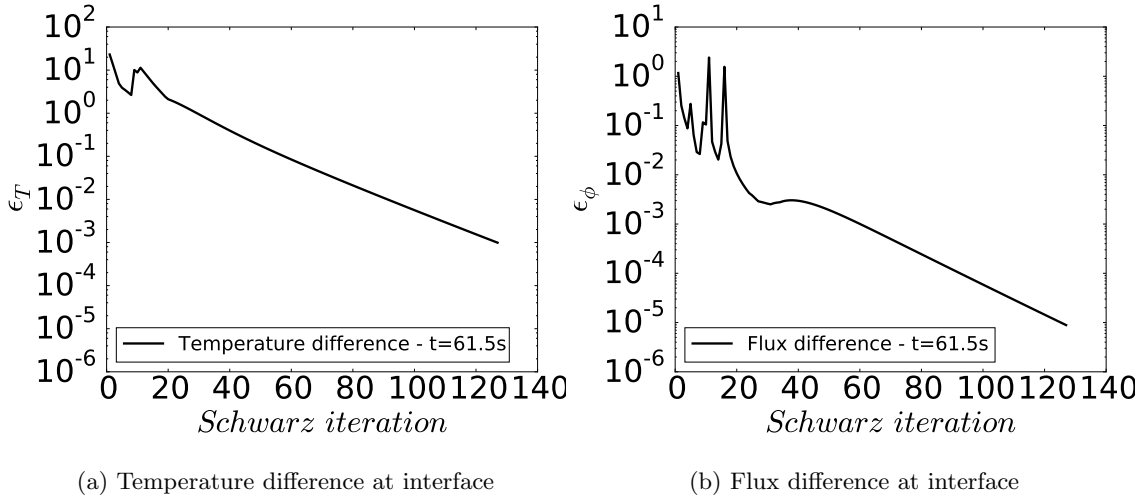


Figure 15: Differences in temperature and flux and the interface at $t = 61s$ showing difficulty in convergence.

340 One may ask which behaviour is most frequently observed in practice. Figure 16 is a histogram repre-
 341 senting, over the whole simulation, the number of times a given amount of Schwarz iterations was required
 342 for convergence. The average number of Schwarz iterations was 17. Moreover, Figure 16 shows that the
 343 most frequent behaviour is a number of Schwarz iterations less than 20. Hence, the cases where convergence
 344 is highly slowed down by mode switching remain reasonable.

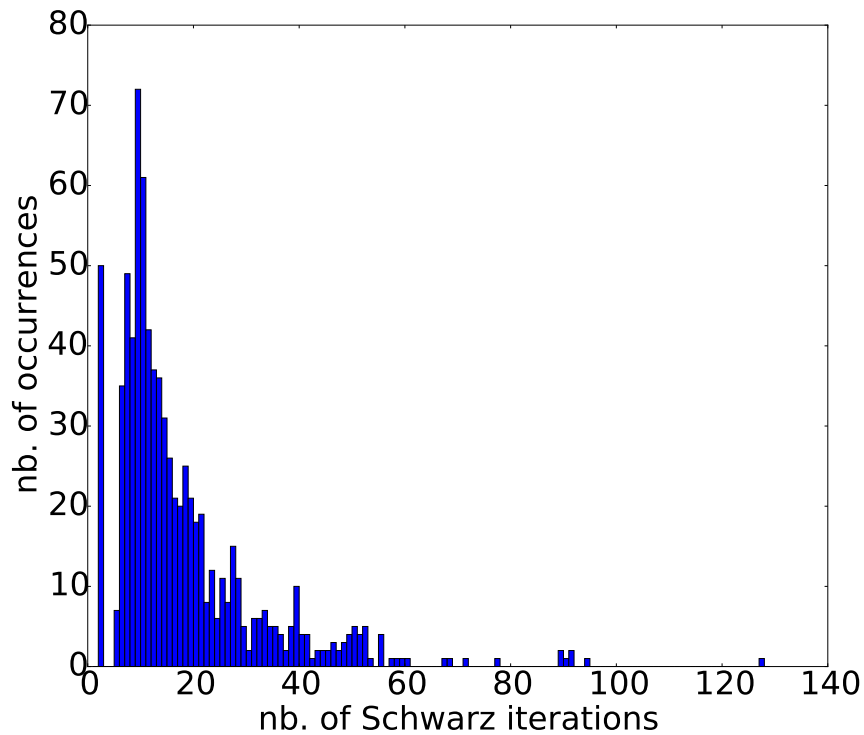


Figure 16: Number of occurrences for each possible number of Schwarz iterations (activation cycle 1). The average number of iterations is 17.

345 To get a more complete view of the convergence of the coupling algorithm during the simulation the
 346 number of iterations required at each time step to reach the desired convergence criterion is shown in Figure
 347 17. Even though some spikes with a high number of iterations are visible, 20 to 25 iterations are usually
 348 sufficient.

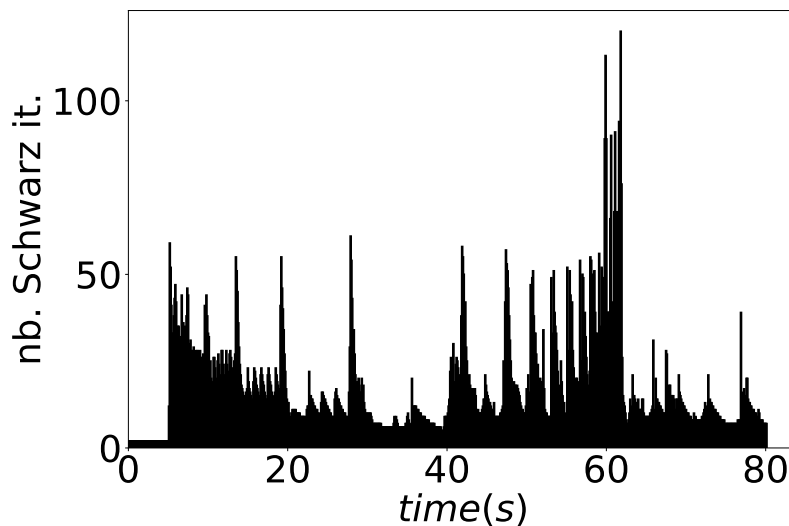


Figure 17: Number of Schwarz iterations at every time step (activation cycle 1).

349 *5.3. Activation cycle 2*

350 In second activation power cycle HM1 and HM2 are deactivated then reactivated during the cycle. The
 351 cycle is given by:

- 352 • PS activated at $t = 0s$ with a power density of $50kW.m^{-2}$. Activation lasts 80s.
- 353 • HM1 activated from $t = 40s$ to $t = 50s$ and reactivated from $t = 60s$ to $t = 70s$ with a power density
 354 of $25kW.m^{-2}$.
- 355 • HM2 activated from $t = 40s$ to $t = 50s$ and reactivated from $t = 60s$ to $t = 70s$ with a power density
 356 of $25kW.m^{-2}$.

357 Up to $t = 50s$ this case is identical to the previous one. At $t = 50s$ heaters HM1 and HM2 are deactivated.
 358 Ice is able to continue to build up while the skin temperature decreases. Figure 18 shows that at $t = 60s$,
 359 just before the reactivation of HM1 and HM2, a small layer of ice is still present.

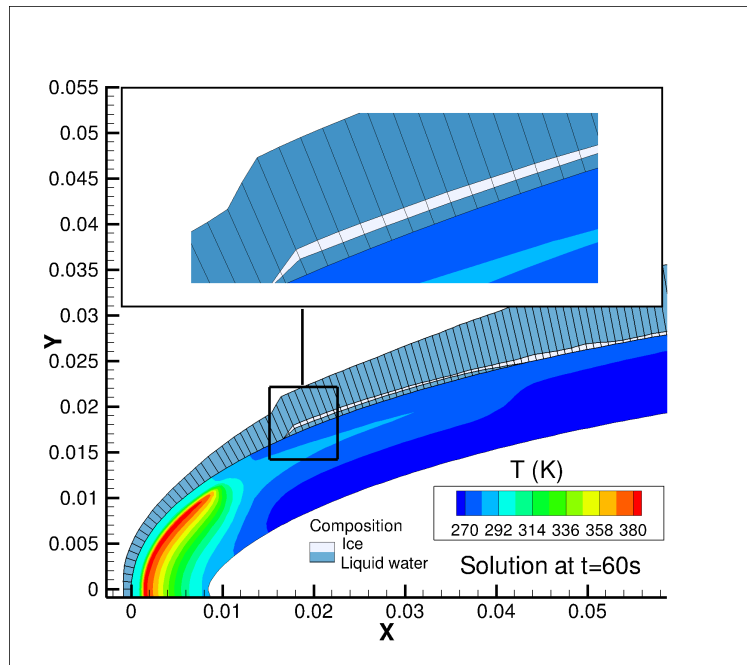


Figure 18: Solution at $t = 60s$ (the running film height is scaled by a factor 500 for visualisation purposes)

360 This layer of ice is progressively melted by the heat provided by HM1 and HM2 after their reactivation.
 361 Figure 19 shows the result at $t = 72s$. Even though HM1 and HM2 are now deactivated again, the thin ice
 362 layer continues to melt due to thermal inertia.

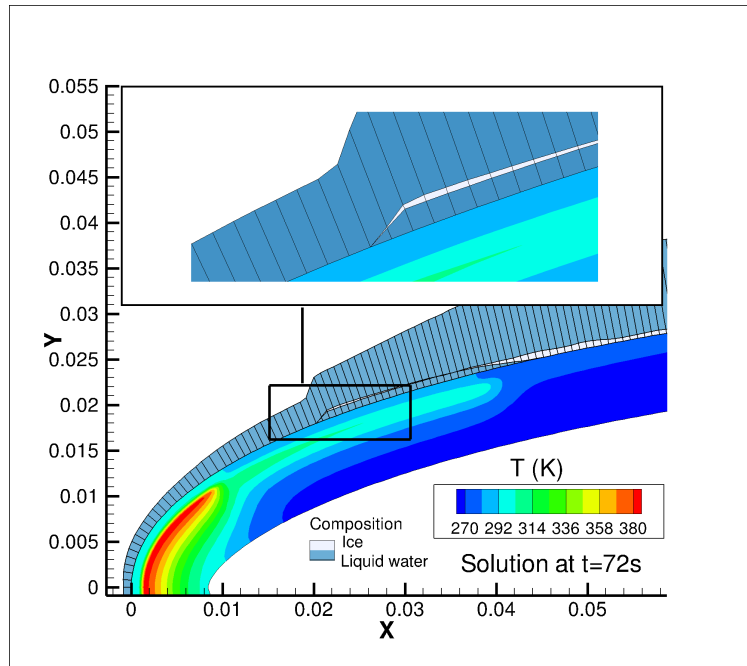


Figure 19: Solution at $t = 60s$ (the running film height is scaled by a factor 500 for visualisation purposes)

363 At $t = 80s$ the area where the heater mats are located is completely free of ice. However, the running
 364 back liquid film generates ice build up beyond the protected area.

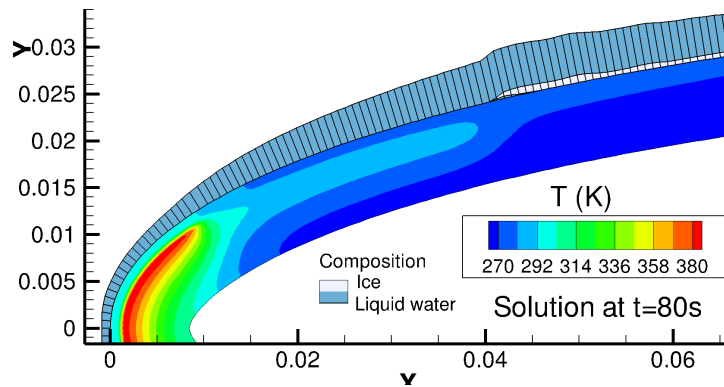


Figure 20: Solution at $t = 80s$ (the running film height is scaled by a factor 500 for visualisation purposes)

365 As in the previous case, Figure 21 is a histogram representing, over the whole simulation, the number of
 366 times a given amount of Schwarz iterations was required for convergence. The behaviour is similar. For some
 367 time steps a large number of coupling iterations was required. However the average number of iterations
 368 was 18.

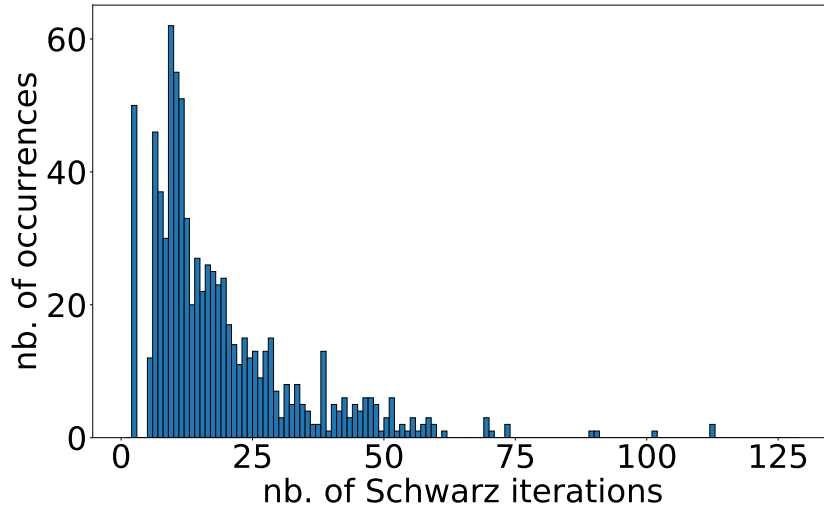


Figure 21: Number of occurrences for each possible number of Schwarz iterations (activation cycle 2).

369 Finally, Figure 22 shows the number of coupling iterations required to reach the convergence criterion
 370 at every time step. The spikes reflect the fact that for some time steps the convergence of the coupling
 371 procedure is slow. Nevertheless, in most cases, 25 iterations are sufficient.

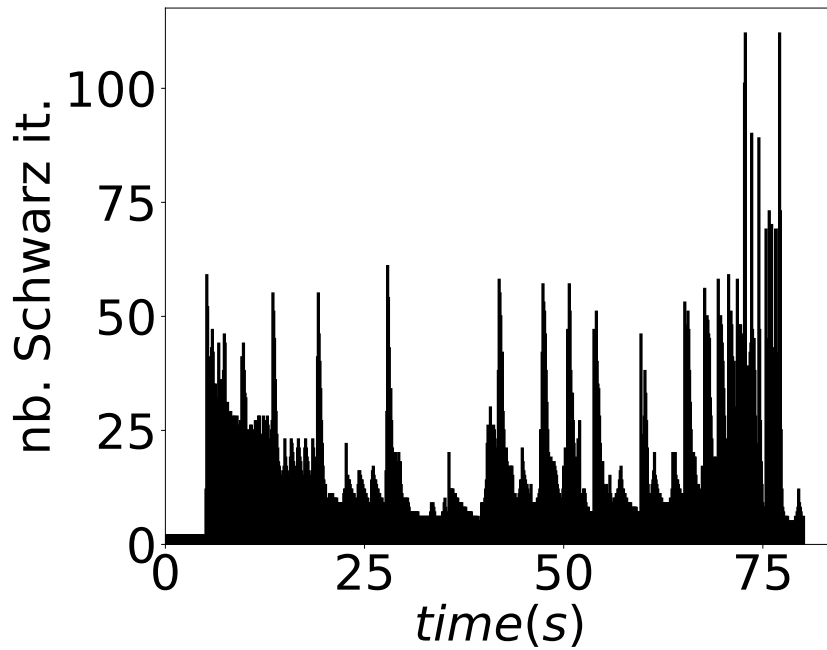


Figure 22: Number of Schwarz iterations at every time step (activation cycle 2).

372 6. Conclusion

373 This paper focuses on a Schwarz coupling methodology. The goal was to obtain coupling coefficients
 374 which could be used to solve coupled heat transfer problems in generic settings. To do so, two cases where

375 theoretical derivations are possible were presented: the unsteady case with linear boundary conditions and
376 the steady case with general boundary conditions. The coefficients obtained with these derivations guarantee
377 the convergence of the algorithm. Moreover, the theoretical derivation also provided the framework to
378 obtain optimized coupling coefficients. In order to treat unsteady cases with generic boundary conditions,
379 the coefficients obtained for linear boundary conditions are extended using a linearisation of the boundary
380 conditions.

381 Two numerical test cases are then performed in order to assess the effective convergence properties of
382 the coupling algorithm. The results for the first test case showed an excellent agreement with the analytical
383 solution. Moreover, for both test cases, the algorithm was shown to efficiently converge. In addition, the
384 numerical investigation also showed that using the optimal set of coefficients yielded a much higher rate of
385 convergence.

386 Finally, an electro-thermal de-icing case was simulated using this method to couple two unsteady solvers.
387 Although in some occurrences convergence was slowed down by the mode switching of the ice accretion solver,
388 the method performed well. Some improvements, such as accounting for the heater mat source terms in the
389 derivation of the coupling coefficients, are possible and are part of ongoing work.

390 The perspectives of this work is to be able to fully simulate the behaviour of an electrothermal ice
391 protection system. One of the last elements to be added is the modelling, simulation and understanding of
392 the ice shedding phenomenon. This requires a mechanical model of the ice blocks and poses the problem
393 of characterizing the physical properties of atmospheric ice. Some work in this direction has already been
394 performed [4, 3]. The next step is to integrate this model into the present coupling procedure.

395 7. Acknowledgements

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398 edged.

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448 Appendix A. - Evaporative boundary condition

449 In sections 4.2 and 5 an evaporative boundary condition is used. It involves two functions $\dot{m}_{ev}(T)$ and
450 $L_v(T)$ representing respectively the rate of evaporation and the latent heat of vaporization which are defined
451 as:

$$L_v(T) = 4185 (T_{boil} - T) + L_{v,boil} + 1850 (T - T_{boil}) \quad (\text{A.1})$$

$$\dot{m}_{ev}(T) = h_m (Y_v(T) - Y_v(T_\infty)) \quad (\text{A.2})$$

452 where

$$h_m = \frac{h_{tc}}{c_{p,air} L_e^{2/3}} \quad (\text{A.3})$$

453 The Lewis number is defines as the ratio of the Schmidt and Prandtl numbers.

$$L_e = \frac{S_c}{P_r} \quad (\text{A.4})$$

454 where the Schmidt number is set to $S_c = 0.67$ and the Prandtl number to $P_r = 0.7$.

455 The vapour mass fraction is linked to the mole fraction and molar masses using:

$$Y_v(T) = \frac{X_v(T)}{X_v(T) + \frac{M_a}{M_v} (1 - X_v(T))} \quad (\text{A.5})$$

and the mole fraction is given by:

$$\begin{cases} X_v(T) = \frac{p_v(T)}{p_e} & (\text{A.6a}) \\ X_v(T_\infty) = \frac{p_v(T_\infty)}{p_\infty} & (\text{A.6b}) \end{cases}$$

456 Finally, the vapour pressure is given by Sonntag’s law:

$$p_v(T) = -\frac{6024.5282}{T} + 29.32707 + 0.010613868 T - 1.3198825 \times 10^{-5} T^2 - 0.49382577 \log(T) \quad (\text{A.7})$$

457 **Appendix B. - Boundary layer and droplet impingement data for de-icer simulation**

458 The boundary layer and droplet impingement data used for the de-icer simulation were obtained with
 459 the IGLOO2D icing code and are shown in Figure B.23. The boundary layer computation is performed with
 460 a smooth wall hypothesis.

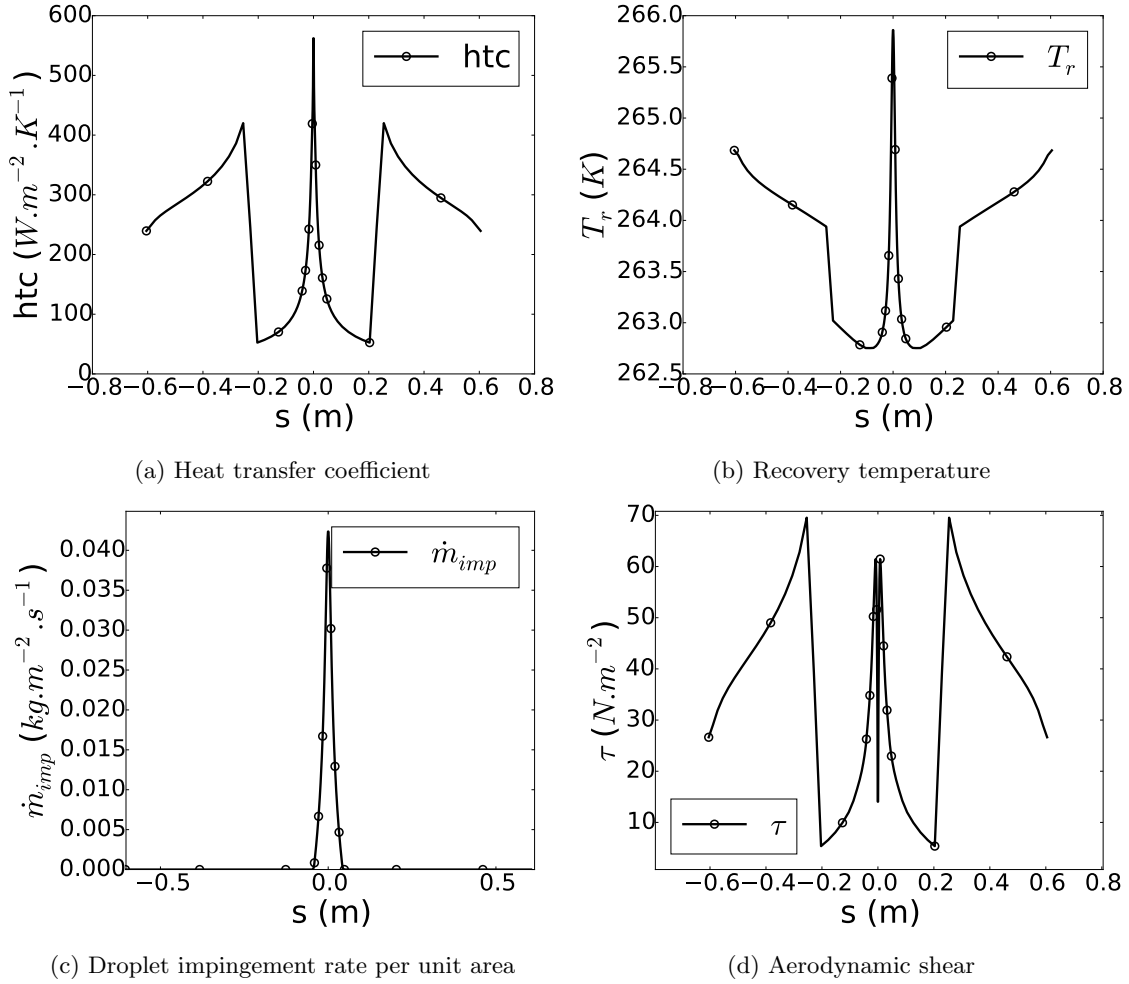


Figure B.23: Boundary layer and droplet impingement data for de-icer simulation