# Advanced computational vademecums for Automated Fibre Placement processes

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Abstract Automated Fibre Placement (AFP) is an incipient manufacturing process for composite structures. Despite its conceptual simplicity it involves many complexities related to the necessity of melting the thermoplastic at the interface tape-substrate, ensuring the consolidation that needs the diffusion of molecules and control the residual stresses installation responsible of the residual deformations of the formed parts. The optimization of the process and the determination of the process window requires many simulations because there are too many parameters involved in the characterization of the material and the process. The exploration of the design space cannot be envisaged by using standard simulation techniques. In this paper we propose the offline calculation of rich parametric solutions that can be then explored online in real time in order to perform inverse analysis, process optimization or online simulation-based control.

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#### 1 Introduction

The production of large pieces made of thermoplastic composites is a challenging issue for today's industry. Thermoplastic composites still represent a niche market because of the difficulties associated to their processing. Several reliable manufacturing processes are now available for building-up thermoplastic laminated structures. Among them, the automated tape placement (ATP) appears to be an appealing process. In this process a tape is placed and progressively welded on the substrate consisting in the tapes previously placed. By placing additional layers in different directions, a part with desired properties and geometry can be produced. However, the welding of two thermoplastic layers requires specific physical conditions: a permanent contact, also called intimate contact, and a temperature that has to be high enough during a large enough time interval to ensure the diffusion of macromolecules, without significant material degradation. Due to the low thermal conductivity of thermoplastics, a high temperature at the interface can be reached with a local heating. ATP uses a laser and a cylindrical consolidation roller to ensure both conditions required for the proper welding, as depicted on Fig. 1.

The numerical simulation of such a process is the subject on an intensive research work. Indeed, because of the successive heating and cooling of the structure during the addition of new tapes, residual stresses appear in the formed part. The evaluation of these residual stresses is crucial because they have a significant impact on both the mechanical properties and the geometry of the manufactured plate or shell due to the springback. In order to evaluate and control the installation of such residual stresses an accurate evaluation of the thermal history is compulsory.

Several models were proposed since the early 90's. We can mention in particular the numerical analysis made by Sonnez et al. [19] and the work by Pitchumani et al. [15] interested in the study of interfacial bonding. In the latter, the domain considered is only 2d and strong assumptions were introduced in the thermal model, in particular concerning the boundary conditions. Moreover, in order to simplify the geometry of the domain, an incoming tow was assumed instantaneously laid down all along the substrate, which is far from being the case in the real process. Finally, the thermal/mechanical contact was assumed to be perfect at the inter-ply interfaces, which again seems to be also a crude assumption. First attempts of the modeling and simulations of this process can be found in [18] [14].

In [8] we proposed some improvements to existing models. First of all, the domain was consider is 3d and the material anisotropic. In order to take into account the imperfect adhesion at the inter-ply interface, thermal contact resistances were introduced. Regarding the mechanical problem, the incoming tow was progressively laid down on the substrate and was subjected to a ten-

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sion force in order to reproduce the pre-tension applied in the real process. But actually, beyond the model itself, the numerical method employed for the solution of the thermal and mechanical problems associated to the ATP process was novel. That work represented a first step towards a global thermomechanical process modeling using robust and efficient numerical tools. The numerical strategy we proposed was based on the Proper Generalized Decomposition (PGD) [1] [2]. This method uses a separated representation of the unknown field, in that case temperature or displacements, and results in a tremendous reduction of the computational complexity of the model solution. Moreover, it entails the ability to introduce any type of parameters (geometrical, material, etc.) as extra-coordinates into the model, to obtain by solving only once the resulting multidimensional model, the whole envelope containing all possible solutions, a sort computational vademecum that can be then exploited on-line even on light, deployed, computing platforms like smartphones of tablets [16] [4] [11] [12] [9].



Fig. 1 Process sketch

However, in those simulations the laying velocity was considered constant. Thus, transient regimes were not taken into account, and these regimes are of special interest for controlling processes that usually involve repeated accelerations and decelerations. When accelerating, the heating power should increase to ensure melting and molecular diffusion, and when decelerating the heating power must decrease in order to prevent thermal degradation. Because the process control must operate in real time, parametric solutions should be computed offline in order to be used online for process control or process optimization purposes.

In what follows we revisit in section 2 the PGD discretization technique and in section 3 its application for computing parametric solutions involving material parameters, initial and boundary conditions and parameters defining the domain in which the problem is defined. Modeling of the AFP manufacturing process is addressed in section 4, with special emphasis in the consideration of the number of plies as model parameter. Section 5 focusses on transient regimes and the use of the resulting parametric solutions for process control purposes. Finally section 6 addresses few conclusions and perspectives.

## 2 PGD at a glance

Consider a problem defined in a space of dimension d for the unknown field  $u(x_1, \dots, x_d)$ . Here, the coordinates  $x_i$  denote any usual coordinate (scalar or vectorial) related to physical space, time, or conformation space in microscopic descriptions [1], for example, but they could also include, as we illustrate later, model parameters such as boundary conditions or material parameters.

We seek a solution for  $u(x_1, \dots, x_d) \in \Omega_1 \times \dots \times \Omega_d$ . The PGD yields an approximate solution in the separated form:

$$u(x_1, \cdots, x_d) \approx \sum_{i=1}^N X_i^1(x_1) \cdot \ldots \cdot X_i^d(x_d) = \sum_{i=1}^N \prod_{j=1}^d X_i^j(x_j)$$
(1)

The PGD approximation is thus a sum of N functional products involving each a number d of functions  $X_i^j(x_j)$  that are unknown *a priori*. It is constructed by successive enrichment, whereby each functional product is determined in sequence. At a particular enrichment step n + 1, the functions  $X_i^j(x_j)$  are known for  $i \leq n$  from the previous steps, and one must compute the new product involving the d unknown functions  $X_{n+1}^j(x_j)$ . This is achieved by invoking the weak form of the problem under consideration. The resulting problem is non-linear, which implies that iterations are needed at each enrichment step. A low-dimensional problem can thus be defined in  $\Omega_j$  for each of the d functions  $X_{n+1}^j(x_j)$ .

If  $\mathcal{M}$  nodes are used to discretize each coordinate, the total number of PGD unknowns is  $N \cdot \mathcal{M} \cdot d$  instead of the  $\mathcal{M}^d$  degrees of freedom involved in standard mesh-based discretizations.

In the case of a field depending on the physical space  $\mathbf{x} \in \Omega_x \subset \mathbb{R}^3$ , the time  $t \in \mathcal{I}_t \subset \mathbb{R}_+$  and Q parameters  $p^1, \ldots, p^Q, p^j \in \Omega_{p^j}, j = 1, \ldots, Q$ , the solution is sought under the separated form

$$u(\mathbf{x}, t, p^1, \cdots, p^Q) \approx \sum_{i=1}^N X_i(\mathbf{x}) \cdot T_i(t) \cdot \prod_{j=1}^Q P_i^j(p^j)$$
(2)

As soon as this solution is available, after solving the multidimensional model within the PGD framework, we can have access to any possible solution.

PGD solution procedures have been extensively described in our former works and successfully applied in a plethora of applications. The interested reader can refer to the reviews [5] [6] [7] as well as to the primer [10] that describes the practical issues related to its computational implementation. For this reason in section 4 we will focus in some novel aspects that ATP processes involve. Among them we are considering two main aspects: (i) the consideration of the number of plies as a model parameter, allowing the solution of the thermal model for any number of plies; and (ii) the consideration of the heating cycle in a parametric way, allowing for a transient parametric solution able to be applied for control purposes.

## 3 Parameters becoming coordinates

In this section we summarize the developments described in [9] in order to illustrate how parameters of different nature become coordinates. In particular we consider three type of parameters: (i) parameters related to the model; (ii) parameters related to initial and boundary conditions; and (iii) geometrical parameters defining the space-time domain in which the model is defined.

#### 3.1 Model parameters as extra-coordinates

We consider the following parametric heat transfer equation:

$$\frac{\partial u}{\partial t} - k \cdot \Delta u - f = 0 \tag{3}$$

with homogeneous initial and boundary conditions. Here  $(\mathbf{x}, t, k) \in \Omega \times \mathcal{I}_t \times \mathcal{I}_k$ , with  $\Omega \subset \mathbb{R}^3$ ,  $\mathcal{I}_t \subset \mathbb{R}_+$  and  $\mathcal{I}_k \subset \mathbb{R}$ . The scalar conductivity k is here viewed as a new coordinate defined in the interval  $\mathcal{I}_k$ . Thus, instead of solving the thermal model for different discrete values of the conductivity parameter, we wish to solve only once a more general problem. For that purpose we consider the weighted residual form related to Eq. (3):

$$\int_{\Omega \times \mathcal{I}_t \times \mathcal{I}_k} u^* \cdot \left(\frac{\partial u}{\partial t} - k \cdot \Delta u - f\right) d\mathbf{x} \cdot dt \cdot dk = 0$$
<sup>(4)</sup>

The PGD solution is sought in the form:

...

$$u(\mathbf{x}, t, k) \approx \sum_{i=1}^{N} X_i(\mathbf{x}) \cdot T_i(t) \cdot K_i(k)$$
(5)

At iteration n < N the solution  $u^n(\mathbf{x}, t, k)$  reads

$$u^{n}(\mathbf{x},t,k) = \sum_{i=1}^{n} X_{i}(\mathbf{x}) \cdot T_{i}(t) \cdot K_{i}(k)$$
(6)

and the new trial function  $u^{n+1}(\mathbf{x}, t, k)$  is searched according to

$$u^{n+1}(\mathbf{x}, t, k) = \sum_{i=1}^{n+1} X_i(\mathbf{x}) \cdot T_i(t) \cdot K_i(k) =$$
$$u^n(\mathbf{x}, t, k) + X_{n+1}(\mathbf{x}) \cdot T_{n+1}(t) \cdot K_{n+1}(k)$$
(7)

with the test function  $u^*$  given by

$$u^{\star}(\mathbf{x}, t, k) = X^{\star}(\mathbf{x}) \cdot T_{n+1}(t) \cdot K_{n+1}(k) + X_{n+1}(\mathbf{x}) \cdot T^{\star}(t) \cdot K_{n+1}(k) + X_{n+1}(\mathbf{x}) \cdot T_{n+1}(t) \cdot K^{\star}(k)$$
(8)

By introducing the trial and test functions, Eqs. (7) and (8) respectively, into the weak form, Eq. (4), and using an appropriate linearization, functions  $X_{n+1}(\mathbf{x})$ ,  $T_{n+1}(t)$  and  $K_{n+1}(k)$  are calculated. When considering the simplest linearization strategy, the alternated direction fixed point algorithm, the following steps are repeated until reaching convergence:

- 1. With  $T_{n+1}^{(r-1)}(t)$  and  $K_{n+1}^{(r-1)}$  given at the previous iteration of the non linear solver (r-1) (arbitrarily initialized at the first iteration:  $T_{n+1}^{(0)}(t)$  and  $K_{n+1}^{(0)}(k)$ ), all the integrals in  $\mathcal{I}_t \times \mathcal{I}_k$  are performed, leading to a boundary value problem involving  $X_{n+1}^{(r)}(\mathbf{x})$ .
- 2. With  $X_{n+1}^{(r)}(\mathbf{x})$  just calculated and  $K_{n+1}^{(r-1)}$  given at the previous iteration of the non linear solver (r-1), all the integrals in  $\Omega \times \mathcal{I}_k$  are performed, leading to an one-dimensional initial value problem involving  $T_{n+1}^{(r)}(t)$ .
- 3. With  $X_{n+1}^{(r)}(\mathbf{x})$  and  $T_{n+1}^{(r)}$  just updated, all the integrals in  $\Omega \times \mathcal{I}_t$  are performed, leading to an algebraic problem involving  $K_{n+1}^{(r)}(k)$ .

4. The convergence is checked by calculating

$$\mathcal{E}^{r} = \|X_{n+1}^{(r-1)}(\mathbf{x}) - X_{n+1}^{(r)}(\mathbf{x})\| + \|T_{n+1}^{(r-1)}(t) - T_{n+1}^{(r)}(t)\| + \|K_{n+1}^{(r-1)}(k) - K_{n+1}^{(r)}(k)\|$$
(9)

When  $\mathcal{E}^r$  becomes small enough the just computed functions are incorporated into the approximation of the solution:

$$\begin{cases} X_{n+1}(\mathbf{x}) = X_{n+1}^{(r)}(\mathbf{x}) \\ T_{n+1}(t) = T_{n+1}^{(r)}(t) \\ K_{n+1}(k) = K_{n+1}^{(r)}(k) \end{cases}$$
(10)

The convergence of the enrichment iteration is checked as soon as the nonlinear iteration converges, by evaluating the norm of the just computed term  $||X_{n+1}(\mathbf{x}) \cdot T_{n+1}(t) \cdot K_{n+1}(k)||$ ; the residual norm, or any appropriate error estimator based on quantities of interest [3].

#### 3.2 Boundary and initial conditions as extra-coordinates

For the sake of simplicity we first consider the steady state heat equation

$$\nabla \cdot (\mathbf{K} \cdot \nabla u(\mathbf{x})) + f(\mathbf{x}) = 0 \tag{11}$$

with  $\mathbf{x} \in \Omega \subset \mathbb{R}^3$ , subjected to the boundary conditions:

$$\begin{cases} u(\mathbf{x} \in \Gamma_d) = u_g \\ (-\mathbf{K} \cdot \nabla u) |_{\mathbf{x} \in \Gamma_n} \cdot \mathbf{n} = \mathbf{q}_g \cdot \mathbf{n} = q_g \end{cases}$$
(12)

with **K** the conductivity tensor and **n** the outwards unit vector defined in the domain boundary  $\Gamma_n$ , with  $\partial \Omega \equiv \Gamma = \Gamma_d \cup \Gamma_n$  and  $\Gamma_d \cap \Gamma_n = \emptyset$ .

In what follows we address the simplest scenarios consisting in constant Neumann, Dirichlet and initial boundary conditions. More complex and general situations were addressed in [9].

## 3.2.1 Neumann boundary condition as extra-coordinate

First, imagine that we are interested in knowing the model solution for values of the heat flux  $q_g \in \mathcal{I}_q = [q_g^-, q_g^+]$ . We could consider the given heat flux as an extra-coordinate and then solving only once the resulting 4D heat equation for calculating the general parametric solution  $u(\mathbf{x}, q)$ . For this purpose the solution is sought in the separated form

$$u(\mathbf{x}, q_g) \approx \sum_{i=1}^{N} X_i(\mathbf{x}) \cdot \mathcal{Q}_i(q_g)$$
(13)

In order to enforce the prescribed Dirichlet boundary condition  $u(\mathbf{x} \in \Gamma_d) = u_g$  the simplest procedure consists of choosing the first functional couple  $X_1(\mathbf{x}) \cdot Q_1(q_g)$  in order to ensure that  $u^1(\mathbf{x} \in \Gamma_d, q_g) = X_1(\mathbf{x} \in \Gamma_d) \cdot Q_1(q_g) = u_g$ . Thus, the remaining terms of the finite sum  $X_i(\mathbf{x}), i > 1$ , will be subjected to homogeneous essential boundary conditions, i.e.  $X_i(\mathbf{x} \in \Gamma_d) = 0, i > 1$ .

In order to use the approximation (13) we start by considering the weak form related to Eq. (11), that writes: Find  $u(\mathbf{x}) \in H^1(\Omega)$ , verifying  $u(\mathbf{x} \in \Gamma_d) = u_g$ , such that

$$\int_{\Omega} \nabla u^* \cdot (\mathbf{K} \cdot \nabla u) \, d\mathbf{x} = \int_{\Gamma_n} u^* \cdot (\mathbf{K} \cdot \nabla u) \cdot \mathbf{n} \, d\mathbf{x} + \int_{\Omega} u^* \cdot f(\mathbf{x}) \, d\mathbf{x} \qquad (14)$$

is verified  $\forall u^* \in H^1(\Omega)$ , with  $u^*(\mathbf{x} \in \Gamma_d) = 0$ .

By introducing the Neumann condition (12) into (14) it results

$$\int_{\Omega} \nabla u^* \cdot (\mathbf{K} \cdot \nabla u) \, d\mathbf{x} = -\int_{\Gamma_n} u^* \cdot q_g \, d\mathbf{x} + \int_{\Omega} u^* \cdot f(\mathbf{x}) \, d\mathbf{x} \tag{15}$$

For using the approximation (13) we must consider the extended-weak form defined in the domain  $\Omega \times \mathcal{I}_q$ 

$$\int_{\Omega \times \mathcal{I}_q} \nabla u^* \cdot (\mathbf{K} \cdot \nabla u) \ d\mathbf{x} \cdot dq_g = -\int_{\Gamma_n \times \mathcal{I}_q} u^* \cdot q_g \ d\mathbf{x} \cdot dq_g +$$

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$$\int_{\Omega \times \mathcal{I}_q} u^* \cdot f(\mathbf{x}) \, d\mathbf{x} \cdot dq_g \tag{16}$$

By assuming at iteration n:

$$\begin{cases}
 u^{n}(\mathbf{x}, q_{g}) = \sum_{i=1}^{n-1} X_{i}(\mathbf{x}) \cdot \mathcal{Q}_{i}(q_{g}) + X_{n}(\mathbf{x}) \cdot \mathcal{Q}_{n}(q_{g}) = \\
 = u^{n-1}(\mathbf{x}, q_{g}) + X_{n}(\mathbf{x}) \cdot \mathcal{Q}_{n}(q_{g}) \\
 u^{*} = X^{*}(\mathbf{x}) \cdot \mathcal{Q}_{n}(q_{g}) + X_{n}(\mathbf{x}) \cdot \mathcal{Q}^{*}(q_{g})
\end{cases}$$
(17)

Now the double iteration described in the previous section, one for enriching the separated representation and the second one for solving the nonlinear problem arising at each enrichment iteration, is performed in order to calculate the solution separated representation.

#### 3.2.2 Dirichlet boundary condition as extra-coordinate

In this section we consider that we are interested in considering the solution of model (11) for any value of  $u_g$  in (12) in a certain interval  $\mathcal{I}_u = [u_g^-, u_g^+]$ . For this purpose we consider the function  $\varphi(\mathbf{x})$  continuous in  $\overline{\Omega}$  such that  $\Delta \varphi \in L_2(\Omega)$  and  $\varphi(\mathbf{x} \in \Gamma_d) = 1$ . Thus, we can define the change of variable [13]

$$u(\mathbf{x}) = v(\mathbf{x}) + u_g \cdot \varphi(\mathbf{x}) \tag{18}$$

that allows rewriting Eqs. (11) and (12) as:

$$\nabla \cdot (\mathbf{K} \cdot \nabla v(\mathbf{x})) + u_g \cdot \nabla \cdot (\mathbf{K} \cdot \nabla \varphi(\mathbf{x})) + f(\mathbf{x}) = 0$$
(19)

subjected to the boundary conditions:

$$\begin{cases} v(\mathbf{x} \in \Gamma_d) = 0\\ (-\mathbf{K} \cdot \nabla v) |_{\mathbf{x} \in \Gamma_n} \cdot \mathbf{n} = u_g \cdot (\mathbf{K} \cdot \nabla \varphi) |_{\mathbf{x} \in \Gamma_n} \cdot \mathbf{n} + q_g \end{cases}$$
(20)

that results in the weak form

$$\int_{\Omega} \nabla v^* \cdot (\mathbf{K} \cdot \nabla v) \, d\mathbf{x} = -\int_{\Omega} \nabla v^* \cdot u_g \cdot (\mathbf{K} \cdot \nabla \varphi) \, d\mathbf{x} + \int_{\Omega} v^* \cdot f(\mathbf{x}) \, d\mathbf{x} - \int_{\Gamma_n} v^* \cdot q_g \, d\mathbf{x} - \int_{\Gamma_n} v^* \cdot u_g \cdot (\mathbf{K} \cdot \nabla \varphi) \cdot \mathbf{n} \, d\mathbf{x}$$
(21)

We can now introduce  $u_g$  as extra-coordinate, searching the solution in the separated form:

$$v(\mathbf{x}, u_g) \approx \sum_{i=1}^{N} X_i(\mathbf{x}) \cdot \mathcal{U}_i(u_g)$$
(22)

that needs for the extended weak-form

$$\int_{\Omega \times \mathcal{I}_u} \nabla v^* \cdot (\mathbf{K} \cdot \nabla v) \ d\mathbf{x} \cdot du_g =$$

$$= -\int_{\Omega \times \mathcal{I}_{u}} \nabla v^{*} \cdot u_{g} \cdot (\mathbf{K} \cdot \nabla \varphi) \, d\mathbf{x} \cdot du_{g} + \int_{\Omega \times \mathcal{I}_{u}} v^{*} \cdot f(\mathbf{x}) \, d\mathbf{x} \cdot du_{g} - \int_{\Gamma_{n} \times \mathcal{I}_{u}} v^{*} \cdot q_{g} \, d\mathbf{x} \cdot du_{g} - \int_{\Gamma_{n} \times \mathcal{I}_{u}} v^{*} \cdot u_{g} \cdot (\mathbf{K} \cdot \nabla \varphi) \cdot \mathbf{n} \, d\mathbf{x} \cdot du_{g}$$
(23)

on which the alternated directions fixed point algorithm applies again to calculate the parametric solution (22).

#### 3.3 Initial conditions as extra-coordinates

We consider in this section the transient heat equation in a homogeneous and isotropic medium

$$\rho C_p \frac{\partial u}{\partial t} = k \Delta u + f \tag{24}$$

 $t \in \mathcal{I}_t = (0, \Theta] \subset \mathbb{R}_+, \mathbf{x} \in \Omega \subset \mathbb{R}^3$  and f = cte. The initial and boundary conditions read:

$$\begin{cases} u(\mathbf{x} \in \Gamma_d) = u_g \\ (-k\nabla u) \mid_{\mathbf{x} \in \Gamma_n} \cdot \mathbf{n} = q_g \\ u(\mathbf{x}, t = 0) = u^0(\mathbf{x}) \end{cases}$$
(25)

The associated weak form reads:

$$\int_{\Omega} u^* \rho C_p \frac{\partial u}{\partial t} \, d\mathbf{x} + \int_{\Omega} k \nabla u^* \cdot \nabla u \, d\mathbf{x} = -\int_{\Gamma_n} u^* \cdot q_g \, d\mathbf{x} + \int_{\Omega} u^* \cdot f(\mathbf{x}) \, d\mathbf{x} (26)$$

that includes explicitly the natural (Neumann) boundary conditions. To prescribe both the initial and the essential (Dirichlet) boundary conditions we proceed to define the following functions:

$$\hat{u}^{0}(\mathbf{x}) = \begin{cases} u^{0}(\mathbf{x}), \, \mathbf{x} \in \Omega\\ 0, \, \mathbf{x} \in \Gamma \end{cases}$$
(27)

$$\Upsilon(t) = \begin{cases} 1, t > 0\\ 0, t = 0 \end{cases}$$
(28)

and  $\varphi(\mathbf{x})$  continuous in  $\overline{\Omega}$ , verifying  $\Delta \varphi \in L_2(\Omega)$  and the essential boundary conditions

$$\varphi(\mathbf{x} \in \Gamma_d) = u_g \tag{29}$$

We could define the function  $\Sigma(\mathbf{x}, t)$  expressed in the separated form

$$\Sigma(\mathbf{x},t) = \hat{u}^0(\mathbf{x}) + \varphi(\mathbf{x}) \cdot \Upsilon(t)$$
(30)

that verifies the initial and essential boundary conditions. However, functions  $\hat{u}^0$  and  $\Upsilon(t)$  are not regular enough to be employed in the weak form of the problem. A direct regularization consists in defining these functions at the nodal positions and then define interpolations with the required regularity.

By applying now the change of variable:

$$u(\mathbf{x},t) = v(\mathbf{x},t) + \Sigma(\mathbf{x},t) = v(\mathbf{x},t) + \hat{u}^{0}(\mathbf{x}) + \varphi(\mathbf{x}) \cdot \Upsilon(t)$$
(31)

and approximating the initial condition as

$$\hat{u}^{0}(\mathbf{x}) \approx \sum_{k=1}^{S_{0}} U_{0}^{k} \cdot \eta_{k}(\mathbf{x})$$
(32)

the parametric solution is assumed having the form

$$v(\mathbf{x}, U_0^1, \cdots, U_0^{S_0}) \approx \sum_{i=1}^N X_i(\mathbf{x}) \cdot \prod_{j=1}^{S_0} \mathcal{U}_i^j(U_0^j)$$
 (33)

with  $U_0^j \in \mathcal{I}_0^j = [(U_0^j)^-, (U_0^j)^+].$ 

# 3.4 Parametric domains

For the sake of clarity and without loss of generality we are addressing in this section the transient one-dimensional heat equation

$$\frac{\partial u}{\partial t} = \alpha \frac{\partial^2 u}{\partial x^2} + f \tag{34}$$

with  $t \in \mathcal{I}_t = (0, \Theta] \subset \mathbb{R}$ ,  $x \in \Omega = (0, L) \subset \mathbb{R}$ , f = cte and u(x = 0, t) = u(x = L, t) = u(x, t = 0) = 0.

The associated space-time weak form reads:

$$\int_{\Omega \times \mathcal{I}_t} u^* \cdot \frac{\partial u}{\partial t} \, dx \, dt = -\alpha \int_{\Omega \times \mathcal{I}_t} \frac{\partial u^*}{\partial x} \cdot \frac{\partial u}{\partial x} \, dx \, dt + \int_{\Omega \times \mathcal{I}_t} u^* \cdot f \, dx \, dt$$
(35)

If we are interested in computing the solution u(x,t) in many domains of length  $L \in [L^-, L^+]$  and for many time intervals of length  $\Theta = [\Theta^-, \Theta^+]$ , more than solving the model for many possible choice in order to define a meta-model, it is preferable to compute the parametric solution  $u(x, t; L, \Theta)$ .

However, Eq. (35) does not involve an explicit dependence on the extracoordinates L and  $\Theta$ , both defining the domain of integration. In order to explicit this dependence, we consider the coordinate transformation

$$\begin{cases} t = \tau \cdot \Theta, \ \tau \in [0, 1] \\ x = \lambda \cdot L, \ \lambda \in [0, 1] \end{cases}$$
(36)

In this case the weak form (35) reads:

$$\int_{[0,1]^2} u^* \frac{\partial u}{\partial \tau} L \, d\lambda \, d\tau = -\alpha \int_{[0,1]^2} \frac{\partial u^*}{\partial \lambda} \frac{\partial u}{\partial \lambda} \frac{\partial}{\partial \lambda} \frac{\partial}{L} \, d\lambda \, d\tau + \int_{[0,1]^2} u^* f L\Theta \, d\lambda \, d\tau (37)$$

that allows calculating the parametric solution derived from (26) after applying the change of coordinates

$$u(\lambda,\tau,L,\Theta) \approx \sum_{i=1}^{N} \tilde{X}_{i}(\lambda) \cdot \tilde{T}_{i}(\tau) \cdot \mathcal{L}_{i}(L) \cdot \mathcal{T}_{i}(\Theta)$$
(38)

# 4 Process modeling

In the AFP process many parametric solutions are of interest. In [8] authors focused on the solution of the steady state thermal problem where the thermal contact resistances, the laser power and the line velocity were considered as parameters and then included into the PGD parametric solution as extracoordinates.

In the present paper we are addressing two major issues: (i) the consideration of the number of plies composing the laminate as a model parameter; and (ii) the consideration of transient solutions induced by non constant laying velocities, both of major interest for controlling the process.

## 4.1 Number of plies as parameter

First we consider the steady state regime in the laser frame. The material domain results  $\Omega = [0, L] \times [0, W] \times [0, H]$  where the heat equation is solved

$$\rho C_p \mathbf{V} \cdot \nabla u(\mathbf{x}) = \nabla (\mathbf{K} \cdot \nabla u) \tag{39}$$

where **V** is the line velocity assumed constant ( $\mathbf{V}^T = (V, 0, 0)$ ) and **K** the conductivity tensor. The domain thickness consists of p plies with equal thickness  $e_p$ , such that  $H = p \cdot e_p$ .

The boundary conditions are given by (see Fig. 2):

$$\begin{cases} u(\mathbf{x} \in \Gamma_D) = u_g \\ -(\mathbf{K} \cdot \nabla u) \cdot \mathbf{n}|_{\mathbf{x} \in \Gamma_N} = q_g(\mathbf{x}) = 0 \\ -(\mathbf{K} \cdot \nabla u) \cdot \mathbf{n}|_{\mathbf{x} \in \Gamma_L} = q_g(\mathbf{x}) = \Phi(\mathbf{x}) \\ -(\mathbf{K} \cdot \nabla u) \cdot \mathbf{n}|_{\mathbf{x} \in \Gamma_R} = h(\mathbf{x})(u(\mathbf{x}) - u_{ext}) \end{cases}$$
(40)

where  $\Gamma_D = (L, y, z)$ ,  $\Gamma_N = \{(0, y, z) \cup (x, 0, z) \cup (x, W, z)\}$ ,  $\Gamma_L$  represents the zone in which the laser applies ( $\Gamma_L = (x \in (x_r - L_{\varPhi}, x_r + L_{\varPhi}), y, z = H)$ ), and  $\varPhi(\mathbf{x})$  the laser heat flux. In the remaining part of the boundary convective heat exchanges occur, governed by the hydrodynamic coefficient  $h(\mathbf{x})$  that takes different values on  $\Gamma_M = (x, y, 0)$  where the part is in contact with the mould and on the upper surface in contact with the air  $\Gamma_a = (x \in \{(0, x_r - L_{\varPhi}, L)\}, y, H)$ .

Moreover, due to an imperfect consolidation at the ply interfaces  $\Gamma_i = \{(x, y, e_p) \cup (x, y, 2e_p) \cup \cdots \cup (x, y, (p-1)e_p)\}$ , with p > 1, a thermal resistance must be considered

$$-(\mathbf{K}\cdot\nabla u)\cdot\mathbf{k}|_{x,y,(l\cdot e_p)^-} = -(\mathbf{K}\cdot\nabla u)\cdot\mathbf{k}|_{x,y,(l\cdot e_p)^+}$$
(41)

with

$$-(\mathbf{K} \cdot \nabla u) \cdot \mathbf{k}|_{x,y,(l \cdot e_p)^-} = h_i(u(x,y,(l \cdot e_p)^-) - u(x,y,(l \cdot e_p)^+))$$
(42)

 $l = 1, \dots, p - 1$ , with p the number of plies and  $h_i$  the interface thermal resistance.



Fig. 2 Problem domain and notation

## 4.1.1 Domain transformation

When the number of plies becomes a parameter, the domain thickness H depends on the considered number of plies as depicted in Fig. 3. In order to define the problem in a reference domain we consider the coordinate transformation

$$z = \lambda \cdot p \cdot e_p \tag{43}$$

with  $\lambda \in [0, 1]$ .

The derivatives are transformed according to

$$\frac{\partial}{\partial z} = \frac{\partial}{\partial \lambda} \frac{d\lambda}{dz} = \frac{\partial}{\partial \lambda} \frac{1}{e_p p} \tag{44}$$

Now the most natural choice for the parametric solution within the PGD framework consists of  $u(x, y, \lambda, p)$  where for taking into account the discrete nature of the extra-coordinate p it suffices considering  $p \in \mathcal{I}_p = [1, 2, \dots, P_M]$ ,  $P_M \in \mathbb{N}$ . The fact of having a discrete nature is not an issue because the model does not imply derivatives with respect to the coordinate p.

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Fig. 3 Evolution of the domain with the number of plies

## 4.1.2 Interface treatment

The fact of considering the number of plies as extra-coordinates does not represent serious difficulties if the interfaces were perfectly consolidated ensuring the temperature continuity across all them. In that case it suffices considering the separated representation

$$u(x, y, \lambda, p) \approx \sum_{i=1}^{N} X_i(x) \cdot Y_i(y) \cdot \mathcal{L}_i(\lambda) \cdot \mathcal{P}_i(p)$$
(45)

that injected into the weak form of Eq. (39) and proceeding as illustrated in section 3 allows calculating the parametric thermal field for any number of plies  $p \in \mathcal{I}_p$ .

However, when interfacial thermal resistances must be considered an important issue appears suddenly. First we must take into account the temperature discontinuity across the plies interfaces. The simplest possibility consists in duplicating the nodes at those interfaces. However the interfaces positions depends on the number of plies considered. For example when considering two plies (p = 2) the interface is located at  $\lambda = 1/2$ . When considering three plies (p = 3) the two interfaces are located at  $\lambda = 1/3$  and  $\lambda = 2/3$ .

The most direct solution consists in duplicating all the nodes located at any possible interface. Thus for example if  $P_M = 3$ , we must take into account the interface located at  $\lambda = 1/2$  for p = 2 and the two interfaces  $\lambda = 1/3$  and

 $\lambda = 2/3$  associated with p = 3. Thus, finally the simplest solution consists in duplicating nodes located at  $\lambda = 1/3$ ,  $\lambda = 1/2$  and  $\lambda = 2/3$  in order to represent any interface. However we must pay special attention when considering the interface transmission conditions, because for example for p = 3 we should enforce temperature discontinuity at  $\lambda = 1/3$  and  $\lambda = 2/3$  but perfect continuity at  $\lambda = 1/2$  because when operating with p = 3,  $\lambda = 1/2$  is not a real interface and then temperature continuity must be enforced. This situation is illustrated in Fig. 4.



Fig. 4 Interface treatment

In order to keep a unified description of interface conditions, needing for each p the consideration of both continuity and discontinuity transmission conditions, we propose enforcing continuity by applying the Nitsche's method. Imagine for a while that we are solving:

$$-\Delta u = f \tag{46}$$

with  $u = u_0$  in  $\Gamma = \partial \Omega$ . The Nitsche's method consists of considering the following symmetric weak form

$$\int_{\Omega} \nabla u^* \cdot \nabla u \, d\mathbf{x} - \int_{\Gamma} u^* \, \nabla u \cdot \mathbf{n} \, d\mathbf{x} - \int_{\Gamma} u \nabla u^* \cdot \mathbf{n} \, d\mathbf{x} + \beta \int_{\Gamma} u \, u^* \, d\mathbf{x} =$$
$$\int_{\Omega} u^* f \, d\mathbf{x} - \int_{\Gamma} u_0 \, \nabla u^* \cdot \mathbf{n} \, d\mathbf{x} + \beta \int_{\Gamma} u_0 \, u^* \, d\mathbf{x}$$
(47)

with  $u \in H^1(\Omega)$  and  $\forall u^* \in H^1(\Omega)$ .

If now we come back to the enforcement of temperature continuity across the interface located at  $\lambda_k$  for a certain p for which  $\lambda_k$  is not a real interface (it will be for another p), the transmission condition writes from one side of the interface where the temperature is denoted by  $u^-$  assuming that  $u^- = u^+$ , and on the other side by enforcing  $u^+ = u^-$ , both written by using the Nitsche's formuation (47).

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Fig. 5 Particularization of  $u(x, y, \lambda, p)$  to: (a) p = 1; (b) p = 2; (c) p = 3 and (d) p = 4.

On the other hand real interfaces ere easy treated because the weak forms from both sides of the interface involve the interface integrals:

$$\begin{cases} -\int_{\Gamma_i} u^* \left( \mathbf{K} \cdot \nabla u^- \right) \cdot \mathbf{n} \, d\mathbf{x} = h_i (u^- - u^+) \\ -\int_{\Gamma_i} u^* \left( \mathbf{K} \cdot \nabla u^+ \right) \cdot \mathbf{n} \, d\mathbf{x} = h_i (u^+ - u^-) \end{cases}$$
(48)

#### 4.1.3 Numerical results

In order to check the solution procedure we consider L = 1 (IS units), W = 0.1,  $e_p = 0.000135$ ,  $P_M = 6$  and the thermal source applying at  $x_r = 0.6$ , at the tape-substrate interface  $z = (p - 1)e_p$ , and having a distribution given by

$$\Phi(x,t) = q(t) \ e^{-\alpha \cdot (x-\mu)^2}.$$
(49)

The parametric solution  $u(x, y, \lambda, p)$ , with  $(x, y, \lambda) \in [0, L] \times [0, W] \times [0, 1]$ and  $p = \{1, 2, \dots, 5\}$ , was calculated and then particularized for the different number of plies as illustrated in Fig. 5.

## 5 From steady state to transient parametric solutions

5.1 Steady-state parametric solution

In order to control de process other parameters should be introduced as extracoordinates, in particular the laser power q and the line velocity V. Thus the parametric solution involves the space coordinates, the number of plies, the laser power and the line velocity, i.e.  $u(x, y, \lambda, p, q, V)$ .



Fig. 6 Highest temperature as a function of the maser power and the line velocity when considering 8 plies:  $u_8(q, V)$ .

Within the PGD framework the separated representation reads:

$$u(x, y, \lambda, p, q, V) \approx \sum_{i=1}^{N} X_i(x) \cdot Y_i(y) \cdot \mathcal{L}_i(L) \cdot \mathcal{P}_i(p) \cdot \mathcal{Q}_i(q) \cdot \mathcal{V}_i(V)$$
(50)

Denoting by  $u_n^h$  the rank-*n* solution, the approximation  $u_{n+1}^h$  results from:

$$(X_{n+1}(x), Y_i n + 1(y), \mathcal{L}_{n+1}(\lambda), \mathcal{P}_{n+1}(p), \mathcal{Q}_{n+1}(q), \mathcal{V}_{n+1}(V)) = \operatorname{argmin}_{(X,Y,\mathcal{L},\mathcal{P},\mathcal{Q},\mathcal{V})} ||\mathbb{L}^h \left( u_n^h + X(x) \cdot Y(y) \cdot \mathcal{L}(\lambda) \cdot \mathcal{P}(p) \cdot \mathcal{Q}(q) \cdot \mathcal{V}(V) \right) - f||(51)$$

where  $\mathbb{L}^h$  is the discretized operator. The enrichment is computed using a fixed point algorithm, where the minimization of the above expression is carried out in turn on each of the 7 functions as previously described. The interested reader can refer to the primer [10] for the details on the separated representation constructor.

## 5.1.1 Heating law determination from the steady-sate parametric solution

If one is interested by ensuring there target temperature  $T_{opt}$  at the hottest point  $(x_r, W/2, (p-1)e_p)$  it suffices to extract from  $u(x, y, z, \lambda, p, q, V)$  the following parametric solutions:

$$u_p\left(q, V; x = x_r, y = W/2, p, \lambda = \frac{p-1}{p}\right), \quad p = 2, \cdots, P_M$$
 (52)

Fig. 6 depicts  $u_8(q, V)$ , that is the highest temperature as a function of the heating power and the line velocity when considering 8 plies, i.e. p = 8.

Now, as soon as the target temperature  $T_{opt}$  is selected, it suffices, for each number of plies, extracting from  $u_p(q, V)$  the curve  $u_p(p, V) = T_{opt}$ . Figure 8 depicts such a curve for three different values of the target temperature  $T_{opt}$ 



**Fig. 7** Process window when considering 8 plies:  $u_8(q, V) = T_{opt}$ .

when considering 8 plies, associated with the parametric solution depicted in Fig. 7  $\,$ 

Now, we could try to use such parametric solutions for controlling transient regimes. We consider the velocity-time profile shown in Fig. 8 (blue curve) and we decided to use the parametric solution  $u_8(q, V)$  for adjusting the laser power (red curve) in order to ensure a constant target temperature  $T_{opt} = 400$ . Fig. 8 proves that despite the fact of using a parametric solution computed under the stationary constraint, the control seems quite good because when considering both inputs (the laser power and the line velocity) the solution of the thermal problem computed by using a stander discretization technique (green curve) remains very close to the target temperature.

#### 5.2 Transient parametric solution

However, a more accurate control requires the solution of the transient model and then the calculation of a parametric transient solution. In this case the time derivative of the temperature is retained in the thermal model (39) that now reads

$$\rho C_p \left( \frac{\partial u}{\partial t} + \mathbf{V} \cdot \nabla u(\mathbf{x}) \right) = \nabla (\mathbf{K} \cdot \nabla u)$$
(53)

and consequently the parametric solution needs considering the time and also the initial velocity and heating power as well as the accelerations of both the line velocity and the heating source, as illustrated in Fig. 9.



Fig. 8 Heating control from the parametric steady-state solution.



Fig. 9 Modeling the transient heating: process parameters

The thermal process is characterized by an initial interval in which the heating power and the line velocity are constants in order to reach the steady state before enforcing the acceleration. Other options are also possible. In this case the parametric solution writes  $u(x, y, \lambda, p, t, q, V, \gamma, c)$  and within the PGD framework the separated representation writes:

 $u(x,y,\lambda,p,t,q,V,\gamma,c)\approx$ 

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Fig. 10 Choosing the power evolution

$$\sum_{i=1}^{N} X_i(x) \cdot Y_i(y) \cdot \mathcal{L}_i(L) \cdot \mathcal{P}_i(p) \cdot T_i(t) \cdot \mathcal{Q}_i(q) \cdot \mathcal{V}_i(V) \cdot \mathcal{G}_i(\gamma) \cdot \mathcal{C}_i(c)$$
(54)

#### 5.2.1 Heating law determination from the transient parametric solution

Now, as soon as the velocity evolution is defined from  $\gamma$ , we must obtain c in order to minimize the gap with respect to the target temperature. The minimization process is illustrated in Fig. 10.

In the case of the more complex heating scenarios, as the one considered in Fig. 8, one could consider the transient parametric solution instead the steady-state one considered in Fig. 8. Fig. 11 proves that when making it, the procedure allows for a better fitting with respect to the target temperature.

## 5.2.2 Other possible computational vademecums

In order to simulate more complex scenarios involving an acceleration phase, followed by a plateau, to finish with a decelerating regime, we decided to create a parametric solution with as parameters the space coordinates, the time, three characteristic thermal resistances (one representative of the interfaces within the substrate, another representing the ply-substrate interface and the last one the one existing with the environement on the upper boundary) and the process parameters.



Fig. 11 Heating control from the parametric steady-state solution.

The process parameters concern 3 times, the plateau velocity and the plateau heating power. Thus, in the interval  $[t_0 = 0, t_1]$  both velocity and laser increase linearly (with respect to time) to reach at time  $t_1$  both target values: the plateau velocity and heating power. Then the system evolves with constant velocity and power within the interval  $[t_1, t_2]$ . If the length of this interval is large enough the steady state conditions predicted by the model [8] are attained. This check served to validate transient model. Finally, within the interval  $[t_2, t_3]$  both the velocity and the power decrease linearly to vanish at the terminal time  $t_3$ . The parametric solution contains in this case 12 coordinates.

#### 6 Conclusion

This paper proposes an original approach to simulation AFP composites manufacturing processes. First, using a spatial transformation to match a reference domain, the number of plies composing the laminate was considered as model parameter and then as problem extra-coordinate within the PGD framework. Different parametric solutions (computational vademecums) were defined by incorporating model parameters, boundary conditions and geometrical parameters. These parametric solutions were then used in order to define the heating laws in a vary simple and efficient manner.

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