Proper Generalized Decomposition of Parameterized Electrothermal Problems Discretized by the Finite Integration Technique

Alexander Krimm\textsuperscript{1}, Thorben Casper\textsuperscript{1,2}, Sebastian Schöps\textsuperscript{1,2}, Herbert De Gersem\textsuperscript{1,2}, Ludovic Chamoin\textsuperscript{3}

\textsuperscript{1}Institut für Theorie Elektromagnetischer Felder, Technische Universität Darmstadt, Darmstadt, Germany
\textsuperscript{2}Graduate School of Computational Engineering, Technische Universität Darmstadt, Darmstadt, Germany
\textsuperscript{3}Laboratory of Mechanics and Technology, ENS Cachan, CNRS, Université Paris-Saclay, Cachan, France

Abstract—The proper generalized decomposition is applied to a static electrothermal model subject to uncertainties. A reduced model that circumvents the curse of dimensionality is obtained. The quadratic electrothermal coupling term is non-standard and requires the introduction of a trilinear form. An existing finite integration technique based solver is used to demonstrate the opportunities and difficulties in integrating the proper generalized decomposition in existing codes.

Index Terms—Curse of dimensionality, electrothermal, model order reduction, proper generalized decomposition.

I. INTRODUCTION

A recurring problem for simulation tools in engineering is the high number of possible parameter combinations. For optimization and uncertainty quantification, surrogate models are required. Classical techniques construct surrogate models from empirical knowledge with a subsequent fitting to measurement data. However, such simplified models are usually only valid in a limited parameter range.

Proper generalized decomposition (PGD) allows to obtain a reduced model for a wide class of problems \[1, 2\]. In contrast to other techniques, its complexity is not subject to the curse of dimensionality. It separates the high dimensional solution into univariate functions using an iterative algorithm. Whereas mainly used in mechanics, few works of PGD in computational electromagnetics have been reported on, e.g., for extracting parasitic capacitances of interconnects in integrated circuits \[3\] and for magnetoequistatic field-circuit couplings \[4\]. Further examples include the application to skin effect problems \[5\] and for magnetoquasistatic field-circuit couplings \[4\].

In this paper, we exploit PGD for reducing the electrothermal model of a microelectronic chip package, see Fig. 1. Such a package typically contains a high number of bond wires affected by variability in the fabrication process. Any such a package typically contains a high number of bond wires affected by variability in the fabrication process. Evaluating this problem for all parameter combinations exceeds the capability of classical techniques as, e.g., Monte Carlo simulation. The construction of a surrogate model as described in this paper can also be used for hybrid approaches as, e.g., reported on in \[8\].

The paper is structured as follows. In Section II, we introduce the parametric electrothermal problem. The PGD algorithm applied to the electrothermal coupling and its usage with the finite integration technique (FIT) is presented in Section III. For illustration and validation purposes, in Section IV we show numerical results when solving a 1D model problem and a 3D microelectronic chip package using PGD. Finally, Section V concludes the paper.

II. PARAMETERIZED ELECTROTHERMAL PROBLEM

We illustrate applications of PGD in vast, the focus in this paper is the static electrothermal problem. We combine the electrokinetic problem on the spatial domain \( \Omega \) of dimension \( d \) obtained from MAXWELL’s equations \[9\] with the static heat equation \[10\] to obtain

\[-\nabla \cdot (\sigma(x; \mu) \nabla \varphi(x; \mu)) = 0, \quad x \in \Omega, \quad (1a)\]

\[-\nabla \cdot (\lambda(x; \mu) \nabla T(x; \mu)) = Q(\sigma, \varphi), \quad x \in \Omega, \quad (1b)\]

with suitable boundary conditions. Here, \( \varphi(x; \mu) \) is the electric potential, \( T(x; \mu) \) the temperature, \( \sigma(x; \mu) \) the electric conductivity, \( \lambda(x; \mu) \) the thermal conductivity and \( Q(\sigma, \varphi) := \sigma |\nabla \varphi|^2 \) describes the resistive losses. The parameters \( \mu \in \Psi := \times_{p=1}^{n_p} I_p \), express variations due to the fabrication process, where \( n_p \) is the number of parameters and \( I_p \in \mathbb{R} \) the domain of parameter \( p \). These variations can, e.g., arise due to an uncertain length of the bond wires in a microelectronic chip package, see Fig. 1. Assuming temperature independent material properties, we obtain a one-directional coupling by the resistive losses \( Q \).

In the following, we treat problem (1) as a high dimensional problem by introducing the space-parameter variable

\[ \tilde{x} = (x, \mu_1, \ldots, \mu_{n_p})^\top \in \Theta := \Omega \times \Psi, \] (2)

such that (1) simplifies to

\[-\nabla \cdot (\sigma(\tilde{x}) \nabla_x \varphi(\tilde{x})) = 0, \quad \tilde{x} \in \Theta, \quad (3a)\]

\[-\nabla \cdot (\lambda(\tilde{x}) \nabla_x T(\tilde{x})) = Q(\sigma, \varphi), \quad \tilde{x} \in \Theta, \quad (3b)\]

where \( \nabla_x \) is the gradient in \( x \)-direction.
III. PROPER GENERALIZED DECOMPOSITION

To apply the PGD to (3), we recognize that the subproblems (3a) and (3b) exhibit an identical mathematical structure except for the different right-hand sides. First, we focus on the left-hand side using (3a) as an example and secondly, we consider the electrothermal coupling term on the right-hand side of (3b). Since the PGD formulation is typically based on the weak formulation, we introduce test functions \( v \) and use integration by parts. Whereas the formulation here is done for an arbitrary number of parameters and becomes quite technical, we show a simplified 1D example with one parameter in Section IV-A.

A. Electrokinetic Problem

The weak form of (3a) reads: find \( \varphi \in V_\Theta \) s.t.

\[
\langle \sigma(\bar{x}) \nabla \varphi(\bar{x}), \nabla v(\bar{x}) \rangle_{V_\Theta} = 0, \quad \forall v \in V_\Theta, \quad (4)
\]

where \( \langle u, v \rangle_V \) denotes the \( L^2 \) inner product of functions \( u \) and \( v \) in \( V \) and \( V_\Theta := \otimes_{p=1}^{n_p} V_p \), with \( V_1 := H^1(\Omega) \) and \( V_p := L^2(\Omega) \) for \( p = 2, \ldots, n_p \) and \( n_p = n_p + 1 \), representing one spatial and \( n_p \) parameter spaces. Following the PGD approach, we assume that the solution can be approximated by

\[
\varphi(\bar{x}) \approx \varphi_{\text{PGD}}(\bar{x}) := \sum_{s=1}^{m} u_s(\bar{x}) := \sum_{s=1}^{m} \prod_{p=1}^{n_p} u_s^p(\bar{x}_p), \quad (5)
\]

where \( m \) is the number of PGD modes. Note that we do not separate the spatial dependencies to enable the usage of existing 3D solvers. Therefore, we have \( \bar{x}_1 = x \in \mathbb{R}^3 \) and \( \bar{x}_p \in \mathbb{R}^p \) for \( p > 1 \). The idea of the PGD is the separation of variables such that a decomposition of all quantities in (4) is required. The electric conductivity is decomposed as

\[
\sigma(\bar{x}) = \prod_{s=1}^{m} \sigma_\varphi(\bar{x}) := \prod_{s=1}^{m} \prod_{p=1}^{n_p} \sigma_{s,p}(\bar{x}_p), \quad (6)
\]

which is trivial for, e.g., piecewise constant parameterized coefficients on \( n_s \) different subdomains of the problem. For more complex cases, appropriate approximations may be used to reduce the required number of terms \( n_\sigma \). The resulting decomposition of (4) reads,

\[
\sum_{s=1}^{m} \sum_{q=1}^{n_s} \langle \sigma_q(\bar{x}) \nabla \varphi(\bar{x}), \nabla v(\bar{x}) \rangle_{V_\Theta} = 0, \quad (7)
\]

where the abbreviation \( \mathcal{L} \) was introduced for conciseness.

To solve (7), we assume that all modes \( s < m \) are known and solve

\[
\sum_{q=1}^{m} \mathcal{L}(q, m) = -\sum_{q=1}^{m} \sum_{s=1}^{m} \mathcal{L}(q, s) \quad (8)
\]

for mode \( m \). Starting with \( m = 0 \), where the right-hand side is zero, an iterative enrichment scheme is established. Let \( p^* \in \{1, \ldots, n_p\} \) be an arbitrary direction, then this nonlinear equation can be solved with a fixed point iteration solving for the \( p^* \)-th direction in each step. Accordingly, the test function for mode \( m \) is separated as

\[
v^m(\bar{x}) = \sum_{p^*=1}^{n_p} v_{p^*}(\bar{x}_{p^*}) \prod_{p=1}^{n_p} u_{p}^{m}(\bar{x}_p), \quad (9)
\]

where the PGD suggests to choose the already calculated functions \( u_{p}^{m} \) as test functions for the directions \( p \neq p^* \). In the following, we omit the parametric dependencies for a compact notation. All parameter functions \( u_{p}^{m} \) with \( p \neq p^* \) are considered known, giving

\[
\mathcal{L}(q, m) = \langle \sigma_q \nabla_{\bar{x}} u_1^m, \nabla_{\bar{x}} v_1 \rangle_{V_\Theta} \prod_{p=2}^{n_p} \langle \sigma_{q,p} u_{p}^{m}, u_{p}^{m} \rangle_{L^2(\Omega)}, \quad (10)
\]

A further simplification of \( \mathcal{L} \) depends on whether the current step solves for the spatial solution or for a parametric solution. In the following, we outline the spatial case \( p^* = 1 \) and leave the parametric case to the reader. If the current step evaluates the spatial problem, we obtain

\[
\mathcal{L}(q, m) = \langle \sigma_{q,1} \nabla_{\bar{x}} u_1^m, \nabla_{\bar{x}} v_1 \rangle_{H^1(\Omega)} \prod_{p=2}^{n_p} \langle \sigma_{q,p} u_{p}^{m}, u_{p}^{m} \rangle_{L^2(\Omega)}, \quad (11)
\]

which can also be written as

\[
\mathcal{L}(q, m) = a_{q,1}(u_1^m, v_1) \prod_{p=2}^{n_p} a_{q,p}(u_{p}^{m}, u_{p}^{m}), \quad (12)
\]

with the bilinear forms defined as

\[
a_{q,1}(u, v) := \langle \sigma_{q,1} \nabla_{\bar{x}} u, \nabla_{\bar{x}} v \rangle_{H^1(\Omega)}, \quad (13)
\]

\[
a_{q,p}(u, v) := \langle \sigma_{q,p} u_1, v \rangle_{L^2(\Omega)}. \quad (14)
\]

Applying a standard GALERKIN finite element discretization to (8) with \( \mathcal{L} \) as given in (12), the linear system for step \( k \) of the fixed point iteration reads

\[
\sum_{q=1}^{m} M_{q,1} u_1^{m,k} \prod_{p=2}^{n_p} a_{q,m,p} = -\sum_{q=1}^{m} \sum_{s=1}^{m-1} M_{q,1} u_1^{m,k} \prod_{p=2}^{n_p} \alpha_{q,p}^{s,k}, \quad (15)
\]

where we have introduced the scalar constants

\[
\alpha_{i,j,p}^{s,k} := (u_{i,j}^{p,k})^\top M_{q,p} u_{p}^{s,k}, \quad (16)
\]

with the integers \( i \) and \( j \), and the finite element matrices \( M_{q,1} \) and \( M_{q,p} \) stemming from the discretization of the bilinear forms \( a_{q,1}(u, v) \) and \( a_{q,p}(u, v) \), respectively, apart from the corresponding weighting vectors.

In Fig. 2 a Nassi-Sneiderman diagram of the PGD algorithm applied to the electrokinetic subproblem is shown. After each parametric (non-spatial) step of the fixed point iteration, a normalization of the solution is applied. For the stopping criterion, the tolerances tolPGD and tolPGD are used to define the accuracy of the fixed point and the PGD iteration, respectively. Here, the solution change \( \Delta^{(k)} = \prod_{p=1}^{n_p} \| u_{p}^{m,k} - u_{p}^{m,k-1} \| \) and the solution’s magnitude \( u_{m}^{m} := \prod_{p=1}^{n_p} \| u_{p}^{m} \| \) of mode \( m \) is used.
B. Electrothermal Coupling

Once the PGD solution \( \varphi^{PGD} \) of the electric potential is obtained, it can be used to couple to the thermal problem via the resistive losses \( Q \). The weak form of the thermostatic problem reads: find \( T \in V_\Theta \) s.t.

\[
\langle \lambda(\tilde{x}) \nabla_x T(\tilde{x}), \nabla_x v(\tilde{x}) \rangle_{V_\Theta} = \langle Q(\sigma, \varphi), v(\tilde{x}) \rangle_{V_\Theta},
\]

(17)

for all \( v \in V_\Theta \). The left-hand side of (17) is mathematically identical to that of (4). Therefore, we focus on the coupling term \( \mathcal{R} \) on the right-hand side. Again, we omit the parametric dependencies and arrive at a separated form of \( \mathcal{R} \) given by

\[
\mathcal{R} = \sum_{q=1}^{n_q} \sum_{j=1}^{m} \frac{1}{2} \langle \sigma_q \nabla_x u^j_q, \nabla_x u^j_q \rangle_{V_\Theta}
\]

(18)

with the trilinear form

\[
Q^j_{q,p} = \langle \sigma_q \nabla_x u^j_q, \nabla_x u^j_p, v \rangle_{V_\Theta}.
\]

Note that we abused notation by using \( m \) here as the number of modes of the electric problem. To finally solve the thermal subproblem, the term \( \mathcal{R} \) needs to be added to the right-hand side of the thermal equivalent to (8) resulting in a discretized linear system similar to (15). Since we consider only a one-directional coupling, the final electrothermal solution is given by the solutions of the two subproblems.

C. Using existing 3D solvers

Although the PGD is commonly based on the weak formulation, as done in Section III discretization schemes that do not require a weak formulation can be used as well. In each step of the fixed point iteration, either the spatial or a parametric problem needs to be solved. Solving a parametric problem is specific to PGD and is typically implemented by the PGD tool itself. On the other hand, the implementation of a spatial solver is more involved, motivating the usage of existing solvers. However, a completely non-intrusive implementation of the PGD algorithm may not be possible, since the coefficients \( \alpha_{q,p} \) and \( \alpha_{q,p}^{m,k} \), that are used for the parametric solver, require information from the spatial solver. Solvers that provide access to the mathematical formulation of the problem enable a straightforward computation of these coefficients. Black box solvers may not give access to the required quantities. Here, we use an in-house electrothermal solver based on the FIT [12].

IV. Numerical Results

In this Section, the PGD algorithm for electrothermal problems, as presented in Section III, is applied to an electrothermal 1D model problem and to the electro simulation of a 3D microelectronic chip package. For all examples, \( \text{tol}_{FP} = 1 \times 10^{-7} \) was used as the tolerance for the fixed point iteration.

A. 1D Electrothermal Model Problem

We consider a 1D problem in the domain \( \Omega = (0, 2L) \), with \( L = 1 \), for which (12) simplifies to

\[
-\partial_x (\sigma(x; \mu) \partial_x \varphi(x; \mu)) = 0, \quad x \in \Omega, \quad (20a)
\]

\[
-\partial_x (\lambda(x; \mu) \partial_x T(x; \mu)) = \sigma(x; \mu) |\partial_x \varphi|^2, \quad x \in \Omega, \quad (20b)
\]

with the boundary conditions

\[
\varphi(0; \mu) = 0, \quad \varphi(2L; \mu) = 1, \quad (20c)
\]

\[
T(0; \mu) = 20, \quad T(2L; \mu) = 20, \quad (20d)
\]

and with the piecewise constant material coefficients

\[
\sigma(x; \mu) = \lambda(x; \mu) = \begin{cases} \mu, & \text{for } x \in (0, L), \\ 1, & \text{for } x \in (L, 2L). \end{cases}
\]

(21)

Problem (20) is discretized using the FIT with 200 elements in \( x \)-direction. Applying the PGD algorithm of Section III to (20), (10) simplifies for the electrokinetic part with \( q = 1 \) and \( p = 1 \) to

\[
\mathcal{L}(1, m) = \int_{(0,L)} \nabla_x u^m_1 \cdot \nabla_x v_1 \, dx \int_{I_1} \mu (u_0^2)^2 \, d\mu.
\]

(22)

Using \( I_1 = [0.2, 1] \), 100 points in \( \mu \)-direction and a tolerance of \( \text{tol}_{PGD} = 1 \times 10^{-2} \), an electrothermal PGD solution of 3 electric and 8 thermal modes is obtained. The relative error of the thermal solution of the resulting electrothermal surrogate model with respect to the analytical solution is shown in Fig. 5a, curve ii). Additionally, \( \|u^m\| \) and the discretization error of the FIT model are shown. In Fig. 5b, the convergence of the fixed point iteration for the individual modes of the thermal solution is observed by plotting the solution change.
\[ \Delta^{(k)} \] In total, only 11 electrothermal PGD modes and 42 spatial field solver calls are required to obtain an accurate model for a wide parameter range.

**B. Chip problem**

A microelectronic chip package geometry [13] with 12 internal bond wires of uncertain length is considered, see Fig. 1. The wires are modeled by an uncertain length \( L \in [10^{-4}, 10^{-3}] \) discretized by 100 points. Following [14], we model the wires using a 1D-3D coupling approach with zero coupling radius and two 1D points each. In Fig. 3 the relative error of the electric PGD solution with respect to the standard FIT solution is shown. The qualitative distribution for the first four spatial modes of the solution is shown in Fig. 5. Whereas the first mode is the main representative of the global field distribution, the following modes show the local influence of the uncertain wires. To obtain an accuracy of \( 1.14 \times 10^{-6} \), 100 modes using 326 spatial field solver calls were required.

**V. Conclusion**

A PGD formulation for static electrothermal problems including an arbitrary number of uncertain parameters has been presented. In contrast to other techniques, the PGD avoids exponential dependence of the computational cost on the number of uncertain parameters and thus, it circumvents the curse of dimensionality. The electrothermal coupling term requires a non-standard trilinear form modeling the resistive losses. For numerical validation, a 1D model problem and a 3D microelectronic chip package example were considered. Extending the algorithm to a transient bi-directional electrothermal coupling and the computation of multi-parameter examples are subject to future work. Using the PGD approach for optimization problems could be another possible application.

**ACKNOWLEDGMENT**

This work is supported by the “Excellence Initiative” of the German Federal and State Governments and by the Graduate School of Computational Engineering at Technische Universität Darmstadt.

**References**

[1] Francisco Chinesta, Roland Keunings, and Adrien Leygue. The Proper Generalized Decomposition for Advanced Numerical Simulations: A Primer. Springer, 2013.

[2] A. Ammar, B. Mokdad, F. Chinesta, and R. Keunings. A new family of solvers for some classes of multidimensional partial differential equations encountered in kinetic theory modelling of complex fluids. J. Non-Newtonian Fluid Mech., 144(2–3):98–121, July 2007.

[3] Yalan Li, Shuai Yan, Xiaoyu Xu, Pengfei Lyu, and Zhouxiang Ren. 3-D IC interconnect parasitic capacitance extraction with a reformulated PGD algorithm. IEEE Trans. Magn., 53(6):1–4, June 2017.

[4] Thomas Henneron and Stéphane Clénet. Proper generalized decomposition method applied to solve 3-D magnetoquasi-static field problems coupling with external electric circuits. IEEE Trans. Magn., 51(6):1–10, June 2015.

[5] M. Pineda-Sanchez, F. Chinesta, J. Roger-Folch, M. Riera-Guasp, J. Pérez-Cruz, and F. Daim. Simulation of skin effect via separated representations. COMPEL, 29(4):919–929, July 2010.

[6] Zhao Qin, Hakeim Talleb, and Zhuoxiang Ren. A proper generalized decomposition-based solver for nonlinear magnetothermal problems. IEEE Trans. Magn., 52(2):1–9, February 2016.

[7] Loïc Girald, Dishi Liu, Hermann G. Matthies, and Anthony Nouy. To be or not to be intrusive? the solution of parametric and stochastic equations. SIAM J. Sci. Comput., 37(1):A347–A368, January 2015.

[8] Thorben Casper, Ulrich Römer, and Sebastian Schöps. Determination of bond wire failure probabilities in microelectronic packages. In Chris Bailey, Sebastian Volz, András Poppe, and John Parry, editors, 22nd International Workshop on Thermal Investigations of ICs and Systems (THERMINIC 2016), pages 39–44, Budapest, Hungary, September 2016. IEEE.

[9] John David Jackson. Classical Electrodynamics. Wiley and Sons, New York, 3rd edition, 1998.

[10] D.V. Widder. The Heat Equation. Academic Press, 1975.

[11] Amine Ammar, Antonio Huerta, Francisco Chinesta, Elías Cueto, and Adrien Leygue. Parametric solutions involving geometry: A step towards efficient shape optimization. Comput. Meth. Appl. Mech. Eng., 268:178–193, 2014.

[12] Thomas Weiland. Time domain electromagnetic field computation with finite difference methods. Int. J. Numer. Model. Electron. Network. Dev. Field, 9(4):295–319, 1996.

[13] Thorben Casper, Herbert De Gersem, Renaud Gillon, Tomáš Göttmans, Tomáš Kratochvíl, Peter Meuris, and Sebastian Schöps. Electrothermal simulation of bonding wire degradation under uncertain geometries. In Luca Fanucci and Jürgen Teich, editors, Proceedings of the 2016 Design, Automation & Test in Europe Conference & Exhibition (DATE), pages 1297–1302. IEEE, April 2016.

[14] Thorben Casper, Ulrich Römer, Herbert De Gersem, and Sebastian Schöps. Coupled simulation of transient heat flow and electric currents in thin wires: Application to bond wires in microelectronic chip packaging. 2018.