

Modelling, Validation, and Design for Additive Manufacturing

Applications of numerical methods to 3D printing processes

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Outline

- Introduction
- Adaptive Isogeometric Analysis for Heat Transfer Problems
- Physical Modelling and Experimental Validations
- Functionally Graded Material Design for Additive Manufacturing
- Conclusions



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Additive Manufacturing generic production process



Laser powder bed fusion process



Image source: EOS Gmbh

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LPBF Numerical simulations: Multi-scale



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Why adaptive isogeometric analysis?

• The multi-scale nature of the problem calls for **adaptive numerical scheme** where refinement and coarsening can be efficiently performed.

• The high temperature gradients in the vicinity of the melt-pool regions can be well-approximated by means of high-order basis functions.

• **Goal**: Develop a multi-level, high-order and adaptive numerical method suitable to simulate heat transfer problems with a localized moving heat source.

Adaptive IGA for heat transfer problems

Truncated Hierarchical

B-splines (THB-splines)

THB-splines can be employed as a basis for adaptive isogeometric analysis, reducing interactions between different levels in the spline hierarchy.





Admissible meshes

A hierarchical mesh is admissible of class m if, for each element Q, the functions that do not vanish in Q belong to at most m successive levels



Giannelli, C., Jüttler, B., Speeler, H., THB-spline: the truncated basis for hierarchical splines, Comp. Aided Geom. Design. (2012)

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Admissible adaptive mesh

• Admissible refinement algorithm: Recursively refine the elements in the refinement neighborhood of the marked elements.



The final mesh automatically fulfills admissibility requirements!

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A. Buffa, C. Giannelli, Adaptive isogeometric methods with hierarchical splines: Error estimator and convergence, Mathematical Models and Methods in Applied Sciences 26 (2016) 1–25

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Admissible adaptive mesh

• Admissible mesh coarsening: A recursive algorithm to efficiently perform coarsening and at the same time automatically fulfills the admissibility requirements.



The final mesh automatically fulfills admissibility requirements!



Carraturo, M., Giannelli, C., Reali, A., Vazguez, R., Suitably graded THB-spline refinement and coarsening: Towards an adaptive isogeometric analysis of additive manufacturing processes, CMAME (2019) 3/18/2020

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Numerical example 2D

- Process parameters:
 - Laser power = 190 [W]
 - Laser speed = 800 [mm/s]
 - Laser radius = 0.05[mm]
 - Hatch distance = 0.05[mm]
 - Initial temperature = 25°
- Material parameters:
 - Absorptivity = 0.33
 - Conductivity = 29 [mW/m/K]
 - Heat capacity = 650 [J/kg/K]
 - Density = 8440[kg/m³]



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Numerical example 2D





Non-admissible grid

Admissible grid

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Numerical example 2D



Carraturo, M., Giannelli, C., Reali, A., Vazquez, R., Suitably graded THB-spline refinement and coarsening: Towards an adaptive isogeometric analysis of additive manufacturing processes, CMAME (2019) 3/18/2020

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Physical modelling

• Temperature-based phase-change model:

• Heat equation is expressed in terms of enthalpy (H) and temperature (T):

$$\begin{aligned} \frac{\partial H}{\partial t} - \nabla(k\nabla T) &= Q & \text{in } \Omega \\ -k\nabla T &= q & \text{on } \Gamma_N \\ H(t) &= \int_{T_{ref}}^{T} \rho c(T) dT + \rho L f_{pc}(T) = H^c + H^l \\ f_{pc}(T) &= \frac{1}{2} \left[\tanh\left(S\frac{2}{T_l - T_s}\left(T - \frac{T_s - T_l}{2}\right)\right) + 1 \right] & \overset{f_{re}}{\overset{1}{=}} \int_{S=3}^{S=2} S = 4 \\ & \overset{f_{re}}{\overset{f_{re}}{=}} \int_{T_{ref}}^{T} \rho c(T) dT + \rho L f_{pc}(T) = H^c + H^l \\ & \overset{f_{re}}{\overset{f_{re}}{=}} \int_{T_{ref}}^{T} \rho c(T) dT + \rho L f_{pc}(T) = H^c + H^l \\ & \overset{f_{re}}{\overset{f_{re}}{=}} \int_{T_{ref}}^{T} \rho c(T) dT + \rho L f_{pc}(T) = H^c + H^l \\ & \overset{f_{re}}{\overset{f_{re}}{=}} \int_{T_{ref}}^{T} \rho c(T) dT + \rho L f_{pc}(T) = H^c + H^l \\ & \overset{f_{re}}{\overset{f_{re}}{=}} \int_{T_{ref}}^{T} \rho c(T) dT + \rho L f_{pc}(T) = H^c + H^l \\ & \overset{f_{re}}{\overset{f_{re}}{=}} \int_{T_{ref}}^{T} \rho c(T) dT + \rho L f_{pc}(T) = H^c + H^l \\ & \overset{f_{re}}{\overset{f_{re}}{=}} \int_{T_{ref}}^{T} \rho c(T) dT + \rho L f_{pc}(T) = H^c + H^l \\ & \overset{f_{re}}{\overset{f_{re}}{=}} \int_{T_{ref}}^{T} \rho c(T) dT + \rho L f_{pc}(T) = H^c + H^l \\ & \overset{f_{re}}{\overset{f_{re}}{=}} \int_{T_{ref}}^{T} \rho c(T) dT + \rho L f_{pc}(T) = H^c + H^l \\ & \overset{f_{re}}{\overset{f_{re}}{=}} \int_{T_{ref}}^{T} \rho c(T) dT + \rho L f_{pc}(T) = H^c + H^l \\ & \overset{f_{re}}{\overset{f_{re}}{=}} \int_{T_{ref}}^{T} \rho c(T) dT + \rho L f_{pc}(T) = H^c + H^l \\ & \overset{f_{re}}{\overset{f_{re}}{=}} \int_{T_{ref}}^{T} \rho c(T) dT + \rho L f_{pc}(T) = H^c + H^l \\ & \overset{f_{re}}{\overset{f_{re}}{=}} \int_{T_{ref}}^{T} \rho c(T) dT + \rho L f_{pc}(T) = H^c + H^c \\ & \overset{f_{re}}{\overset{f_{re}}{=}} \int_{T_{ref}}^{T} \rho c(T) dT + \rho L f_{pc}(T) = H^c + H^c \\ & \overset{f_{re}}{\overset{f_{re}}{=}} \int_{T_{ref}}^{T} \rho c(T) dT + \rho L f_{pc}(T) = H^c + H^c \\ & \overset{f_{re}}{\overset{f_{re}}{=}} \int_{T_{ref}}^{T} \rho c(T) dT + \rho L f_{pc}(T) = H^c + H^c \\ & \overset{f_{re}}{\overset{f_{re}}{=}} \int_{T_{ref}}^{T} \rho c(T) dT + \rho L f_{pc}(T) = H^c \\ & \overset{f_{re}}{\overset{f_{re}}{=}} \int_{T_{ref}}^{T} \rho c(T) dT + \rho L f_{pc}(T) \\ & \overset{f_{re}}{\overset{f_{re}}{=}} \int_{T_{ref}}^{T} \rho c(T) dT \\ & \overset{f_{re}}{\overset{f_{re}}{=} \int_{T_{ref}}^{T} \rho c(T) dT \\ & \overset{f_{re}}{\overset{f_{re}}{=}} \int_{T_{ref}}^{T} \rho c(T) dT \\ & \overset{f_{re}}{\overset{f_{re}}{=} \int_{T_{ref}}^{T} \rho c(T) dT \\ & \overset{f_{re}}{\overset{f_{re}}{=} \int_{T_{ref}}^{T} \rho c(T) dT \\ & \overset{f_{re}}}{\overset{f_{re}}{=} \int_{T_{ref}}^{T} \rho c(T) dT \\ & \overset{f_{re}}{\overset{$$



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Physical modelling

• Energy input from power density measurements:



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Physical modelling

• Anisotropic conductivity: allows to partially consider melt-pool dynamics in the model.



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Kollmannsberger, S., Carraturo, M., Auricchio, F. and Reali, A. (2019), Accurate Prediction of Melt Pool Shapes in Laser Powder Bed Fusion by the Non-Linear Temperature Equation Including Phase Changes. IMMI. 3/17/2020

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Experimental validation (AMBench2018)

- Adjacent, independent laser scans using 3 different combinations (case A, B, and C) of laser power and speed
- Material:INCONEL 625 (a Nickelbased superalloy widely used in AM applications)
- No powder is involved





Source: https://www.nist.gov/ambench/amb2018-02-description

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Source: https://www.nist.gov/ambench/amb2018-02-description

Experimental validation (AMBench2018)

• *Ex-situ* measurements of the melt-pool cross section



• *In-situ* measurements of the meltpool length. Mod<u>elled</u> Equiv. Signal [DL] Optical Blur, Motion Blur, Spatial Digitization (3.3 µm/pixel)







Kollmannsberger, S., Carraturo, M., Auricchio, F. and Reali, A. (2019), Accurate Prediction of Melt Pool Shapes in Laser Powder Bed Fusion by the Non-Linear Temperature Equation Including Phase Changes. IMMI.

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Objective:

- Minimize the compliance (i.e., maximize the stiffness) of the structure for linear elastic problems
- Obtain a functionally graded lattice design with varying density

Phase-field Method:

- No filtering methods required (cfr. SIMP approaches)
- No function re-initialization required (cfr. Level-set method)



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Objective

Minimize the compliance of the structure defined as:

 $\int_{\Gamma_N} \mathbf{g} \cdot \mathbf{u}(\varphi, \chi) \mathrm{d}\Gamma$

Phase-field variables $0 \le \varphi \le 1$ a.e. in Ω Defines the material regions $0 < \chi < \varphi$ a.e. in Ω Defines the density of the material **Extended** objective functional Double-well potential $\mathcal{J}^{\varepsilon}(\mathbf{u},\varphi,\chi) = \kappa_{\varphi} \int_{\Omega} \left(\frac{\mathcal{W}(\varphi)}{\varepsilon_{\varphi}} + \varepsilon_{\varphi} \frac{|\nabla\varphi|^2}{2} \right) \mathrm{d}x + \kappa_{\chi} \int_{\Omega} \varepsilon_{\chi} \frac{|\nabla\chi|^2}{2} \mathrm{d}x + \int_{\Gamma_N} \mathbf{g} \cdot \mathbf{u} \mathrm{d}x$ Free-energy functional Density gradient term 3/17/2020 M. Carraturo 23

Objective

Minimize the extended objective functional: $\mathcal{J}^{\varepsilon}(\mathbf{u},\varphi,\chi)$

under the constraints:

• Volume constraint is introduced using the Lagrange multiplier $\lambda_{\varphi} \in \mathbb{R}$

$$\Longrightarrow \lambda_{\varphi} \left(\int_{\Omega} \varphi \mathrm{d}x - \int_{\Omega} m_{\varphi} \mathrm{d}x \right) =: \lambda_{\varphi} M_{\varphi} = 0$$

• Mass constraint is introduced using the Lagrange multiplier $\lambda_{\chi} \in \mathbb{R}$ and controlled by a mass fraction parameter $m_{\chi} < m_{\varphi}$ defining the target mass fraction in the optimized structure w.r.t. the mass of the initial volume filled with purely bulk material.

Objective

Minimize the extended objective functional: $\mathcal{J}^{\varepsilon}(\mathbf{u},\varphi,\chi)$

under the constraints:

• The mechanical equilibrium equations are satisfied:

$$-\nabla \cdot \boldsymbol{\sigma} = \mathbf{0} \quad \text{in } \Omega$$
$$\boldsymbol{\sigma} \cdot \mathbf{n} = \mathbf{g} \quad \text{on } \Gamma_g$$
$$\mathbf{u} = \mathbf{u}_0 \quad \text{on } \Gamma_d$$
$$\boldsymbol{\sigma} = \mathbf{C}(\varphi, \chi) : \boldsymbol{\varepsilon}(\mathbf{u}) \quad \text{in } \Omega$$

Obtained from **asymptotic homogenization** on a lattice RVE with periodic BCs





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MBB 2D example

- Messerschmitt-Bölkow-Blohm (MBB) beam problem:
 - g=25 N
 - Material: RGD851 rigid polymer from Stratasys (E=2.3GPa and v=0.3)
 - 3D printer machine: Stratasys Objet 260 Connex 3
 - Volume fraction = 0.6
 - Mass fraction = 0.4





MBB 2D example



From numerical analysis to 3D printing



Alaimo G., Carraturo M., Rocca E., Reali A., Auricchio F., Functionally graded material design for plane stress structures using phase field method, II International Conference on UNIVERSITÀ Simulation for Additive Manufacturing - Sim-AM 2019 3/17/2020

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Experimental measurements

• **Objective:** Evaluate the improvement in terms of max. displacements obtained in the optimized specimen w.r.t. a uniform specimen of the same weight.



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Conclusions

- Adaptive IGA is an attractive methodology to deal with heat transfer problem with moving heat source.
- A thermal model based on a **data fitting** of power density measurements and **anisotropic conductivity** is validated w.r.t. thermal camera measurements.
- A phase-field topology optimization approach can be used to design functionally graded lattice structures with higher stiffness compared to uniform lattice structures of the same weight.



Further outlook



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The 7 categories of AM (ASTM/ISO)



Fused deposition modeling (FDM)Materials: ABS, ceramics

• Applications: tissue/scaffolds



Binder jetting

- Materials: polymers, ceramics, and metals
- Applications: arts, prototyping



Material jetting

- Materials: polymers
- Applications: electrical and chemical industry



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Vat Photopolymerization

- Materials: polymers
- Applications: coating and printing industry



Sheet lamination

- Materials: metals
- Applications: prototyping



Direct energy deposition (DED)

- Materials: metals
- Applications: repairing/joining metal components



Laser powder bed fusion (LPBF)

- Materials: metals
- Applications: aerospace and biomedical industry

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Laser powder bed fusion features

Benefits:

- Complex geometries
- Better material properties

Drawbacks:

- High residual stresses induced by the process
- Needs of supports
- Higher accuracy and surface finish
- High costs



LPBF Numerical simulations: Challenges

Complex geometrical features







LPBF Numerical simulations: Challenges



THB-splines features

- Key features:
 - Local Linear independence
 - Local and compact support
 - Two scale relation
 - Partition of unity
 - Non-negativity





Adaptive IGA

- From Buffa and Giannelli (2016) we take the definitions of multilevel *support extension* and *refinement neighborhood* of an active element *Q* of level *I*:
 - Definition support extension: the support of B-splines of level k which do not vanish on the element Q

$$S(\hat{Q},k) := \left\{ \hat{Q'} \in \hat{G}^k : \exists \hat{\beta} \in \hat{\mathcal{B}}^k, \operatorname{supp} \hat{\beta} \cap \hat{Q'} \neq \emptyset \land \operatorname{supp} \hat{\beta} \cap \hat{Q} \neq \emptyset \right\}.$$

• Definition refinement neighborhood: set of elements of level l - m + 1 with a child in the support extension of Q.

$$\mathcal{N}_r(\hat{Q},m) := \left\{ \hat{Q}' \in \hat{\mathcal{G}}^{\ell-m+1} : \exists \, \hat{Q}'' \in S(\hat{Q},\ell-m+2), \hat{Q}'' \subseteq \hat{Q}' \right\}.$$

A 3/18/202

39

Adaptive IGA

• Definition *coarsening neighborhood* of an element *Q* of level *l* : the set including all the active elements of level *l*+*m* which are in the support extension of the children of the element *Q*:

$$\mathcal{N}_c(\hat{\mathcal{Q}}, \hat{Q}, m) := \left\{ \hat{Q}' \in \hat{\mathcal{G}}^{\ell+m} : \exists \, \hat{Q}'' \in \hat{\mathcal{G}}^{\ell+1} \text{ and } \hat{Q}'' \subset \hat{Q}, \text{ with } \hat{Q}' \subset S(\hat{Q}'', \ell+1) \right\}.$$

• Property: When the coarsening neighborhood of an element to be reactivated turns out to be empty it ensures that all the re-activated functions are fully truncated on level *I*+1.



3/17/2020

40

Adaptive IGA

Algorithm 5 coarsen Input: Q, M_c, m Output: Q 1: for $Q \in \mathcal{M}_c$ do 2: $\mathcal{R}_c \leftarrow \mathcal{R}_c \cup \texttt{get_parent}(Q)$ 3: end for 4: for $Q \in \mathcal{R}_c$ do ▷ This loop must be done from the finest to the coarsest level $Q_c \leftarrow \texttt{get_children}(Q)$ 5: if $(Q_c \subset \mathcal{M}_c \text{ and } \mathcal{N}_c(\mathcal{Q}, Q, m) = \emptyset)$ then 6: update Q by activating Q and removing its children Q_c 7: end if 8: 9: end for





Thermografic measurement model

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43

Part-scale Additive Manufacturing Thermal Process Simulations using the Finite Cell Method

Standard AM-design process



FCM AM-design-through-analysis

• The thermo-mechanical analysis is performed directly on the CAD model.



• The STL repair step is required only once the final design is ready to be printed.

• Initial domain discretization



• Numerical integration using Quad/Octree partitioning





Other integration schemes could be used:

Х

Х

Х

- Smart-Octree (Kudela et al. 2016)
- Moment fitting (Hubrich et al. 2016)



- We distinguish between two different kind of layers:
 - 1. Cell layer: the layer of the finite cells supporting the basis functions
 - 2. Powder (or physical) layer: it corresponds to the actual layer of powder spread by the machine



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• Growing domain





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Part-scale immersed thermal analysis



Carraturo M., Jomo J., Kollmannsberger S., Reali A., Auricchio, F. and Rank E. (*submitted*), Modeling and experimental validation of an immersed thermo-mechanical part-scale analysis for laser powder bed fusion processes. Additive Manufacturing. 3/17/2020 M. Carraturo 49

GE-bracket immersed thermal process analysis

GE-bracket:

- Optimized AM component
- Material: Stainless steel 316L
- Dimensions: 48.5x80.0x28.5 mm^3
- # cells: 60x100x72
- # integration voxels/cell: 4x4x4
- # GPs/voxel: 8
- # powder-layer/cycle: 5
- Powder-layer thickness: 50 μm
- Number of time steps: 120





GE-bracket immersed thermal process analysis





Growing domain thermo-mechanical problem solver



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Experimental Validation

• Experimental setup





Experimental Validation

• Simulation vs. Experiment comparison

Max. deflection relative error < 5% Data correlation ~ 99%





Functional Design for AM

• Multi-step design process driven by optimization, taking advantages from Additive Manufacturing but also knowing the limitations of the process.



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Objective

Minimize the Lagrangian functional such that the admissibility conditions are fulfilled: $\min \mathcal{L}(\mathbf{u}, \mathbf{p}, \varphi, \chi, \lambda_{\varphi}, \lambda_{\chi}) = \mathcal{J}^{\varepsilon}(\mathbf{u}, \varphi, \chi) + \lambda_{\varphi} \mathcal{M}_{\varphi} + \lambda_{\chi} \mathcal{M}_{\chi} + \mathcal{S}(\mathbf{p}, \varphi, \chi)$

$$D_{\varphi} \mathcal{L}(\bar{\varphi}, \bar{\chi}, \bar{\mathbf{u}}, \bar{\mathbf{p}}, \bar{\lambda}_{\varphi}, \bar{\lambda}_{\chi}) (\varphi - \bar{\varphi}) \ge 0 \quad \forall \varphi \in \Phi_{ad}$$
$$D_{\chi} \mathcal{L}(\bar{\varphi}, \bar{\chi}, \bar{\mathbf{u}}, \bar{\mathbf{p}}, \bar{\lambda}_{\varphi}, \bar{\lambda}_{\chi}) (\chi - \bar{\chi}) \ge 0 \quad \forall \chi \in \Xi_{ad}$$

 $\Phi_{ad} := \{ \varphi \in H^1(\Omega) : 0 \le \varphi \le 1 \text{ a.e. in } \Omega \}$ $\Xi_{ad} := \{ \chi \in H^1(\Omega) : 0 \le \chi \le \phi \text{ a.e. in } \Omega \}$



Auricchio F., Carraturo M., Bonetti E., Hömberg D., Reali A. and Rocca, E. (submitted). A phase-field based graded-material topology optimization with stress constraint. M3AS.

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Homogenization

• Asymptotic homogenization using Ansys



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Topology Optimization

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Phase-field based topology optimization of MBB-beam problem



3D virtual model reconstruction



