Digital transformation in materials science and the role of a common ontologie

C. Eberl, et al., Regensburg, 2019



Digital transformation in materials science and engineering: Boundary conditions for manufacturers in Europe

- Materials expenses for industrial manufacturing are 56.7 % (personnel expenses 18.6, destatis Germany 2014) – improving on materials efficiency has a 10-fold higher impact than energy efficiency or 30-fold higher than logistics improvement.
- Ressource-rich countries use their commodities strategically (z.B. Steel, Cu, Rare Earth Metals, Oil).
- The global competition shifts from a productivity challenge to a purchase market although they are connected!
- Climate changes, ressource scarcity and the increasing population need political as well as technological solutions.
- Technological shifts have been accelerating, new tools have been made available, especially in the big data community

Speed, flexibility and adaptivity in product development depends on the ability to develop novel materials and bring them into production will be key to compete in the future market.



Digital transformation in materials science and engineering: Make materials behavior available in a digital form

- Connecting product development to materials development
- Through Industry 4.0: Connecting materials information into the processing chain
- Higher safety, reliability, functionality and adaptivity to market changes









McLean P Echlin*, William C Lenthe and Tresa M Pollock The size of the representative volume element - RVE





Vision meets pragmatism The digital representation of materials



- Holistic approach versus relevant materials information
- Availability of materials behavior during processing and in applications
- Information on processing and loading history to predict behavior



Development of microstructure-property relation in lamellar cast iron



M. Metzger, C. Schweizer et al.



intern

Development of microstructure-property relation in lamellar cast iron





intern

What will be necessary to connect materials knowledge to processing Transient behavior based on the simulation of the microstructural evolution



Materials Data Space:

Experimental and sensor raw- and meta data, physical and data based materials models,

D. Helm et al.



11 © Fraunhofer IWM Adaptive materials processing is based on materials knowledge Digitizing Materials within the Industry 4.0 Initiative



Materials Data Space:

Experimental and sensor raw- and meta data, physical and data based materials models,



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The key performance indicators change!

Real time physical/statistical material modells, e.g.: neural networks trained with data from the **Materials Data Space**

Materials Data Space:

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Sensor Input

Digital Workflows We need to develop the digital infrastructure



- Automated data generation: processing , experimentation and simulation
- Automated 3D microstructural analysis
- Filling in missing materials data through virtual testing
- Establishing materials data spaces containing the materials history and predict ist future behavior
- Development of real time materials models through machine learning



Digital Workflows Digital infrastructure needs a common ontology



The European Materials Ontology is out and we can start implementing:

- <u>https://github.com/emmo-repo</u>
- https://emmc.info/wp-content/uploads/2019/06/1-Gerhard-Goldbeck-EMMO.pdf
- https://emmc.info/wp-content/uploads/2019/04/Part 1 Ontology Intro.pdf
- https://emmc.info/wp-content/uploads/2019/04/Part 2 EMMO Intro.pdf
- <u>https://emmc.info/wp-content/uploads/2019/04/Part 3 Interoperability Intro.pdf</u>



Developing a structured data space (not yet EMO) From single process steps towards a structured data space



C. Schweizer, H. Oesterlin, E. Augenstein, A. Hashibon, V. Friedmann



How to structure knowledge and data? (not yet EMO) Materials ontology implemented into knowledge graphs



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C. Schweizer, H. Oesterlin, E. Augenstein, V. Friedmann, J. Preußner Example work flow: Casting process







C. Schweizer, H. Oesterlin, E. Augenstein, V. Friedmann, J. Preußner Example work flow: Casting process - sawing







C. Schweizer, H. Oesterlin, E. Augenstein, V. Friedmann, J. Preußner Example work flow: Casting process - embedding





21

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C. Schweizer, H. Oesterlin, E. Augenstein, V. Friedmann, J. Preußner Example work flow: Casting process - polishing









C. Schweizer, H. Oesterlin, E. Augenstein, V. Friedmann, J. Preußner Example work flow: Casting process - etching







C. Schweizer, H. Oesterlin, E. Augenstein, V. Friedmann, J. Preußner Example work flow: Casting process - microscopy







C. Schweizer, H. Oesterlin, E. Augenstein, V. Friedmann, J. Preußner **Example work flow: Casting process – a glimpse of the workflow**









HUB Digital@IWM

New structures are needed for the digital transformation

MBE Material Di leibnit HCE Jairan Eigenforschung MANO UNVERT Semantics and Ontology MAVO hALU_3D MaterialDigital@BW HUB Digital ICT, IPM, Landesinstitute **Structured Data** MAYO Mesic ENI MANS BANA CPM **Data Analysis**



Are there initiatives which help us in the process?





Bundesministerium für Bildung und Forschung Bundesministerium für Wirtschaft und Energie

European Materials Modeling Council EMMC **European Materials Characterization Council** EMCC



Fraunhofer Materials Data Space



Fraunhofer Industrial Data Space









VDI

28 © Fraunhofer IWM We can only master the challenges within the digital transformation together!