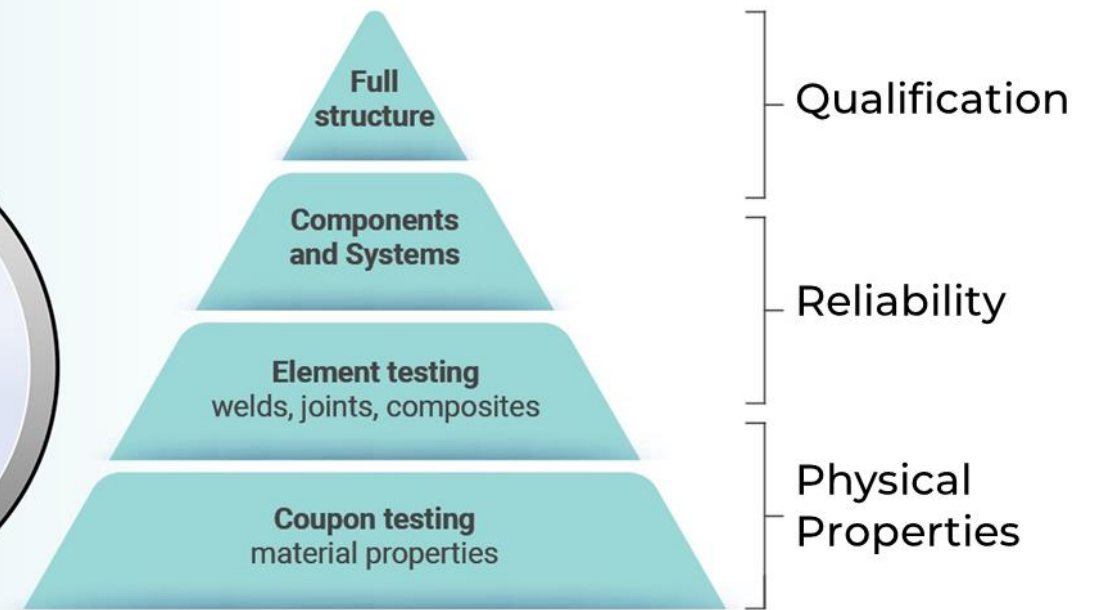
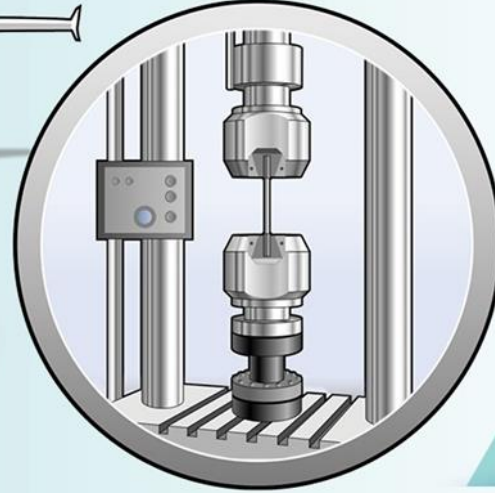
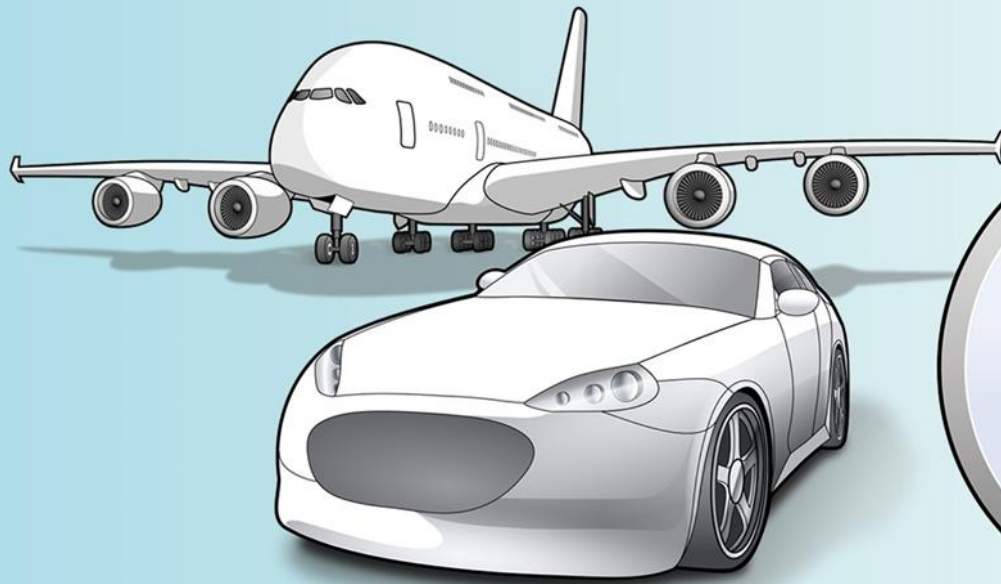
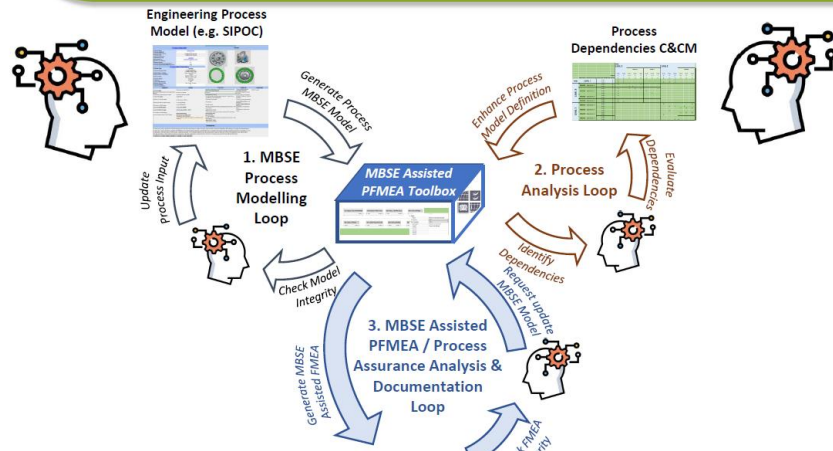




2023 HBK Technology Days

21 November 2023

MBSE-assisted FMEA Generation



- Placing the Expert in-the-Loop: Framework for MBSE-assisted FMEA Generation

Prof. Felician Campean,
University of Bradford

Process Function Failure Analysis				Process Controls		
Process Functions / Requirements	Failure Modes	Effects	Causes	CC	Prevention	Detection

Placing the Expert in-the-Loop: Framework for MBSE-assisted FMEA Generation

Abstract

- While significant progress has been made with the development and adoption of computer modelling and simulation tools to assist with the systems design, many of the analysis methods, in particular those focussed on robustness and reliability for the design and process assurance, including the FMEA, remain expert-centred and time and resource intensive. This talk outlines the development of an MBSE-assisted automatic FMEA generation tool, that places the expert in-the-loop, tasked with providing inputs and systematically reviewing the outcomes as generated system models. The approach emphasizes integration and traceability of requirements and critical characteristics across the levels of decomposition and analysis of the system.
- A case study of a robotic manufacturing process design for an electric drive unit is used to illustrate the use of the tool in an industrial context. This illustrates the cascade of the critical characteristics across multiple levels of process and tooling design analysis, supporting the synthesis of robust process controls plans for an Industry 4.0 implementation. As well as discussing the effectiveness of the tool, a reflection on the interaction between the analyst / expert, the MBSE model and FMEA, facilitated by the tool is provided. The practical benefits are the robustness and integrity of the FMEA and other design assurance documentation, and their governance for future use.

Placing the Expert in-the-Loop: Framework for MBSE-assisted FMEA Generation

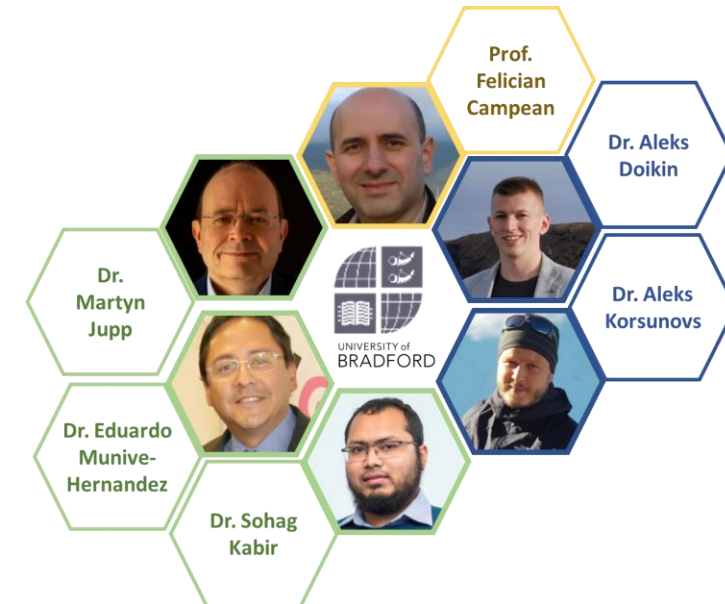
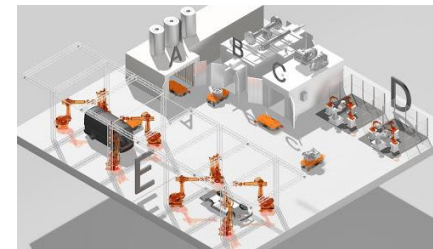
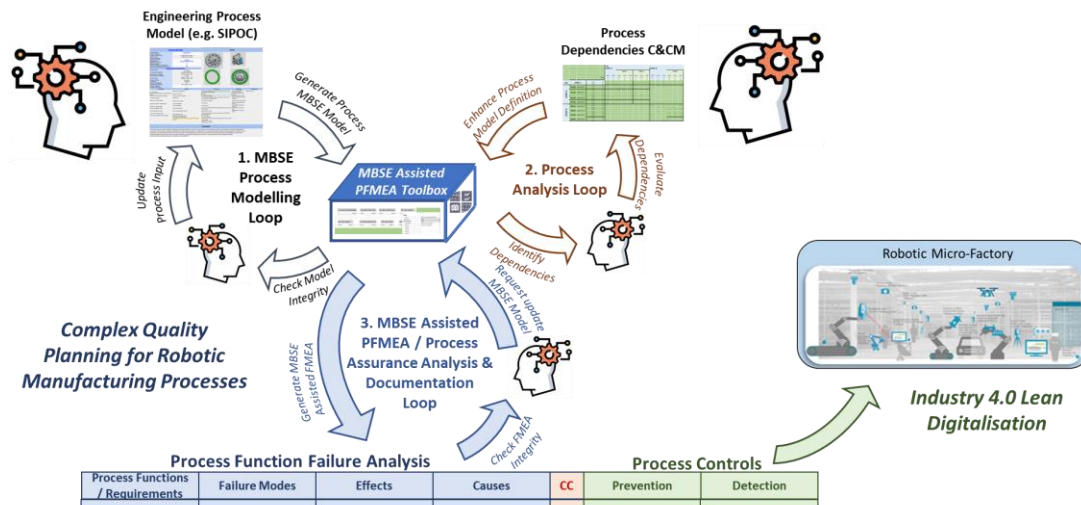
▲ About the presenter, Prof. Felician Campean, University of Bradford

- Felician is a Professor in Reliability Engineering, Director of the Automotive Research Centre at the University of Bradford, and Associate Dean for Research & Innovation in the Faculty of Engineering and Digital Technologies. He holds a PhD in Reliability from Brunel University (1998) and a Mechanical / Manufacturing Engineering Degree from Transilvania University (1990).
- His research interests revolve around the development of methods and tools to support design for performance, reliability and robustness in the product development of automotive and aerospace complex systems. This includes model-based methods for systems engineering, reliability, robustness and resilience analysis, AI big data analytics methods for intelligent systems lifecycle management, multi-disciplinary design optimization, modelling complex manufacturing and product development processes. He has an extensive track record of working with industry on collaborative research – both fundamental and applied, as well as effective knowledge transfer, with strong evidence of impact to industry.



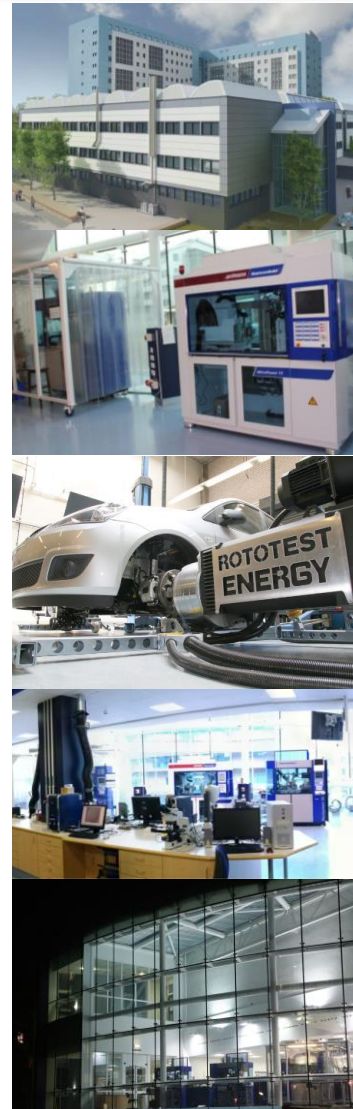
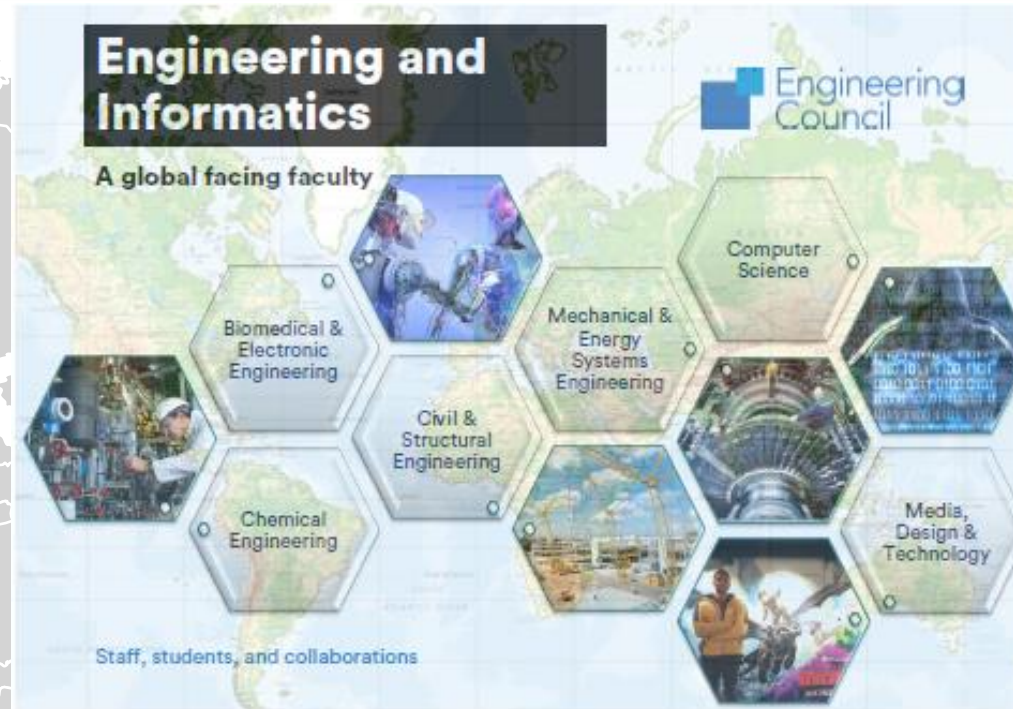
Placing the Expert in-the-Loop: Framework for MBSE-assisted Automatic FMEA Generation

Felician Campean
Aleks Korsunovs
Aleks Doikin
University of Bradford, UK



University of Bradford

UNIVERSITY of
BRADFORD



Automotive Research Centre



- Launched in 1995 – in partnership with Ford
- Enduring collaboration with the Automotive Industry

Research focus:

- Reliability; product development; systems engineering; “Failure Mode Avoidance”



- Launched in 2016 – in partnership with JLR
- Cross-disciplinary collaboration ARC – Computer Science

Research focus:

- ML and AI applied to automotive and transportation problems, CPS & Industry 4.0

Brakes Research Centre

- Launched in 2010 – as Brakes Research Centre
- Strong track record of expertise in brakes

Research focus:

- Brakes Noise, Vibration & Harshness
- USPs: Unique Brakes NVH expertise and Testing Facilities; BRV Course.

Engine & Powertrain Research Laboratory

- Launched in 2010 – as Hyper-C (£2m development)
- 2 transient dynos, including 500KW regen dyno
- Emissions capability – Particle counter

Research focus:

- Powertrain modelling & Optimisation

Tribology Research Laboratory

Research focus:

- Engine tribology
- Lubricants

Materials & Surface Engineering Lab

Research focus:

- Surface coating;
- Joining & welding.
- Corrosion

Acoustics Research Laboratory

Research focus:

- Acoustic materials characterisation;
- New materials with acoustic properties

Automotive Systems Reliability Engineering



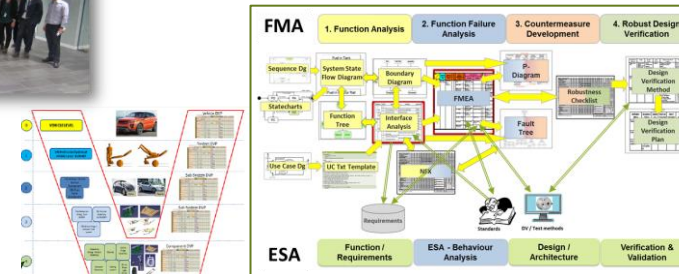
Bradford Engineering Quality Improvement Centre



- Launched in 1995
- Enduring collaboration with the Automotive & Manufacturing Industries

Research / Knowledge Exchange focus:

- Reliability & Robustness / "Failure Mode Avoidance"
- Process / Productivity Improvement
- Product Development
- Systems Engineering
- Multidisciplinary Design Optimisation
- Skills development



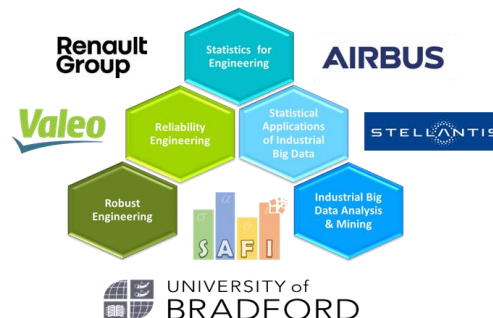
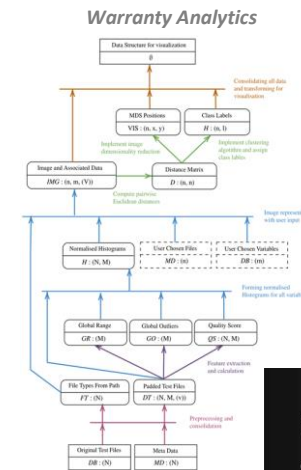
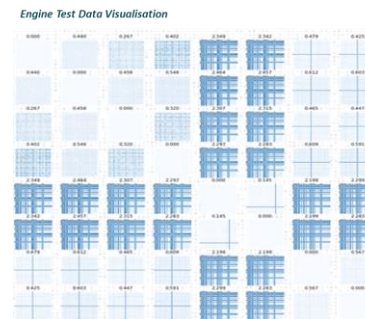
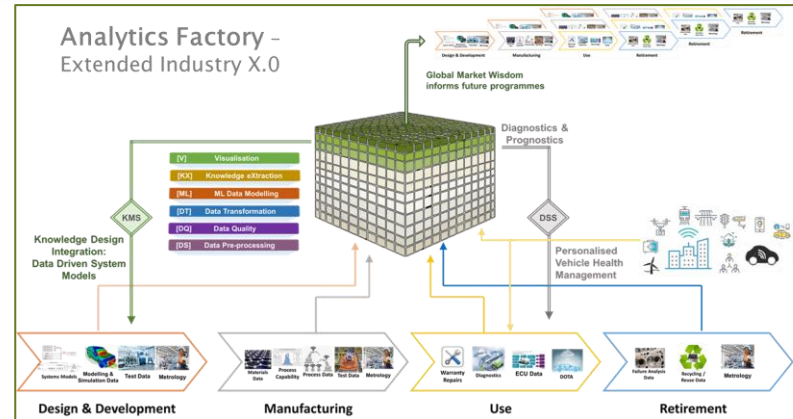
Advanced Automotive Analytics Laboratory



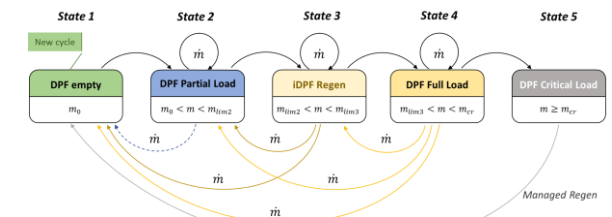
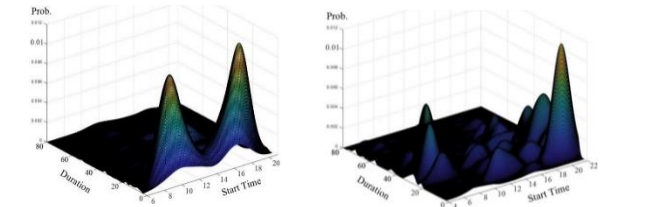
- Launched in 2016 Interdisciplinary collaboration

Research / Knowledge Exchange focus:

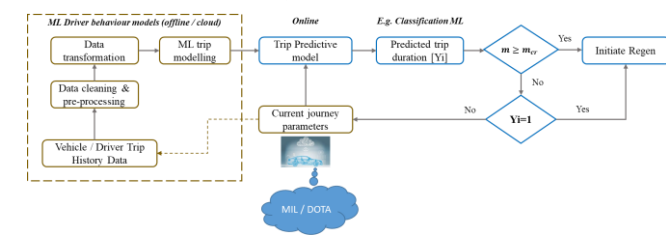
- Intelligent Systems Health Management, CPS & Industry-X.0
- Robustness & Safety of embedded AI
- Augmented intelligence for design & manufacturing
- Resilience & trust of autonomous systems
- Data centric life-cycle optimisation
- Skills development - SAFI



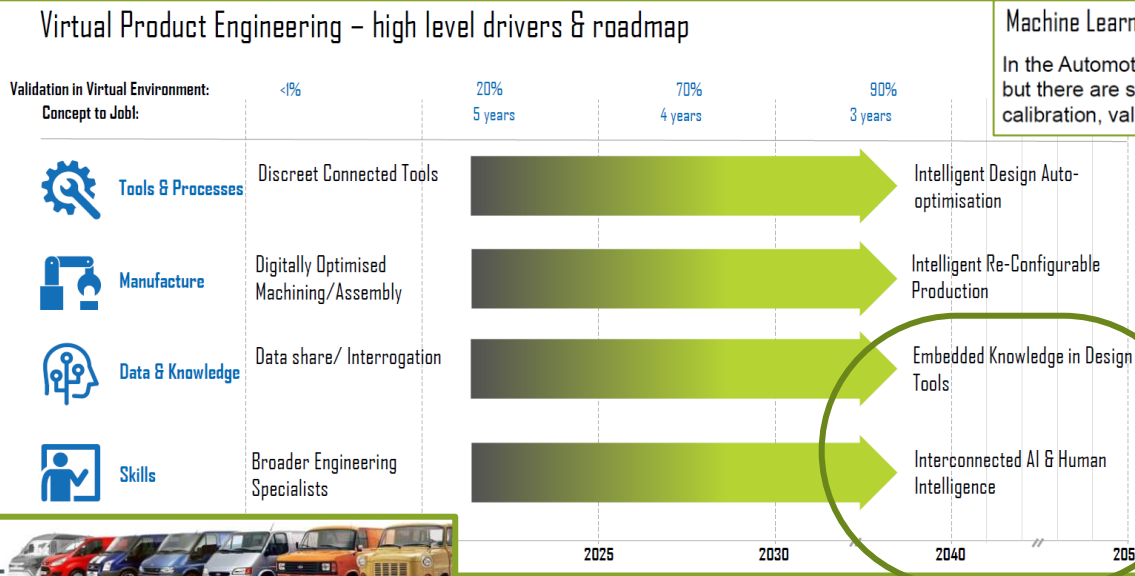
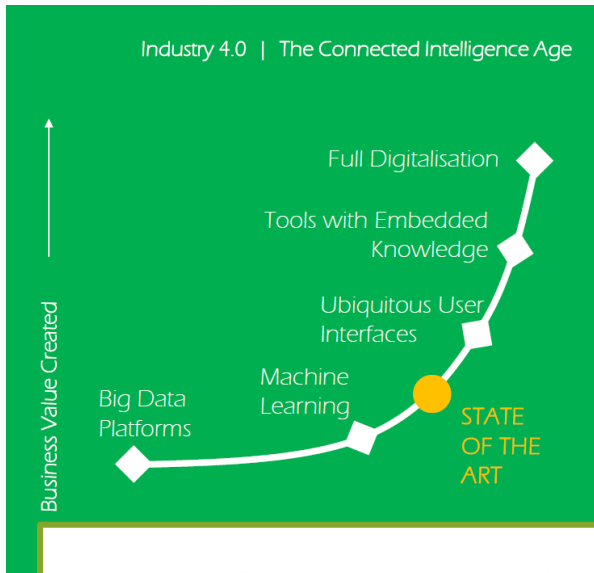
Driver Behaviour Modelling



iDPF Control Strategy



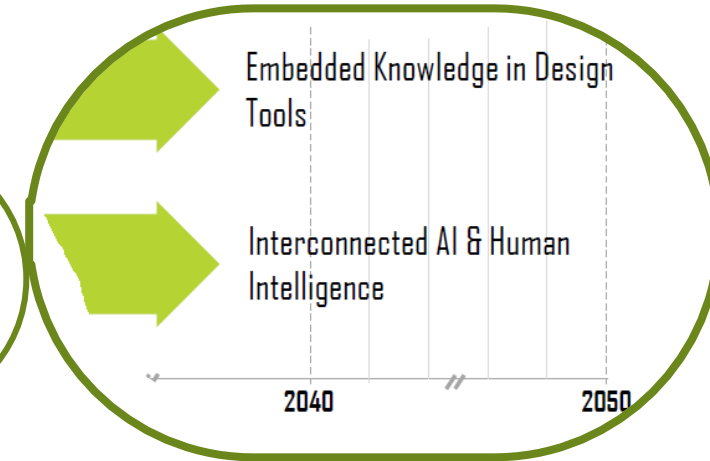
Motivation: Automotive Council Digitalisation Roadmap 2050



Machine Learning & AI in Product design and development

In the Automotive sector, machine learning and AI are mostly associated with automated driving but there are significant opportunities to employ these techniques in design, development, calibration, validation & manufacturing processes

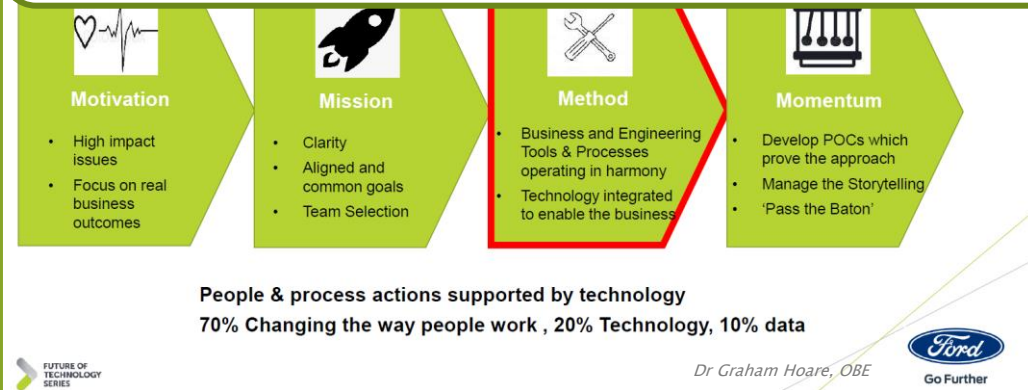
Prof Neville Jackson, FEng



CREATING AMPLIFIED INTELLIGENCE



Augmenting People with Technology, Data & Insights. Leveraging the Agile Methodology for Speed



Research challenge: how do we evolve the PD methodologies, methods and tools to create **amplified intelligence** through **interconnected AI & Human Intelligence**?

Motivation: R&D Opportunity



UK Research
and Innovation



ADVANCED
PROPULSION
CENTRE UK

NextGenDrive R&D Project



- Aim: develop the prototype for a “robofactory” (robotic micro-factory) for EDUs
 - The robofactory is based on the “vertical integration” principles – the whole unit made and assembled in a fully robotised microfactory, leveraging I4.0 manufacturing & automation methods and local supply chains
 - EDU redesigned for robotic manufacturing
 - Full virtual integration of EDU and robotic manufacturing process design for early validation
- Research challenge for the UoB team: robotic manufacturing process design assurance and I4.0 smart lean digitalisation
 - Use the FMEA integrated with the manufacturing process design to derive a *sufficient* robust control plan and design the I4.0 implementation
- Research approach: MBSE modelling for robotic manufacturing process design with MBSE-Assisted Automatic FMEA Generation

MBSE-assisted Auto-FMEA Approach

FMEA Challenges

Human factors / effectiveness:

- Labour intensive / time consuming
- Error prone
- Difficult to update & interpret

Outcomes:

- Uncertain completeness and integrity of the analysis
- Unwieldy / not reflecting the recursive analysis at different levels

Integration with Process Design flow:

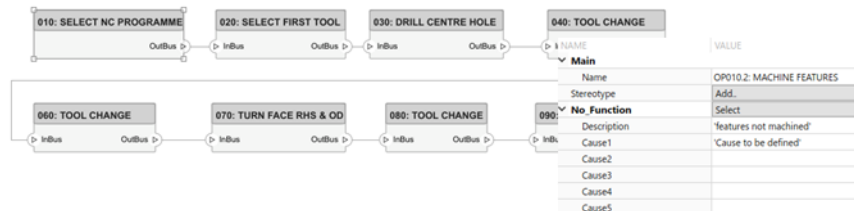
- FMEA not updated to reflect changes to design / model
- Lack of traceability between high level requirements and process controls

MBSE-assisted Auto-FMEA approach

Engineering Process Model e.g. SIPOC



MBSE Model



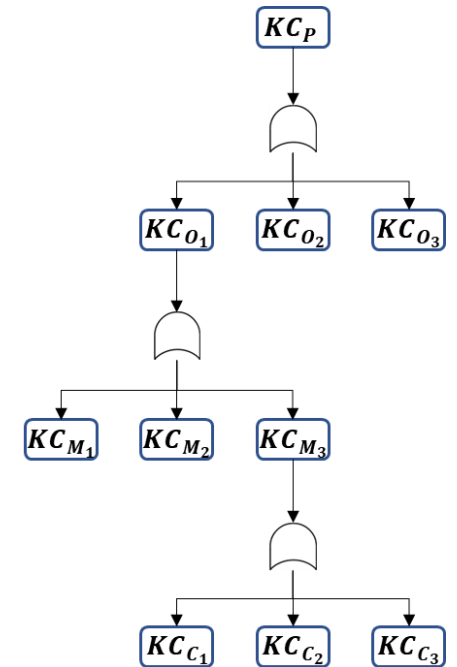
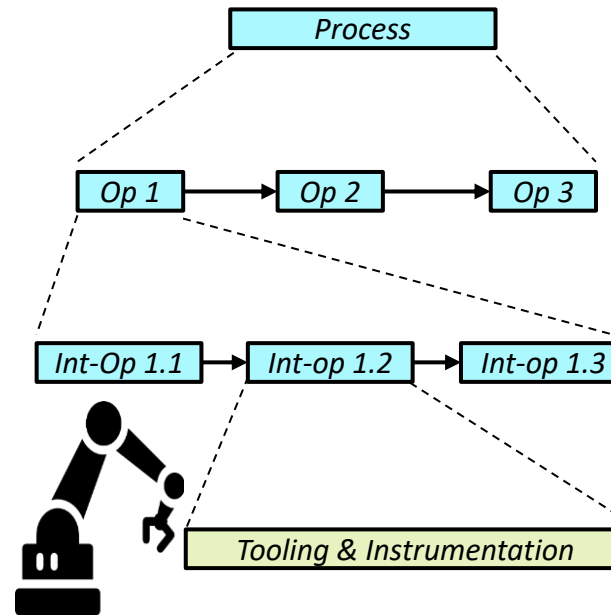
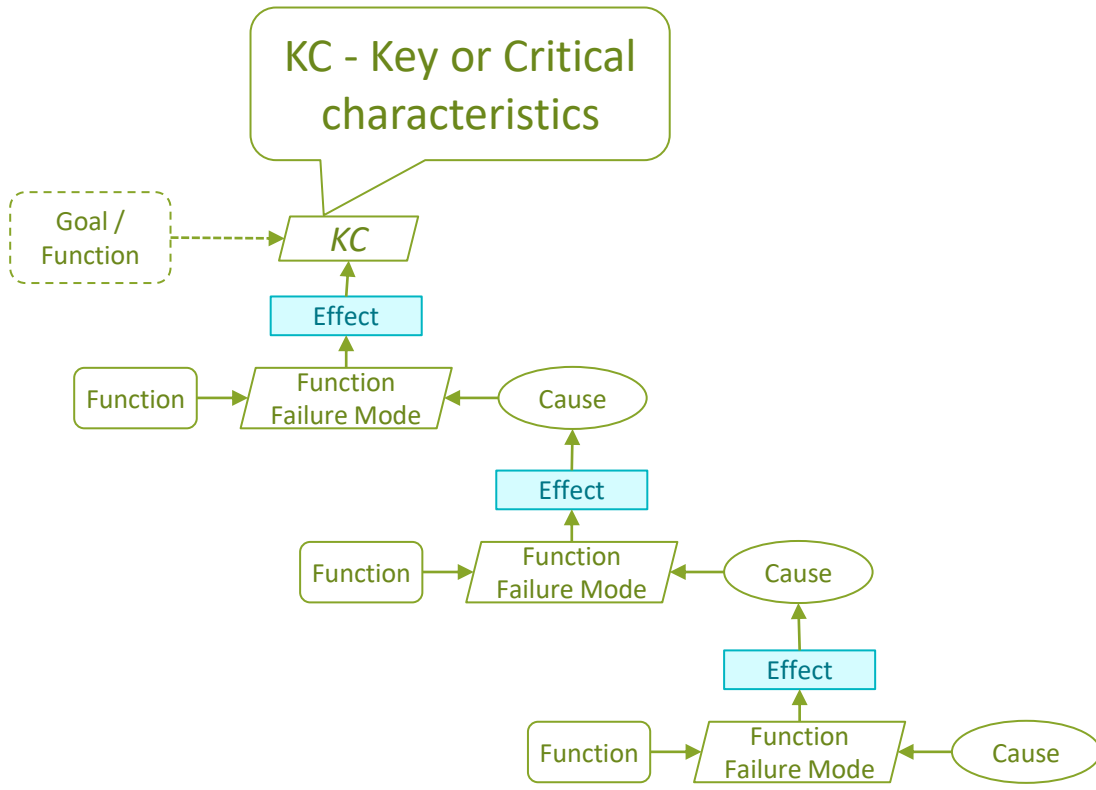
Generated PFMEA

Process Functions / Requirements	Potential Failure Mode	Potential Effect(s) of Failure	Sev	Class	Potential Cause / Mechanism of Failure	Prevention Controls	Occ	Detection Controls
PROCESS: ASSEMBLE DIFFERENTIAL SH004: ASSEMBLE DIFFERENTIAL CORE 004-000: PICK DIFFERENTIAL GEAR FROM PALLET	differential gear not picked	Ultimate customer: differential not assembled			differential gear not located			Monitor location coordinates
		Level up: differential core not assembled			differential gear not secured			Monitor clamping force via electric current
		Next operation: differential gear not placed			differential gear not lifted			Monitor lifting force via electric current
	Incorrect differential gear picked	Ultimate customer: differential damaged			Incorrect differential gear located			Monitor location coordinates
		Level up: differential core damaged						
		Next operation: Damage to the components or equipment						
		Governmental regulations: No effect						
		Associate Safety: No effect						
		Environment: No effect						

Initial technical challenges:

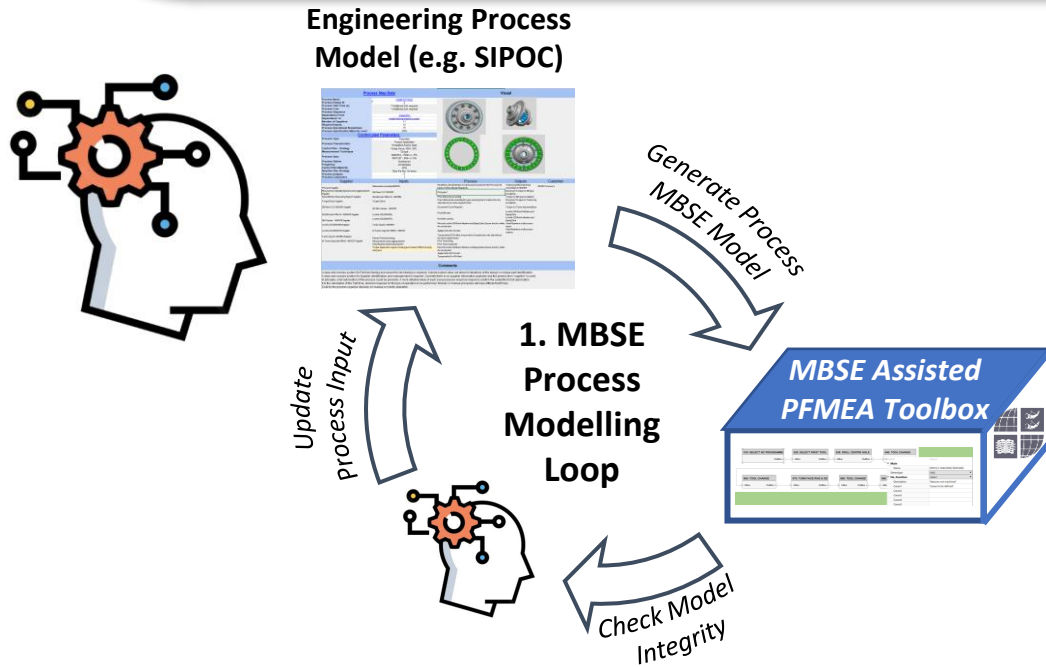
1. Develop an “X-ITL” (Expert / Engineer in the Loop) approach to a comprehensive MBSE for manufacturing process design with integrated FMEA generation;
2. Automate the traceability of key characteristics across the levels of process-as-a-system decomposition

Understanding the Systemic FMEA Reasoning

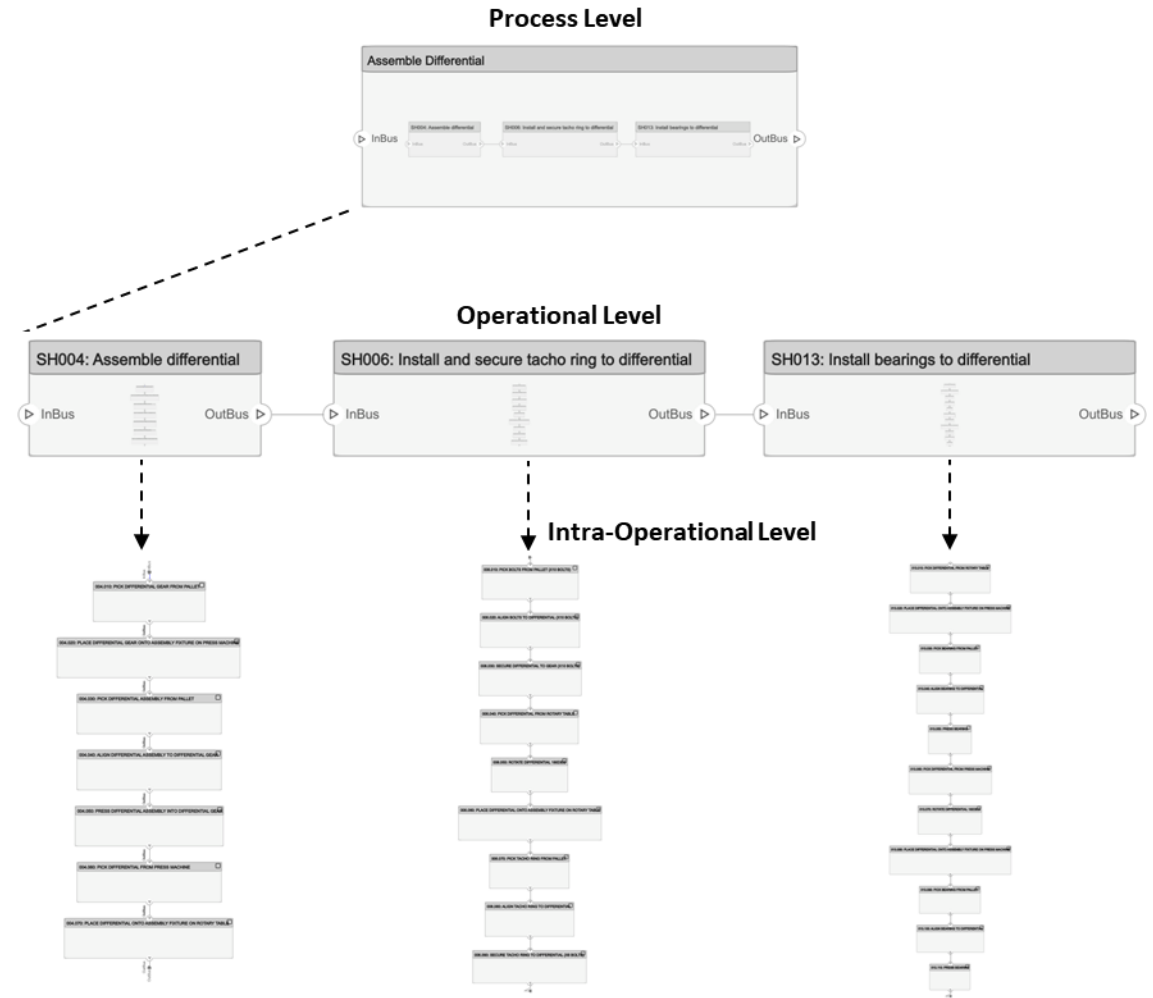


- Research / technical challenges:
 1. Develop an “**X-ITL**” (Expert / Engineer in the Loop) approach to a comprehensive MBSE for manufacturing process design with integrated FMEA generation;
 2. Automate the traceability of key characteristics across the levels of process-as-a-system decomposition

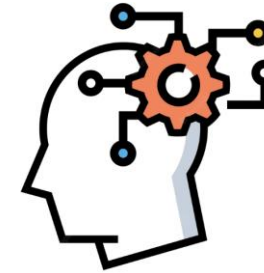
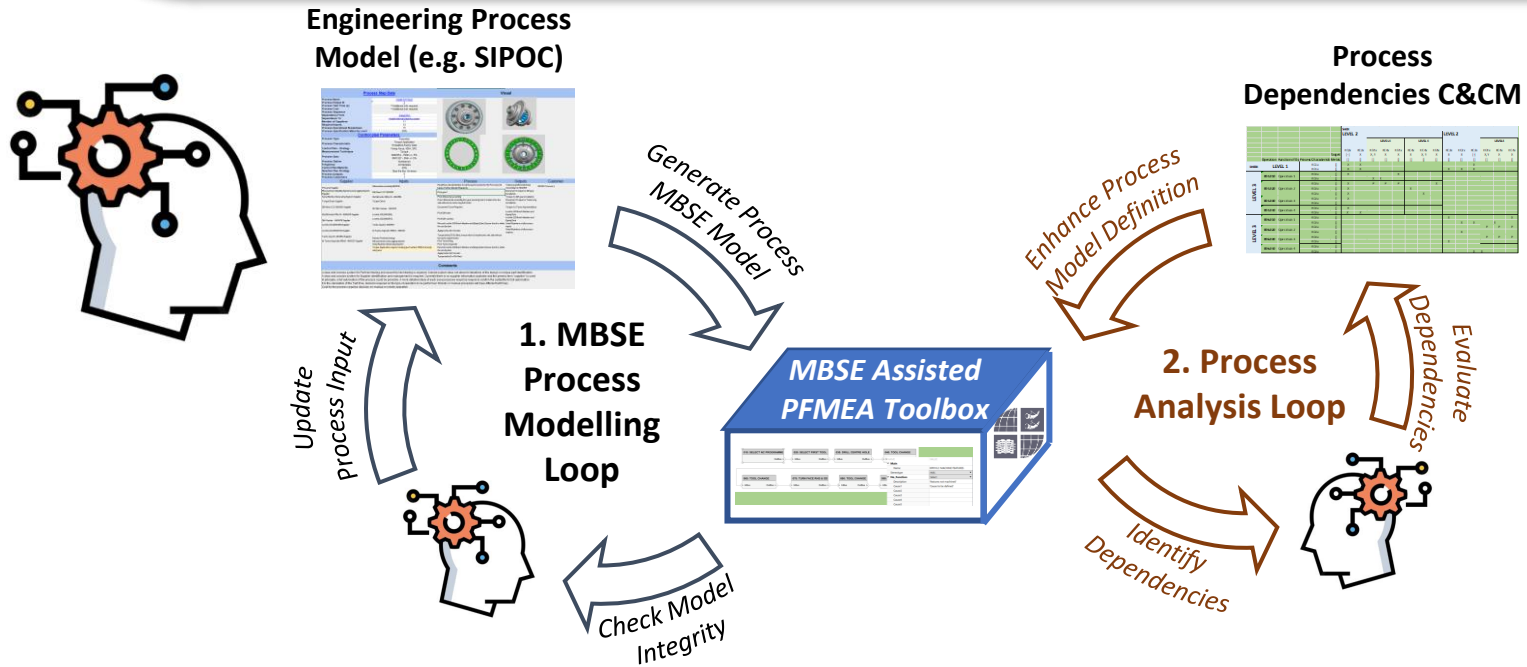
X-ITL MBSE Process Modelling



- The MBSE model is generated automatically from the Engineering Process Model generated by experts;
- If the expert / analyst is not happy with the model s/he will have to adjust the input;
- NB Reverse flow enabled – i.e. Engineering process model (SIPOC, IDEF, etc) can be generated from MBSE



X-ITL Process Analysis



We have developed and implemented the Ford¹ approach based Critical & Correlation Matrix (CCM) to characterise dependencies

- The CCM data structure is quite complex as we need to capture both inter- and intra-level dependencies:
 - Inter-level dependencies – allocation of KCs to next level of decomposition;
 - Intra-level dependencies – propagation of KC flags within the same level where operations have functional dependencies

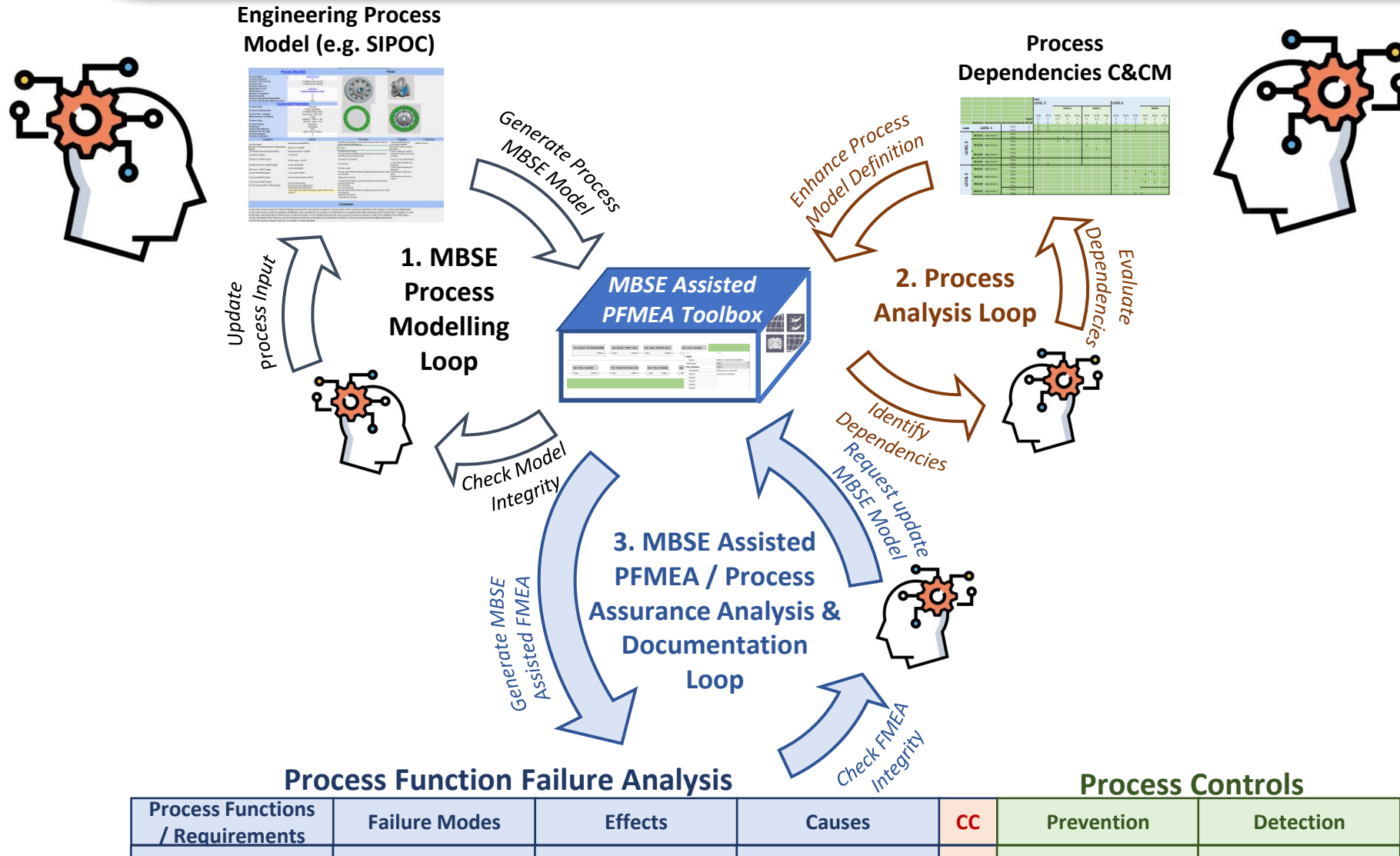
It also means a two-step process for the CCM – and FMEA

This is dependent on the process / solution – hence analysis is concurrent with process design

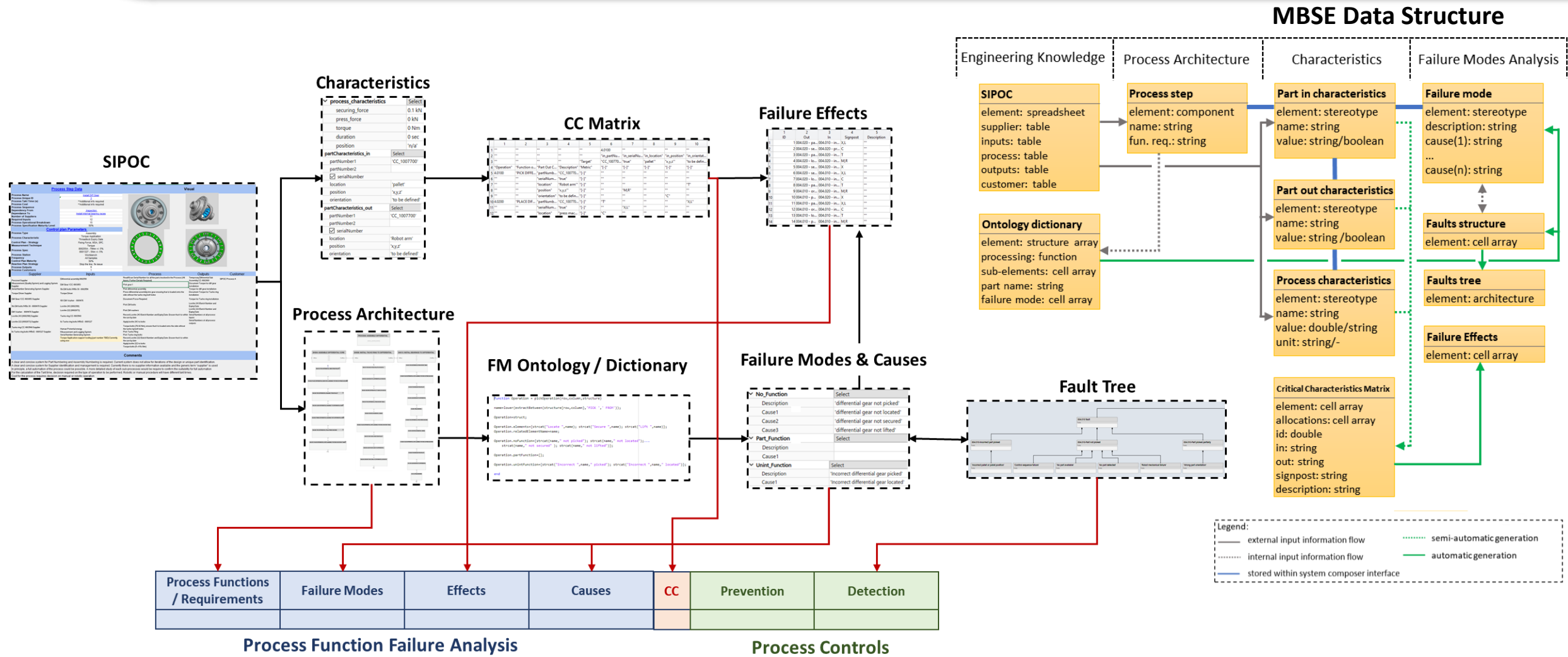
		Key Process Characteristics											
		Operation											
		10.2			10.3			20.1			20.2		
		Operator	Ambient temperature	Operator	Ambient temperature	Number of rivets per basket	Operator	Ambient temperature	Distribution of rivets in tray	Operator	Tray position	Furnace temperature	Time in furnace prior to heat treatment
		[°C]	[°C]	[°C]	[°C]	[#]	[°C]	[°C]	[°C]	[°C]	[°C]	[°C]	[min]
Function of Operation	Part Characteristic	[t]	[°C]	[°C]	[°C]	[°C]	[°C]	[°C]	[°C]	[°C]	[°C]	[°C]	[min]
20.1 Load tray into furnace	Location	Baskets, Tray, Furnace	[t]						X				
	Material	Al Alloy xxxx	[Spec]									R	
	Part	xxxx	[Part Number]	R									
	Number	xxxx	[#]	R									
	Temperature	Rising to Furnace Temp	[°C]					R		X	X	X	X
20.2 Heat rivets	Hardness	X	[BHN]							X	X	X	X
	Tensile Strength	Y	[MPa]							X	X	X	X
	Location	Baskets, Tray, Furnace	[t]						R				
	Material	Al Alloy xxxx	[Spec]										
	Part	xxxx	[Part Number]	R									
	Number	xxxx	[#]	R									
	Temperature	495 ± 5	[°C]			R		R		R	R	R	X
	Hardness	<X	[BHN]							R	R	R	X
	Tensile Strength	>Y	[MPa]							R	R	R	X

¹FMEA with Robustness Linkages, Ford Motor Company, 2004

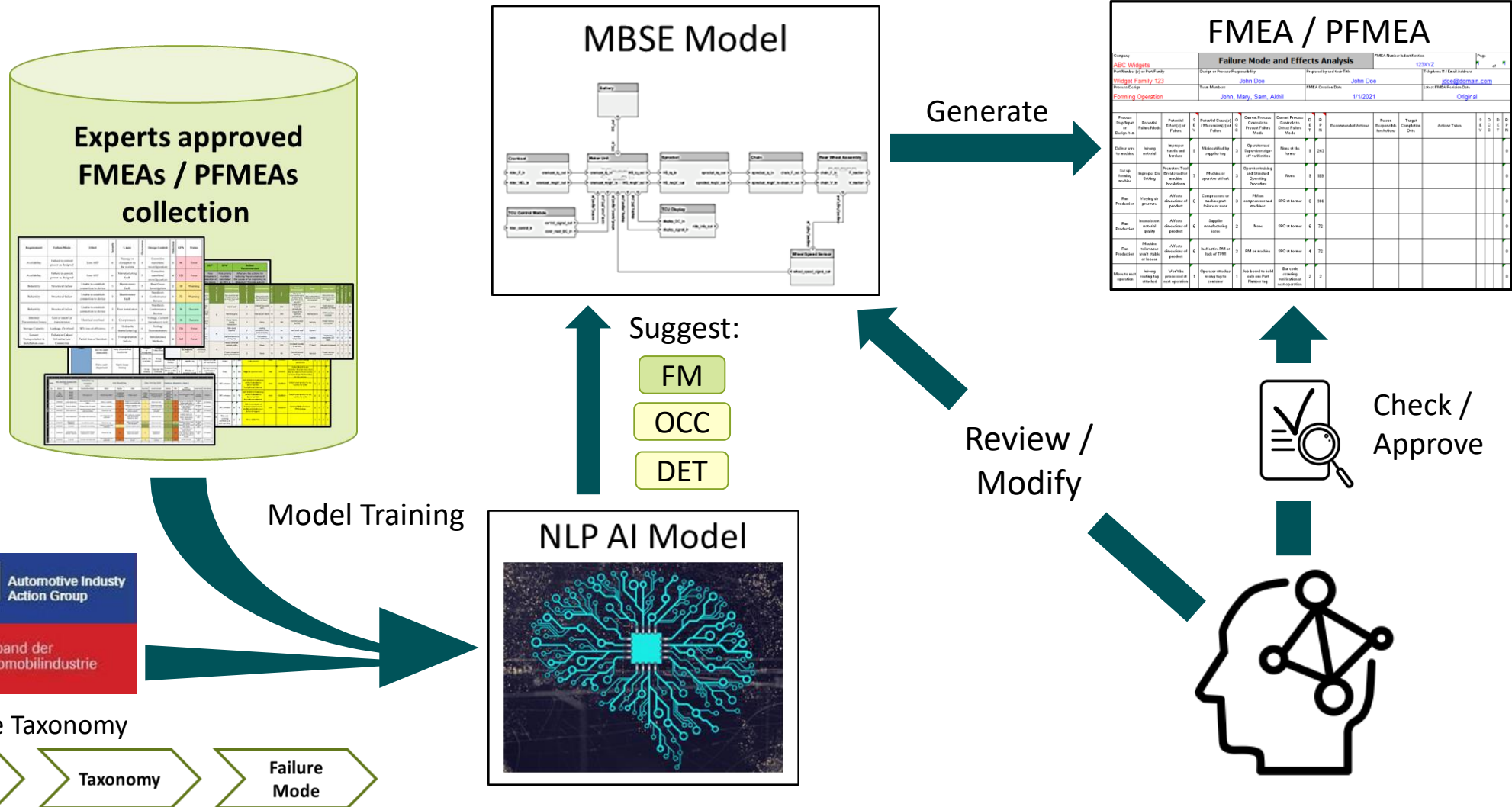
MBSE-assisted FMEA Generation



MBSE-assisted FMEA Generation



Intelligent Failure Modes Dictionary



Failure Mode Taxonomy

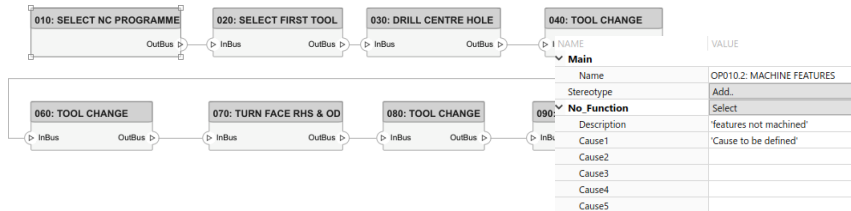


Example

SIPOC



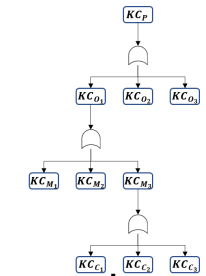
Process Flow MBSE Model



CC Matrix & Characteristics Cascade

CC Matrix & Characteristics Cascade diagram showing a matrix and a cascade diagram.

Operation / Function of Process Characteristics Matrix	LEVEL 2	LEVEL 4	LEVEL 4	LEVEL 2	LEVEL 4
SH004	LEVEL 2	LEVEL 4	LEVEL 4	LEVEL 2	LEVEL 4
SH004.020 Operation 1	LEVEL 2	LEVEL 4	LEVEL 4	LEVEL 2	LEVEL 4
SH004.020 Operation 2	LEVEL 2	LEVEL 4	LEVEL 4	LEVEL 2	LEVEL 4
SH004.020 Operation 3	LEVEL 2	LEVEL 4	LEVEL 4	LEVEL 2	LEVEL 4
SH004.020 Operation 4	LEVEL 2	LEVEL 4	LEVEL 4	LEVEL 2	LEVEL 4
SH004.020 Operation 5	LEVEL 2	LEVEL 4	LEVEL 4	LEVEL 2	LEVEL 4
SH004.020 Operation 6	LEVEL 2	LEVEL 4	LEVEL 4	LEVEL 2	LEVEL 4
SH004.020 Operation 7	LEVEL 2	LEVEL 4	LEVEL 4	LEVEL 2	LEVEL 4
SH004.020 Operation 8	LEVEL 2	LEVEL 4	LEVEL 4	LEVEL 2	LEVEL 4
SH004.020 Operation 9	LEVEL 2	LEVEL 4	LEVEL 4	LEVEL 2	LEVEL 4
SH004.020 Operation 10	LEVEL 2	LEVEL 4	LEVEL 4	LEVEL 2	LEVEL 4

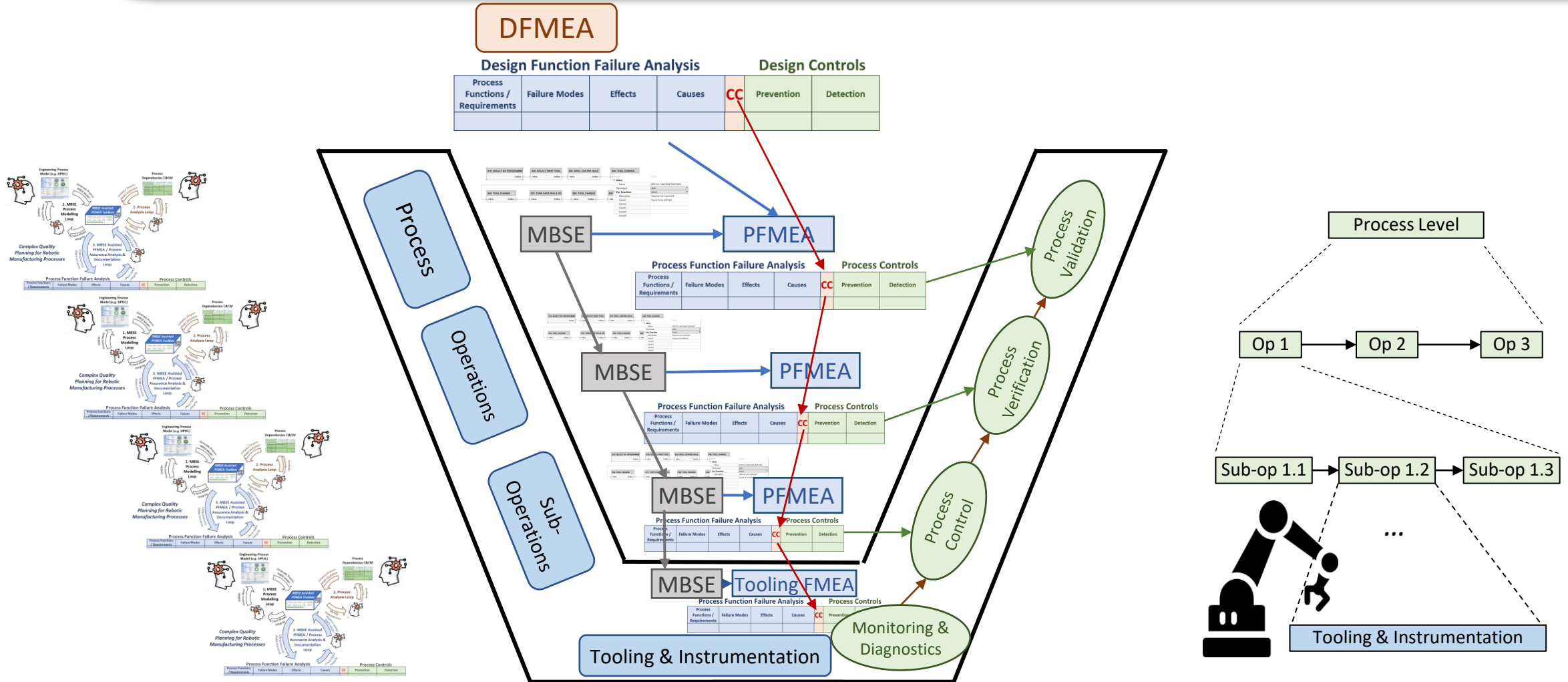


Effects of failure mode are automatically suggested from the C&CM

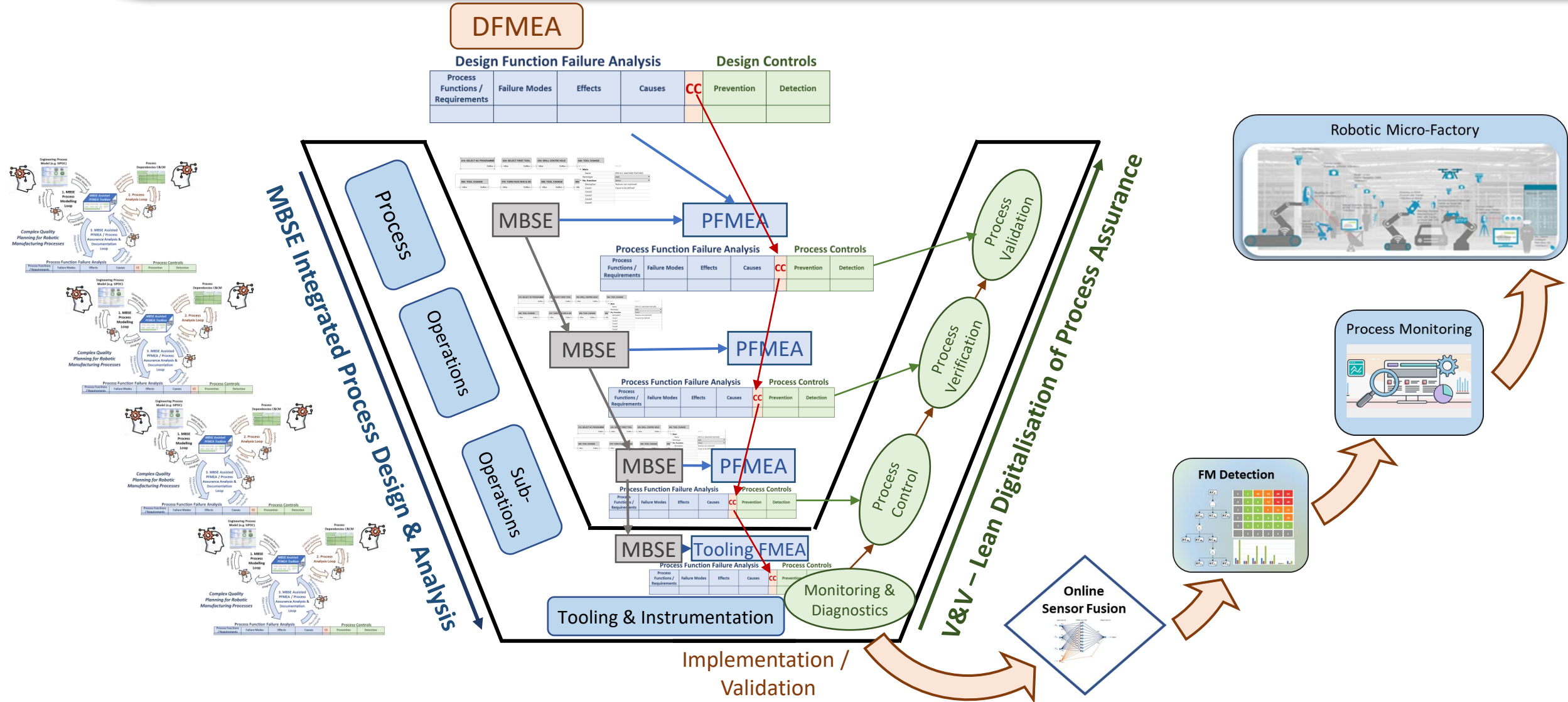
Severity automatically cascaded based on the traceability to DFMEA

Process Functions / Requirements	Potential Failure Mode	Potential Effect(s) of Failure	Sev	Class	Prevention Controls	Occ	Detection Controls
PROCESS: ASSEMBLE DIFFERENTIAL CORE SH004: ASSEMBLE DIFFERENTIAL CORE 004.010: PICK DIFFERENTIAL GEAR FROM PALLET	differential gear not picked	Ultimate customer: effect on operation SH004, Part Geometry (5)	5		differential gear not located		Monitor Location via Location Coordinates
		Level Up: effect on operation SH004, Part Geometry (5)			differential gear not secured		Monitor Clamping Force via Electric Current
		Next operation: effect on operation 004.020, Path Availability (5)			differential gear not lifted		Monitor Lifting Force via Electric Current
	Incorrect differential gear picked	Ultimate customer: effect on operation SH004, Part Geometry (5)	5		Incorrect differential gear located		Monitor Location via Location Coordinates

Integrated MBSE Robofactory Process Design



Integrated MBSE Robofactory Process Design

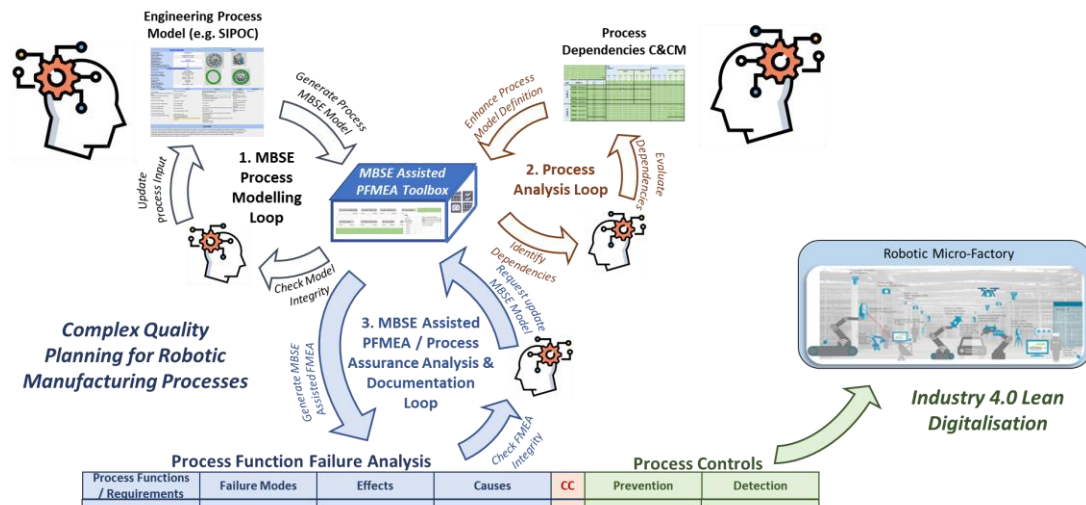


Discussion & Conclusions

- Successful proof of concept for the next generation design tools that deliver amplified intelligence through interconnected AI & Human Intelligence – i.e. “Expert in the Loop”
 - Current state: alpha prototype developed and validated with a subsystem of the EDU robotic manufacturing process
 - » System Design / DFMEA in progress
- Current work:
 - Develop an AI ontology for function failure representation, with an extended taxonomy for failure modes to support integrated MBSE-assisted FMEAs for intelligent physical systems – equipped with embedded AI
 - Develop intelligent data-driven FMEAs – i.e. automatic build of the MBSE-assisted FMEA from real world failure data (e.g. warranty data)

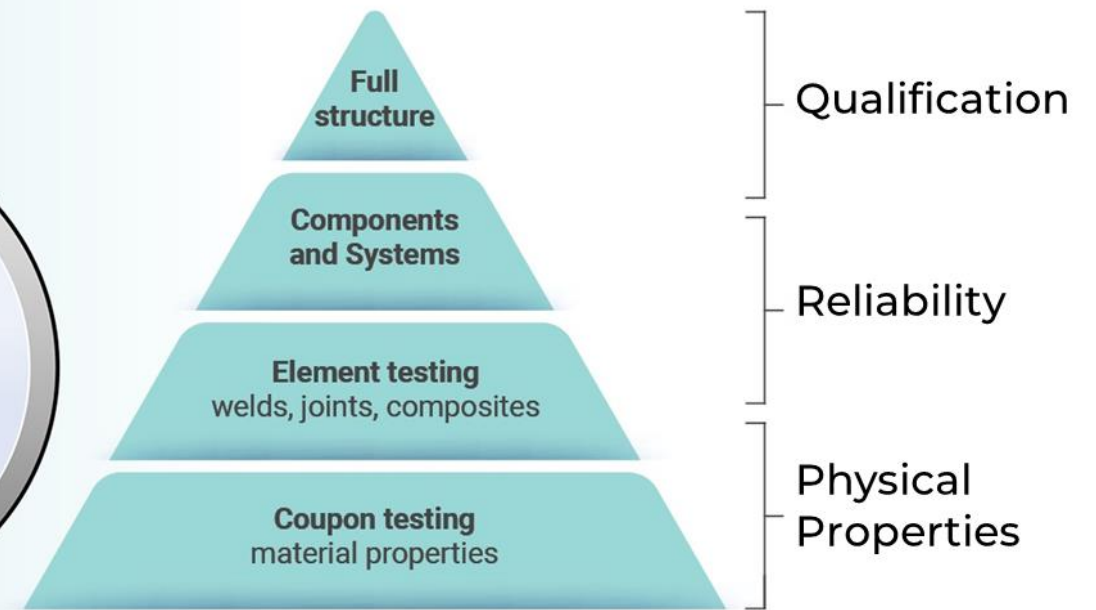
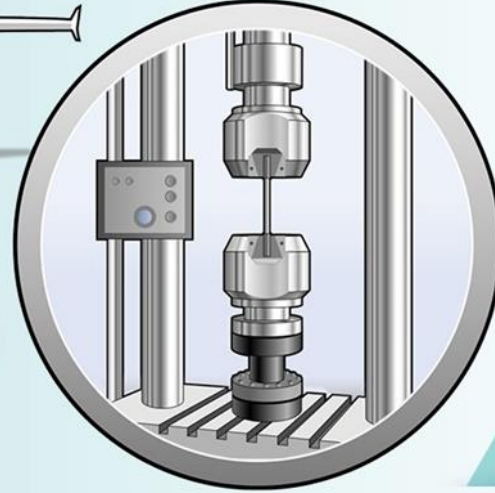
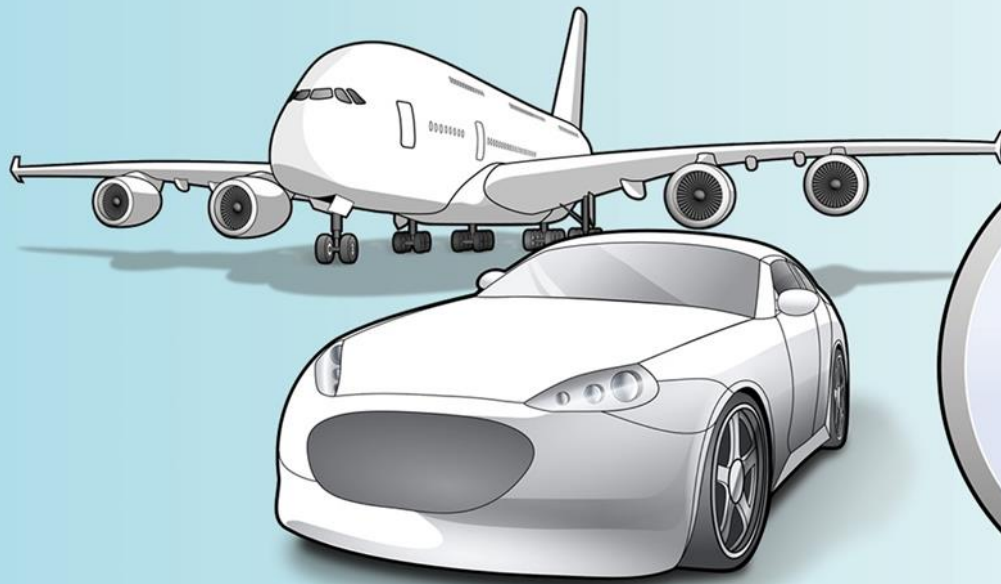


Placing the Expert in-the-Loop: Framework for MBSE-assisted Automatic FMEA Generation



Thank You!

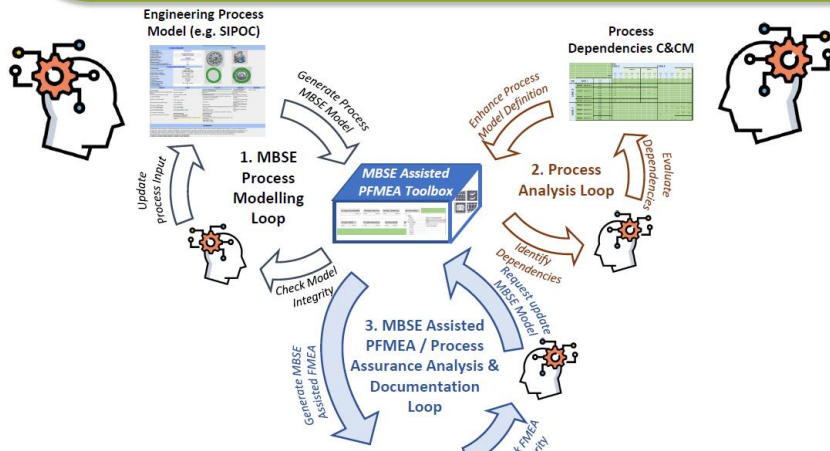
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- Placing the Expert in-the-Loop: Framework for MBSE-assisted FMEA Generation

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Process Function Failure Analysis				Process Controls		
Process Functions / Requirements	Failure Modes	Effects	Causes	CC	Prevention	Detection