# MathWorks **AUTOMOTIVE CONFERENCE 2024** India

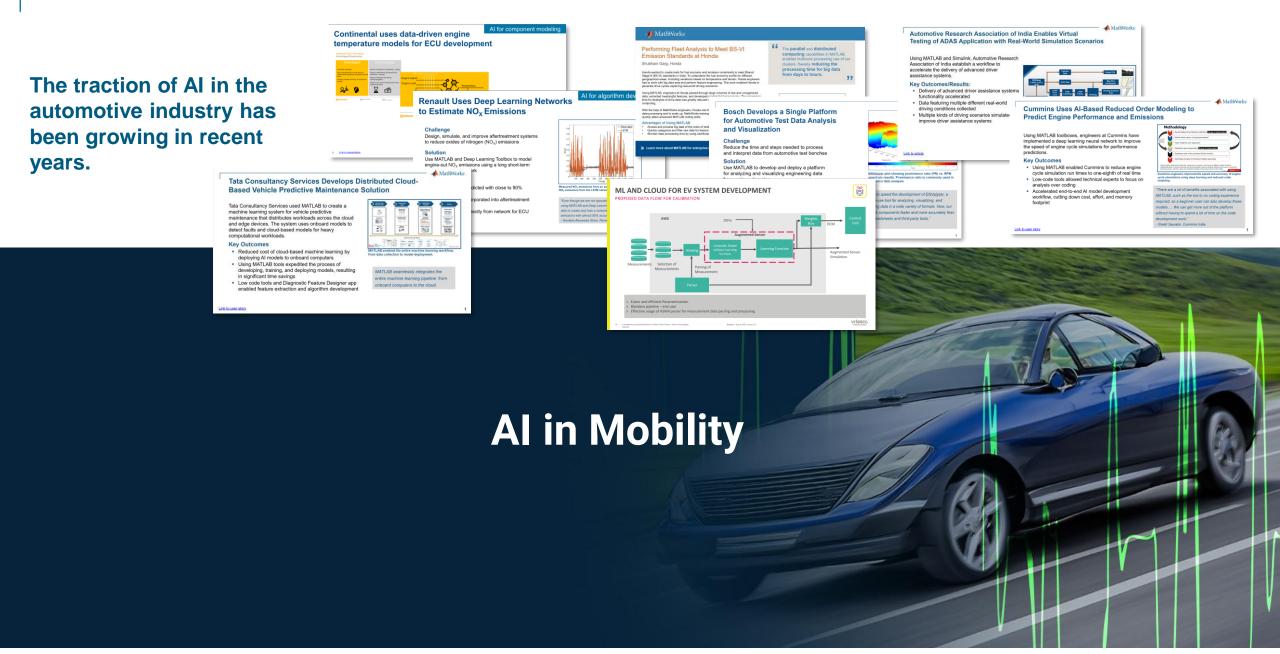
19 November | Pune

# Accelerating Al Adoption: From Design to Deployment in Mobility



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#### MathWorks AUTOMOTIVE CONFERENCE 2024





Core Technology Foundation







**Specialized Talents** 

# AI in Mobility



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Senior Application Engineer, MathWorks



## Koustubh Shirke

Senior Application Engineer, MathWorks



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Senior AI Academic Liaison, MathWorks

## Integrating AI is a priority for companies today but...

#### Top Barriers to Al Implementation

Unable/Hard to Measure the Value	119	6	30%
Complexity of Al Solution(s) Integration With	119	6	309
Data Volume and/or Complexity	8%	22%	
Potential Risks or Liabilities	7%	22%	
Data Scope or Quality Problems	6%	20%	
Lack of Understanding AI Benefits and Uses	6%	20%	
Lack of Technology Knowledge	5%	19%	
Data Accessibility Challenges	6%	19%	
ttle Improvement Over Existing Technologies	6%	18%	
Lack of Skills of Staff	7%	18%	
Technology Is Too Difficult to Use or Deploy	5%	17%	
Governance Issues or Concerns	5%	17%	
Lack of Capability to Leverage AI Techniques	6%	16%	
Difficulty Finding Use Cases	5%	15%	
Unable/Hard to Measure the Value	3%	14%	
0%		15%	30%
= 601 All Respondents, excluding "not sure." What are the top 3 barriers to the implementation of AI technique urce: 2020 Gartner AI In Organizations Survey	as within your organization	7	
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#### n = 601

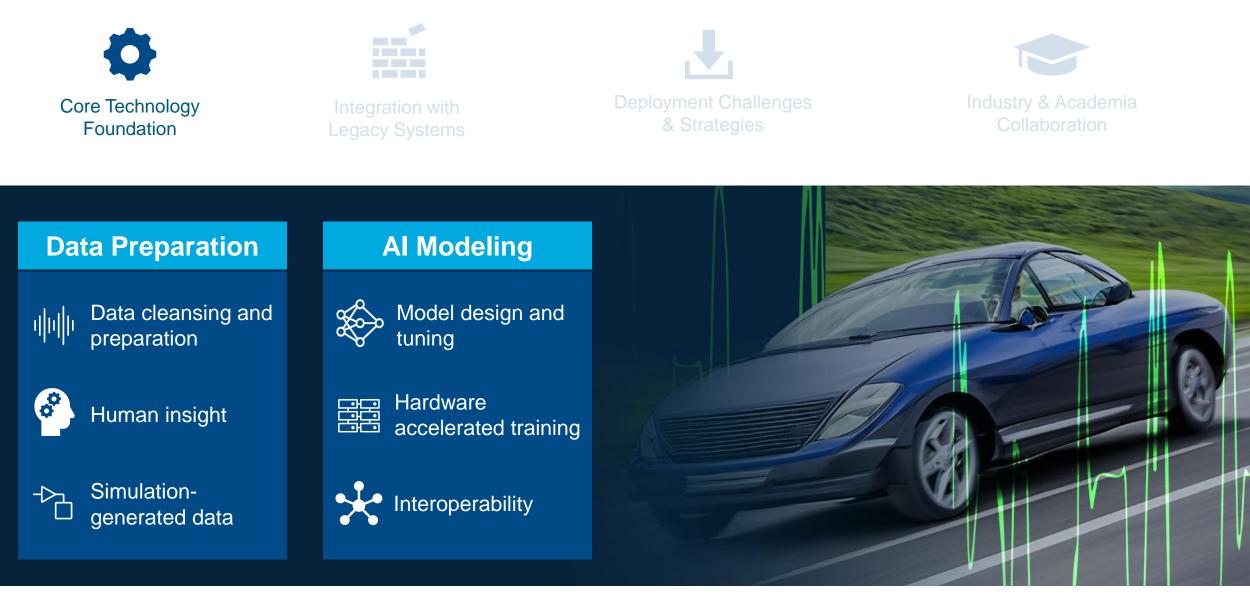
Gartner Research Circle members, excluding "unsure" Source: 2020 Gartner AI in Organizations Survey Q: What are the top three barrier to the implementation of AI techniques within your organization? *Rank up to three.* ID: 719012 C

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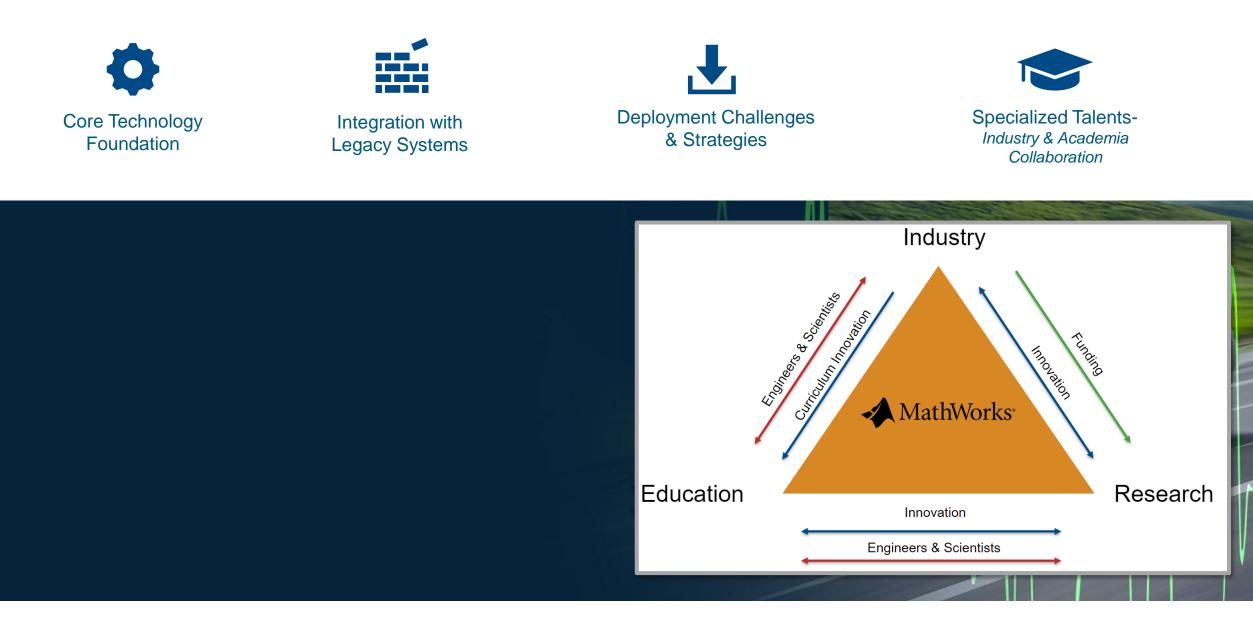
- Source: "" How to Build Knowledge Graphs That Enable AI-Driven
- Enterprise Applications" Gartner Research Note, <D#>, published 27 May 2020

# Top barriers to successful adoption of AI

- 1. Integration with existing technology
- 2. Data Complexity/Quality
- 3. Lack of Skills



Core Technology Foundation	Integration with Legacy Systems	Deployment Challenges & Strategies	Industry & Academia Collaboration
Data Preparation	Al Modeling	Simulation & Test	Deployment
االالالالالالالالالالالالالالالالالالا	Model design and tuning	Integration with complex systems	Embedded devices
Human insight	대대 Hardware 고급 accelerated training	System simulation	Enterprise systems
-D Simulation- generated data	Interoperability	<ul> <li>× System verification</li> <li> and validation</li> </ul>	Edge, cloud, desktop



## Typical Data Driven Workflow

**Domain Experts** 

From data to deployment, overseeing

the entire AI workflow in MATLAB

From data preparation to deployment



#### **Embedded Software Engineers**

Implementing and optimizing AI models on edge hardware, collaborating seamlessly with data science teams

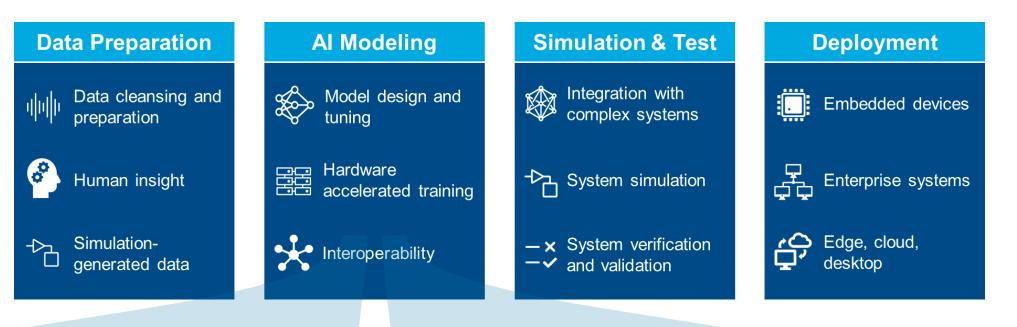
# Data Preparation





Deployment

## Leveraging AI-driven Automotive Application design



## AI for component modeling

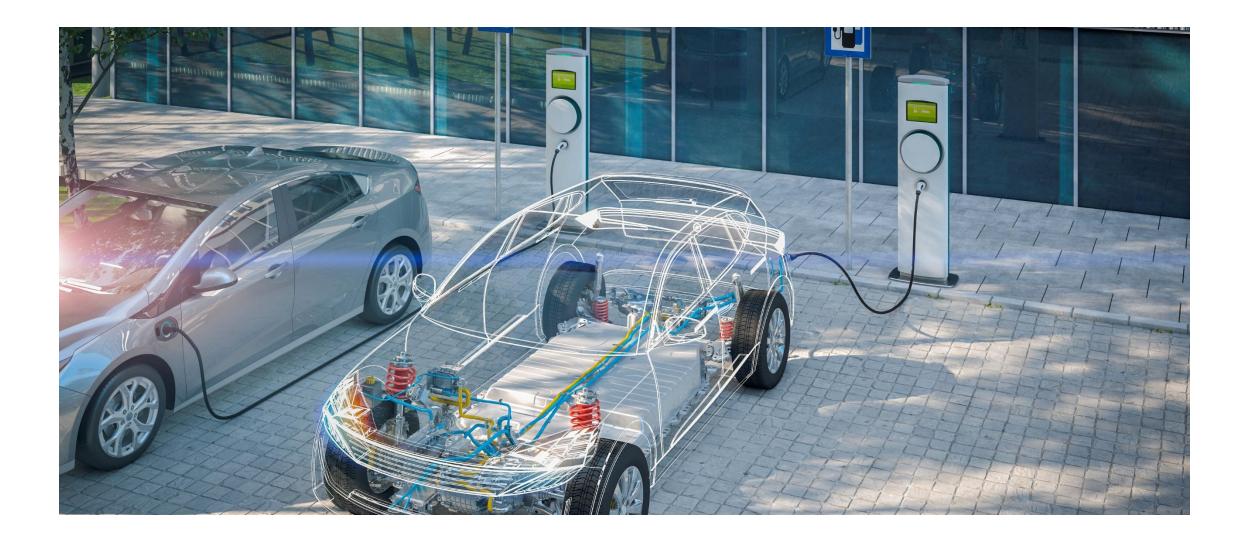
- Speeding up desktop and HIL simulations
- Modeling component dynamics from data when first-principles models cannot be obtained

## Al for Algorithm development

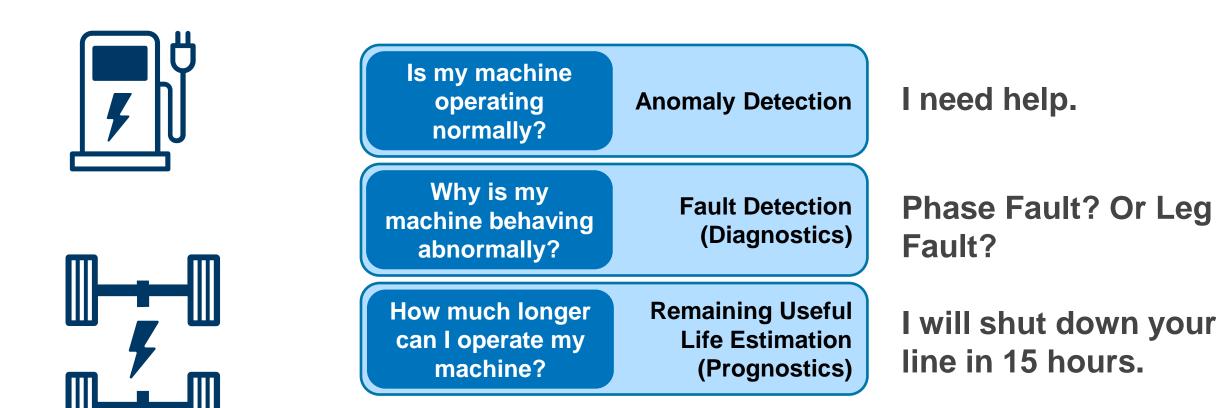
- Virtual sensor modeling
- Sensor fusion
- Object detection
- Remaining Useful Life Estimation/Predictive Maintenance

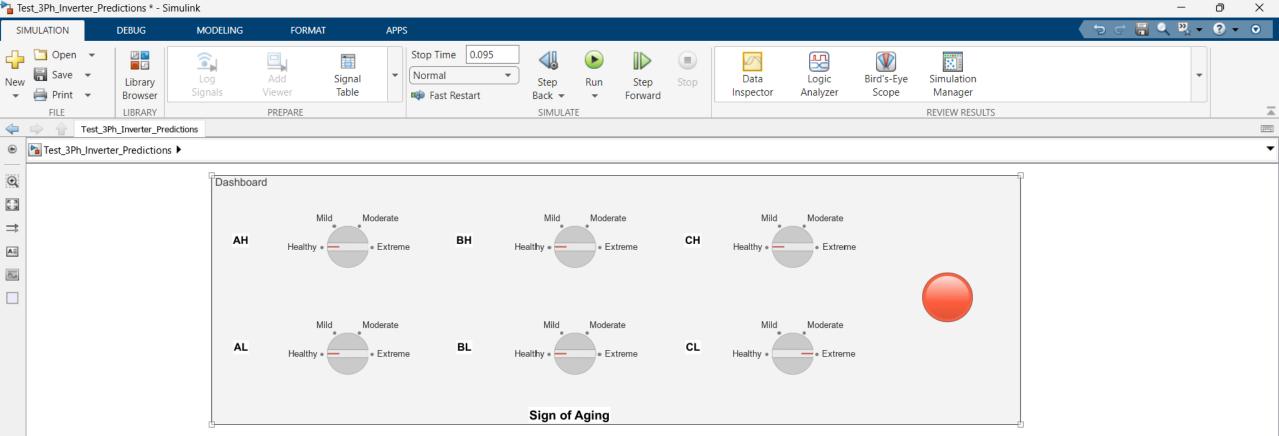
Leveraging Data Driven Application Software Development

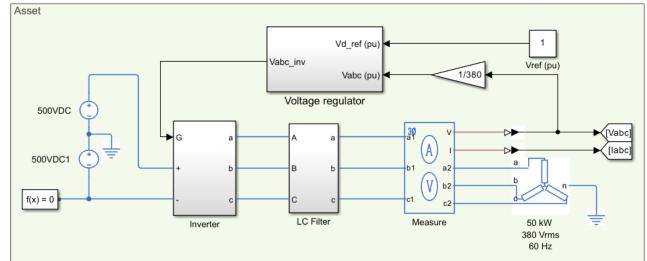
## Power Converters: Driving the Future of Electrification

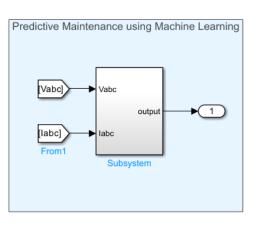


## **Predictive Maintenance of Power Converters**









Three-Phase Inverter Voltage Control

## Daihatsu Uses AI to Classify Engine Sounds

## Challenge

Develop an AI solution that can judge the level of engine knocking sound, which only skilled workers could judge

## **Solution**

Create classification models and easy-to-use interface with MATLAB, making it possible to examine features multiple times

## **Key Outcomes**

- Performed knocking sound analysis with the same accuracy as skilled workers
- Increased AI expertise through MATLAB training
- Promoted visualization of AI and increased awareness of AI



Daihatsu used AI to identify knocking sounds from its engines.

"Although we tried other programming languages, it was hard to implement. We decided to use MATLAB, which allows us to easily import the necessary data by dragging and dropping, and we could easily see the result by ourselves."

- Takuya Kumagae, Daihatsu Motor Co., Ltd.

## Challenges in AI Model Development

Accuracy	Balancing precision with real-world applicability
Development Time	Time to market
Effective Testing	Ensuring reliable model behavior
Compatibility	<ul> <li>Integration with the production systems</li> </ul>
Maintainability	Efficient model updating
Robustness	Maintaining consistent performance in diverse situation
Latency	Meeting real-time processing

# Hands-on learning: IIT Madras and HAW Hamburg



**Dr. V Krishna Teja Mantripragada** (He/Him) • 1st Tyre & Vehicle dynamics, JK Tyres | Adjunct Faculty, IIT Madras 2w • Edited • 🔇

#### Hey Automotive Enthusiasts!

Ever wondered about so many buzz words in suspension tuning that makes your vehicle ride and handling smooth...

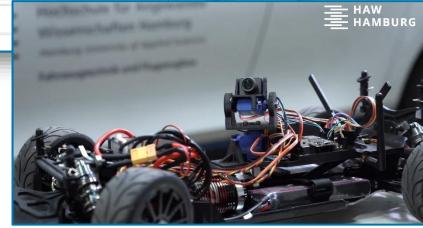
Here is a simple <u>#MATLAB</u> app that I created during my university days for quickly visualising McPherson suspension and calculating some important metrics. The app can be downloaded from Mathworks FileExchange

#### https://lnkd.in/gNfudcN6

Feel free to customise as needed! Share your insights on suspension optimization and tuning experiences...

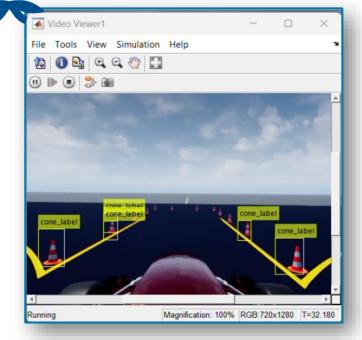
#AutomotiveEngineering #vehicledynamics #engineering #design #vehicledesign #students #formulastudent #modelling #simulation

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#### Student competitions – <u>TUM at the Indy</u> <u>Autonomous Challenge</u>

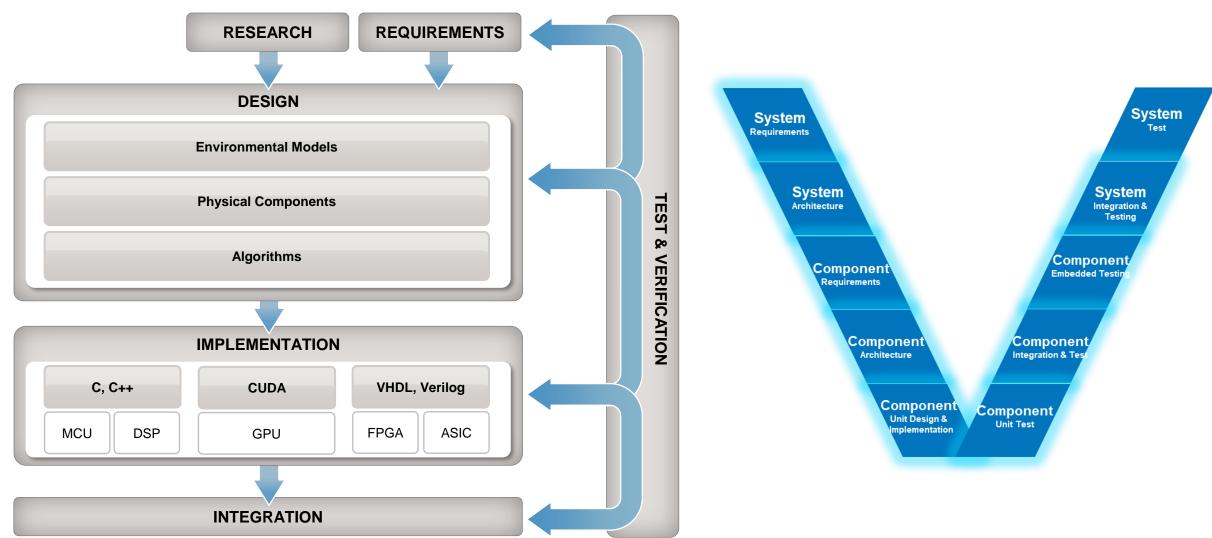




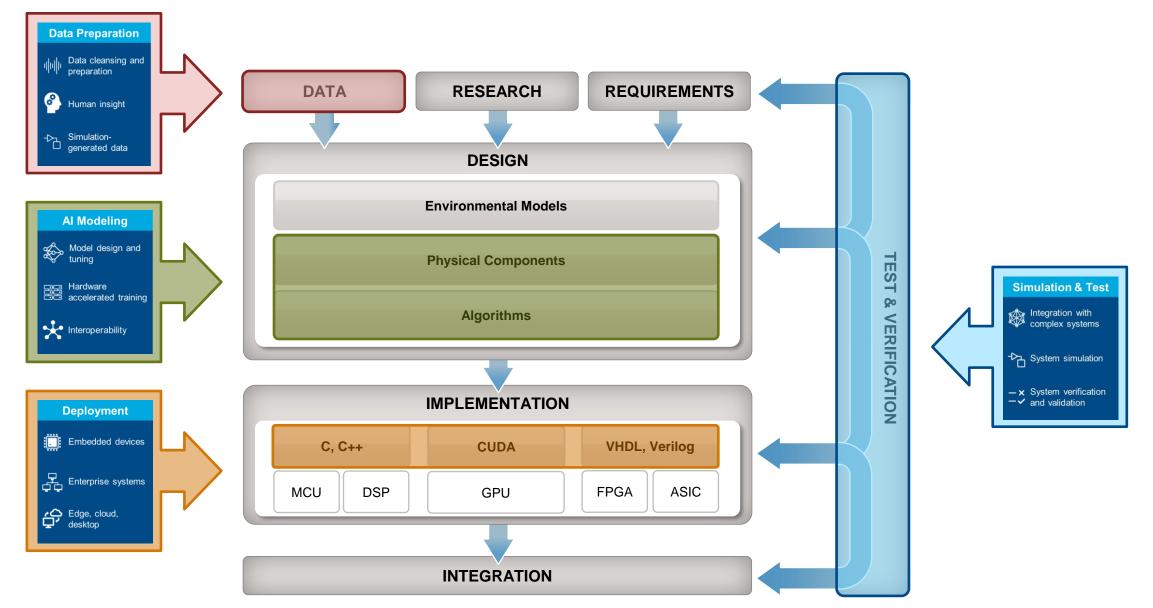


# **Audience Question**

## Application Development workflow with Model-Based Design



## Integrating AI in Model-Based Design



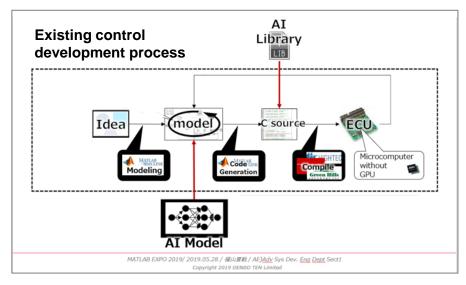
## Use AI to improve engine control unit development efficiency

# **DENSO TEN**

- Used Deep Learning to formulate a model for complex vehicle control issues
- Applied model-based development to integrate AI model into existing control model
- Developed a pathway to c-code generation for ECU implementation

"A model-based development workflow is essential in order to use AI for control ECUs. Combining the existing control model and the AI model enables us to establish a simulation environment and accelerate product development." - Natsuki Yokoyama, Denso Ten

#### **Model-Based Development Workflow**



## Verification and Validation of AI Model

Model-Based Design	Al Models
Ensuring algorithms represent the intended physical or logical system accurately	Verifying model training aligns with intended objectives. Explainability Rigor & Trust
Errors stem from incorrect logic, parameter values, or design mistake.	Explainability Rigor & Hust Er ors may arise from data quality hyperparameters, or learning process
Ensure the system performs as intended under real-world scenarios	Ensure the AI model generalizes well to unseen data and meets real world recomments Robustness
Relies on diverse and representative datasets for validation	Uses predefined scenarios design requirements and physical constraints
	Ensuring algorithms represent the intended physical or logical system accurately Errors stem from incorrect logic, parameter values, or design mistake. Ensure the system performs as intended under real-world scenarios Relies on diverse and representative

## Developing Onboard SOH Estimation Using DVA and ICA for LFP Batteries

#### Challenge

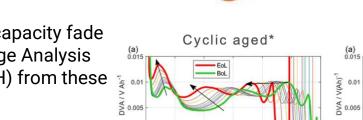
Li-on Batteries suffer from a variety of degradation mechanisms that lead to either capacity fade or power fade. Techniques like Incremental Capacity Analysis and Differential Voltage Analysis can be used to estimate DQ and DV curves but inferring battery State of Health (SOH) from these curves still requires domain expertise.

#### **Solution**

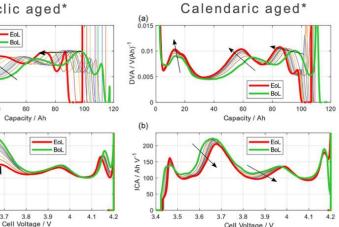
Gotion used MATLAB to develop feature extraction methods that detect the most important features in DQ and DV curves, then trained a linear regression model that correlates these features with capacity fade. This regression model was then used alongside temperature data in a 2-D look-up table that estimates SOH. The solution was implemented in Simulink for testing, requirements validation, and certification.

### **Benefits of using MATLAB and Simulink**

- Easy data analysis for visualization and identification of key trends in battery aging
- Built-in tools to extract meaningful features from differential voltage curves (peak detection)
- V-diagram workflow support including requirements management, automatic code generation, and ISO 26262/IEC 61508 certification



(b)



Gotion

Link to MathWorks Automotive Conference slides Link to MathWorks Automotive Conference recording

## There is an increased focus in government **regulation and certification** efforts



## Automotive



### ISO/CD PAS 8800

Road Vehicles — Safety and artificial intelligence

**Under development** A draft is being reviewed by the committee.



#### Aerospace



#### Process Standard for Development and Certification/Approval of Aeronautical Safety-Related Products Implementing AI ARP6983

This document discusses guidelines for the development of Aircraft Systems leveraging Al capabilities, taking into account the overall aircraft operating environment and functions. This includes validation of requirements and verification of the design implementation for certification and product assurance and guidelines with the assessment of safety. It provides practices for showing compliance with the regulations and serves to assist a company in developing and meeting its own internal standards by considering the guidelines herein.

-/V-Medical Devices

← <u>Software as a Medical Device (SaMD)</u>

#### Artificial Intelligence and Machine Learning in Software as a Medical Device

May 13, 2024 update: 191 Artificial Intelligence and Machine Learning (AI/ML)-Enabled Medical Devices were added to the list below. With this update, the FDA has authorized 882 AI/ML-enabled medical devices. Of those newly added to the list, 151 are devices with final decision dates between August 1, 2023, and March 31, 2024, and 40 are devices from prior periods identified through a further refinement of methods used to generate this list.

**European Parliament** 

WIP 2023-06-26

2019-2024



TEXTS ADOPTED

#### P9\_TA(2024)0138

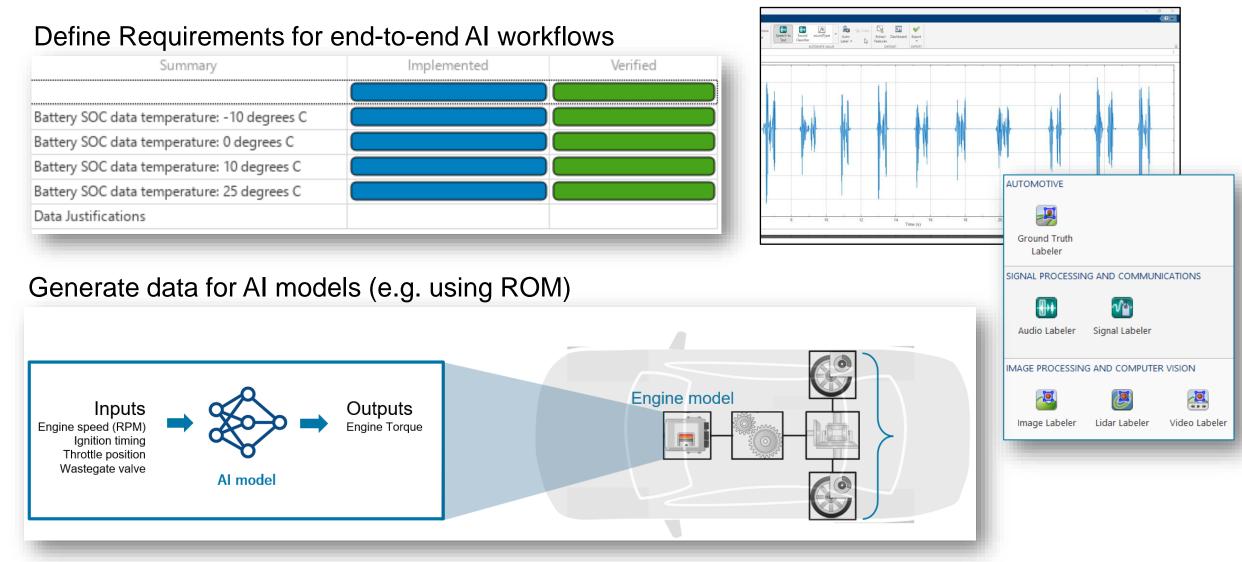
#### **Artificial Intelligence Act**

European Parliament legislative resolution of 13 March 2024 on the proposal for a regulation of the European Parliament and of the Council on laying down harmonised rules on Artificial Intelligence (Artificial Intelligence Act) and amending certain Union Legislative Acts (COM(2021)0206 - C9-0146/2021 - 2021/0106(COD))



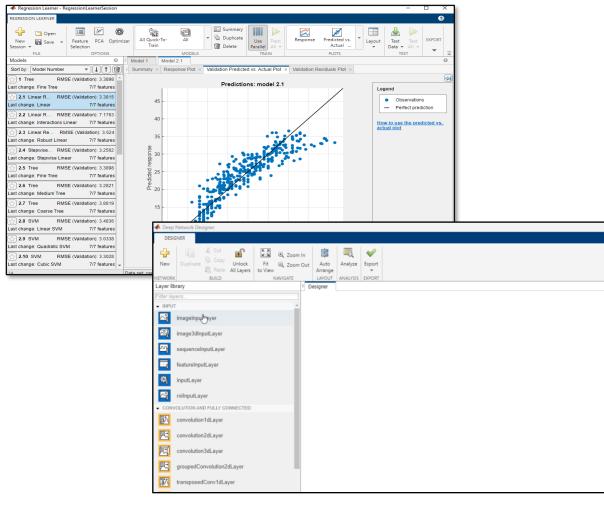
## Requirements and Data Management

### Automate Data Labeling



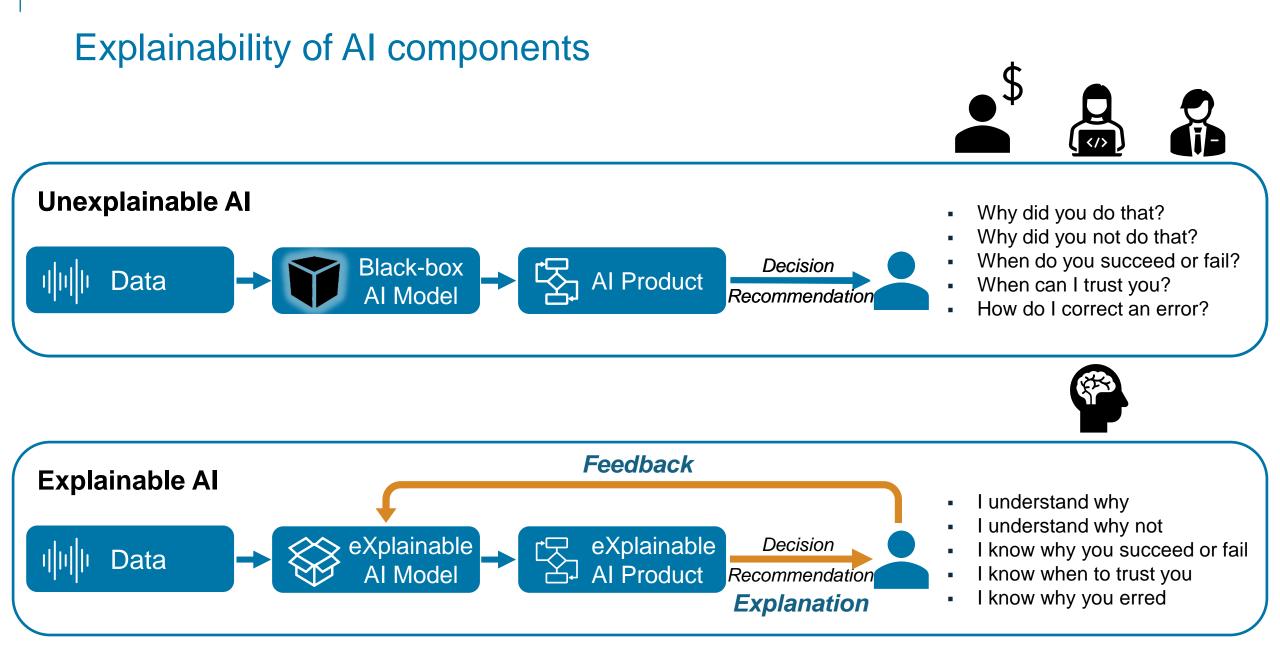
## Learning Process - AI Modeling, Hyperparameters & Experimentation

# Design & Train Deep Learning, Machine Learning models



#### Experiment over models, hyperparameters

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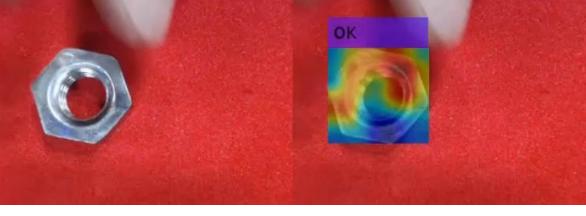
# Musashi Seimitsu Industry Uses Deep Learning for Visual Inspection of Automotive Parts



## How XAI was used: Estimate and visualize the defect area using Class Activation Mapping

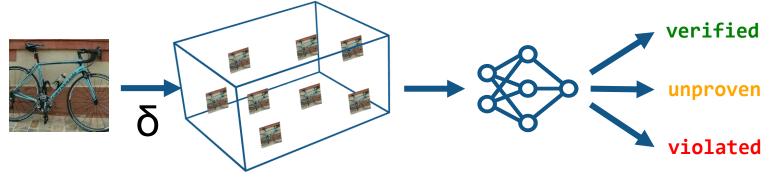
Using camera connection, preprocessing, and various pretrained models in MATLAB enabled us to work on the entire workflow. Through discussions with consultants, our team gained many tips for solving problems, growing the skills of our engineers.

#### **Class Activation Mapping**

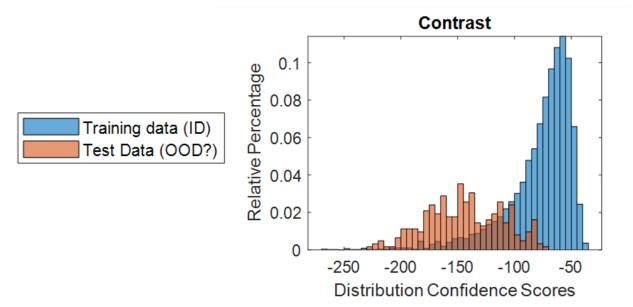


## **Robustness and Out of Distribution Detection**

Is the AI model robust against small input changes?



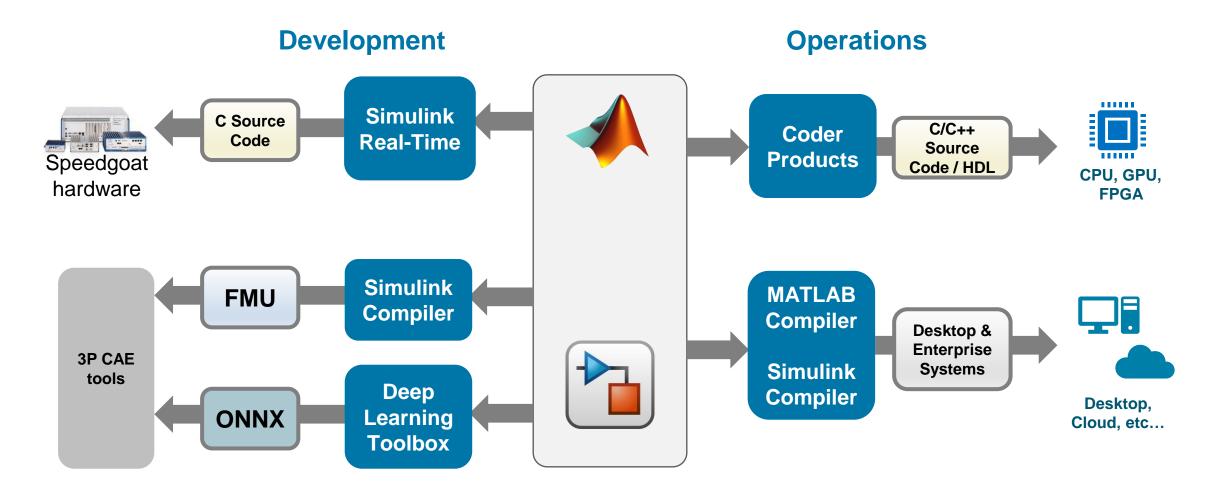
Can the AI model reliably detect unknown samples?





**Deep Learning Toolbox Verification Library** by MathWorks Deep Learning Toolbox Team **STAFF** Verify and test robustness of deep learning networks

# Deploy to your choice of embedded hardware, or integrate with a variety of platforms in the cloud





## **Operationalize Embedded AI with On-Device Learning**



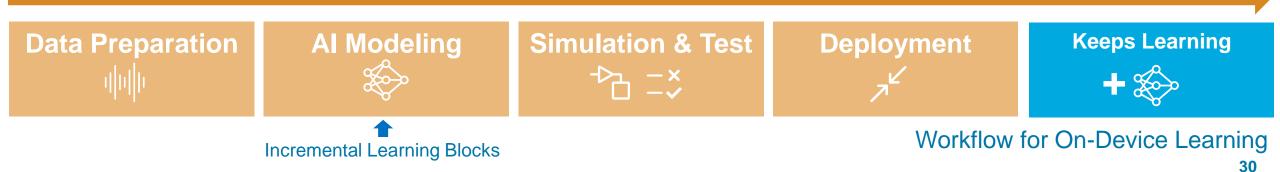
Domain Experts From data to deployment, overseeing the entire AI workflow in MATLAB



#### **Embedded Software Engineers**

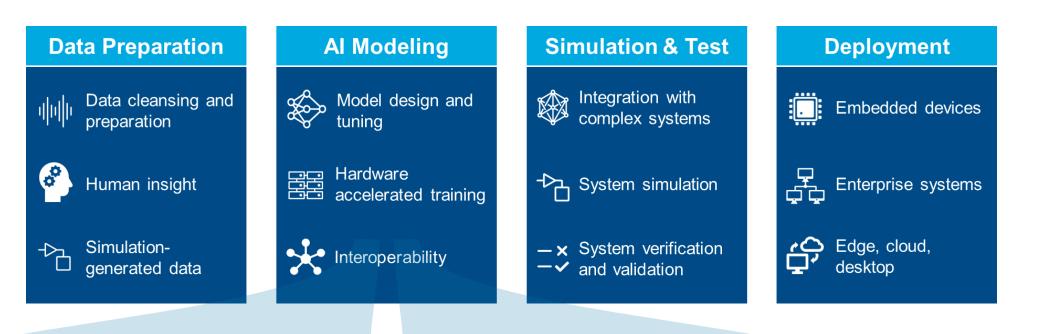
Implementing and optimizing AI models on edge hardware, collaborating seamlessly with data science teams

Workflow for Embedded AI





## Leveraging Al-driven Automotive Application design



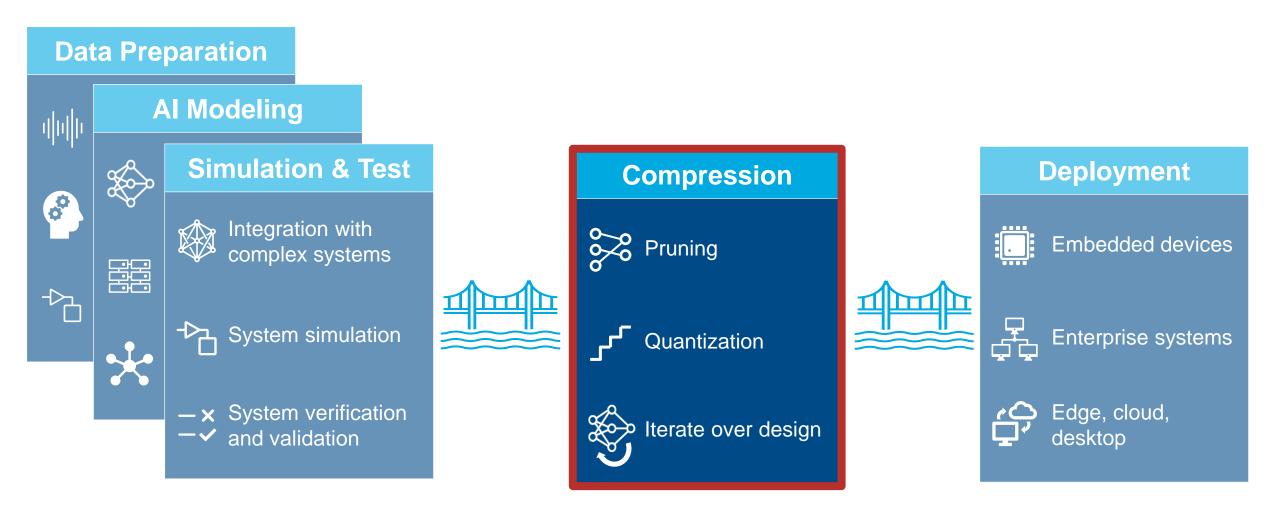
## AI for component modeling

- Speeding up desktop and HIL simulations
- Modeling component dynamics from data when first-principles models cannot be obtained

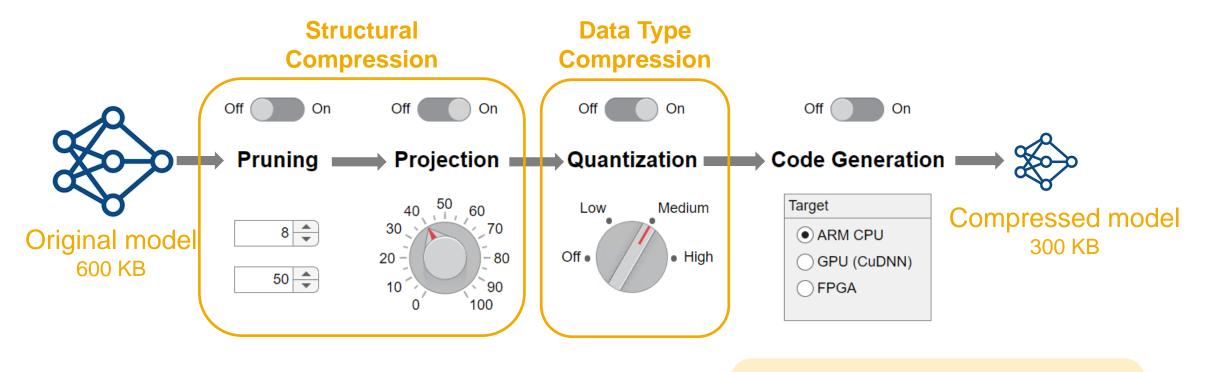
## Al for Algorithm development

- Virtual sensor modeling
- Sensor fusion
- Object detection
- Remaining Useful Life Estimation/Predictive Maintenance

Leveraging Data Driven Application Software Development Model compression can bridge the gap between AI modelling and embedded deployment



# Reduce model footprint and accelerate inference of DL models for deployment to the edge



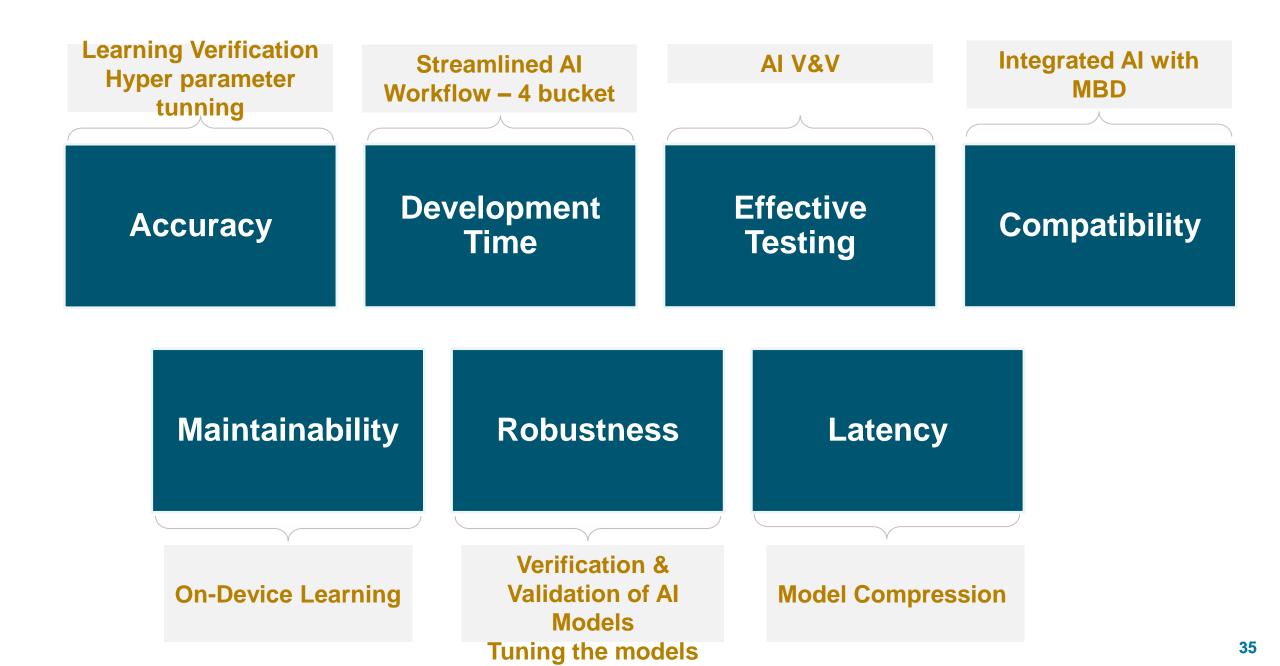
"original [network] was 40MB, was told needed to be **less than 10MB to fit**."



"model is 600kb and want to **reduce it to 300kb**. if I'm not fitting it in, I don't have a working solution"

## Challenges in AI Model Development

Accuracy	<ul><li>Learning Verification</li><li>Hyper parameter tunning</li></ul>
Development Time	<ul> <li>Streamlined AI Workflow – 4 bucket</li> </ul>
Effective Testing	• AI V&V
Compatibility	Integrated AI with MBD & Deployment portfolio
Maintainability	On-Device Learning
Robustness	<ul><li>Verification &amp; Validation of AI Models</li><li>Tuning the models</li></ul>
Latency	Model Compression



## **Industry-Academia initiatives**

CONTRACT OF	Continental
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At Continental, we believe in the power of partnerships to shape the future of mobility. Our collaboration with Amrita Vishwa Vidyapeetham and MathWorks reflects our commitment to bridging the gap between academia and industry, ensuring that future engineers are equipped with both cutting-edge knowledge and industry readiness.

Through the Automotive Systems and Layered Architecture course, students gain exposure to advanced automotive technologies such as **#AUTOSAR**, a critical standard in the **#industry**. By combining academic expertise with Continental's industry insights, we are creating an environment where students can develop the skills needed to address real-world challenges in automotive engineering.

This **#partnership** is not just about **#education** —it's about creating future-ready professionals who will lead the way in innovation. We are proud to support initiatives like this that empower the next generation, preparing them for the dynamic demands of the **#automotive** world.

#### #Collaboration and #innovation are smarter, safer #future for mobility at Bosch and National Institute of Technology Calicut Collaborate on EV Course to Prepare Students for Industry

#### #ContinentalIndia #India #bangalore Challenge

Address the shortage of automotive engineers with system engineering skills

#### Solution

Jointly create a new undergraduate course in model-based system engineering as part of a collaboration between academia and industry

#### Results

- Months of on-the-job training eliminated
- Enrollment increased by 250%
- 90%+ positive feedback received



Pradeep Kumar of Bosch India lighting the ceremonial lamp with Dr. Sivaji <u>Chakravorti</u> of NIT Calicut before signing the agreement.

"The collaboration between NIT Calicut, MathWorks, and Bosch narrowed the gap between academia and industry, producing an electric vehicle system engineering course that has been both well received by our students and highly useful for them as well." - Dr. Kumaravel Sundaramoorthy, NIT Calicut

#### Upcoming course on SDV

#### **Course Outcomes:**

By the end of this course, students will be able to:

1. Understand the architecture and evolution of software-defined vehicles and their role in the automotive industry.

2. <u>Analyze</u> the components, systems, and communication protocols central to software-defined vehicles.

3. Apply software development and validation techniques specific to automotive applications.

4. Explore trends and emerging technologies shaping the future of software-defined vehicles.

#### Module 1: Introduction to Software-Defined Vehicles

Definition, significance, SDVs - Evolution, automotive technology, softwarecentric systems - Software-defined architecture - vehicle design, functionality **Module 2: Fundamentals of SDV Architecture and Components** 

SDV architecture, embedded systems, ECUs - Software platforms, methodologies, SDVs - Software development lifecycle (SDLC) - automotive engineering

#### Module 3: Communication, Connectivity, and ADAS in SDVs

V2V, V2I, V2X communication - In-vehicle networking, CAN, Ethernet -Advanced Driver Assistance Systems (ADAS) - key features - autonomous driving **Module 4: Software Development, Validation, and Future Trends** 

Software development, validation, testing, simulation, ISO 26262 - Case studies, practical implementations - Future trends, OTA updates, cloud-based platforms and Role of AI in SDVs

Bosch and National Institute of Technology Calicut Collaborate on EV Course to Prepare Students for Industry

#### Key Summary...



#### Key Summary...





# **Audience Question**