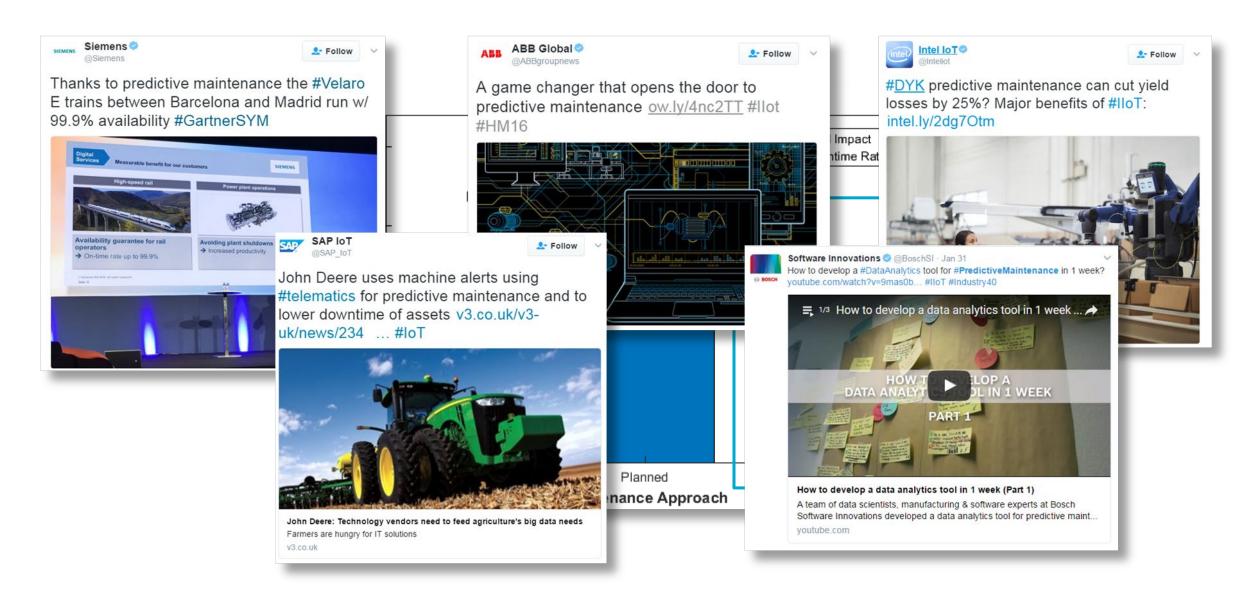
MATLAB EXPO

Deploying Predictive Maintenance Solutions To The Cloud & The Edge

Presenter Name



Predictive Maintenance Promises Improved Operating Efficiency, New Revenue Streams, & A Competitive Differentiator



Predictive Maintenance Matters To Every Single Industry That Is Manufacturing Or Operating Machinery



Aerospace and Defense



Electronics



Process Manufacturing



Automotive



Energy Production



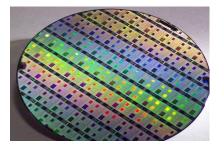
Railway Systems



Medical Devices



Industrial Machinery



Semiconductors

MATLAB & Simulink Are Being Used Today For Predictive Maintenance



The Challenges Associated With Predictive Maintenance Are Consistent Across Industries, for both Data Scientists & Engineers



Too many options for machine learning, feature extraction, etc.



Integrating algorithms with existing infrastructure



Lack of failure data



Hard to get started

Our Solution Addresses Every Challenge By Providing a Workflow That Spans Algorithm <u>Development & Deployment</u>



- Explore and automate feature extraction & machine learning tasks
- Target edge devices through C/C++ codegen
- Integrate with on-premise or cloud-based Enterprise IT/OT systems

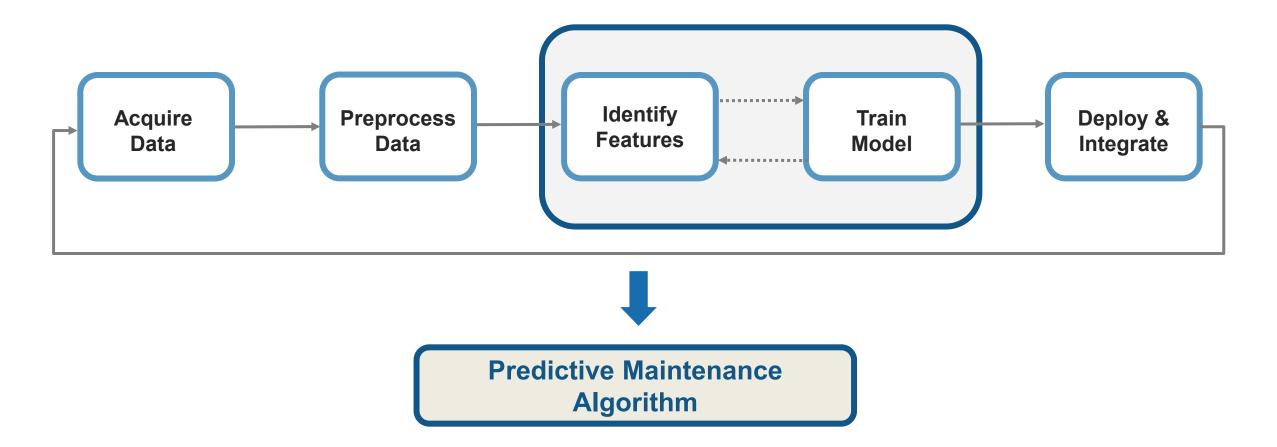
Generate failure data from Simulink & Simscape models of machines



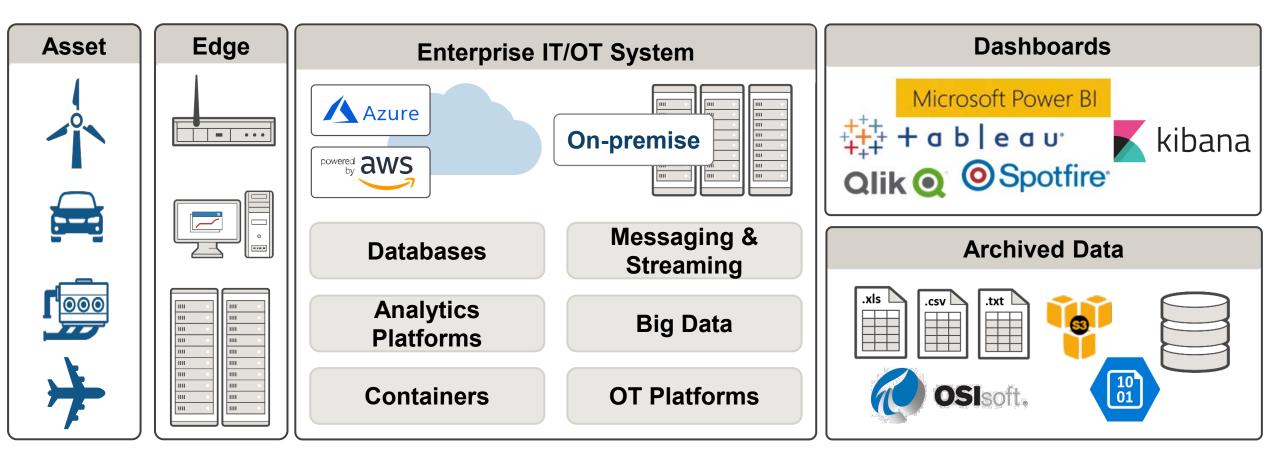


- Get started using Reference Examples
- Work with our Consulting group to scope & define a project

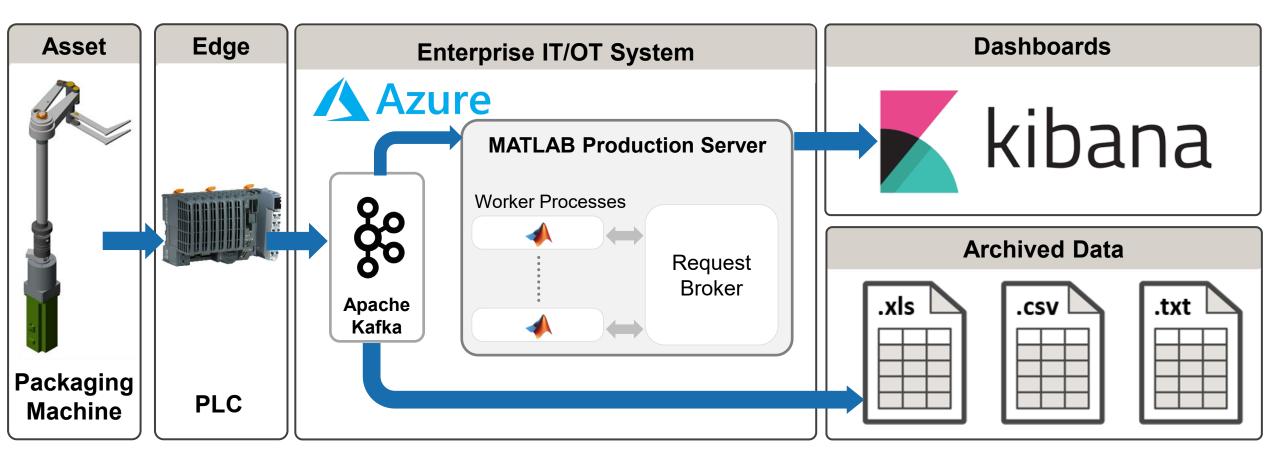
Developing A Predictive Maintenance Algorithm Requires Domain Expertise and Machine Learning Techniques...



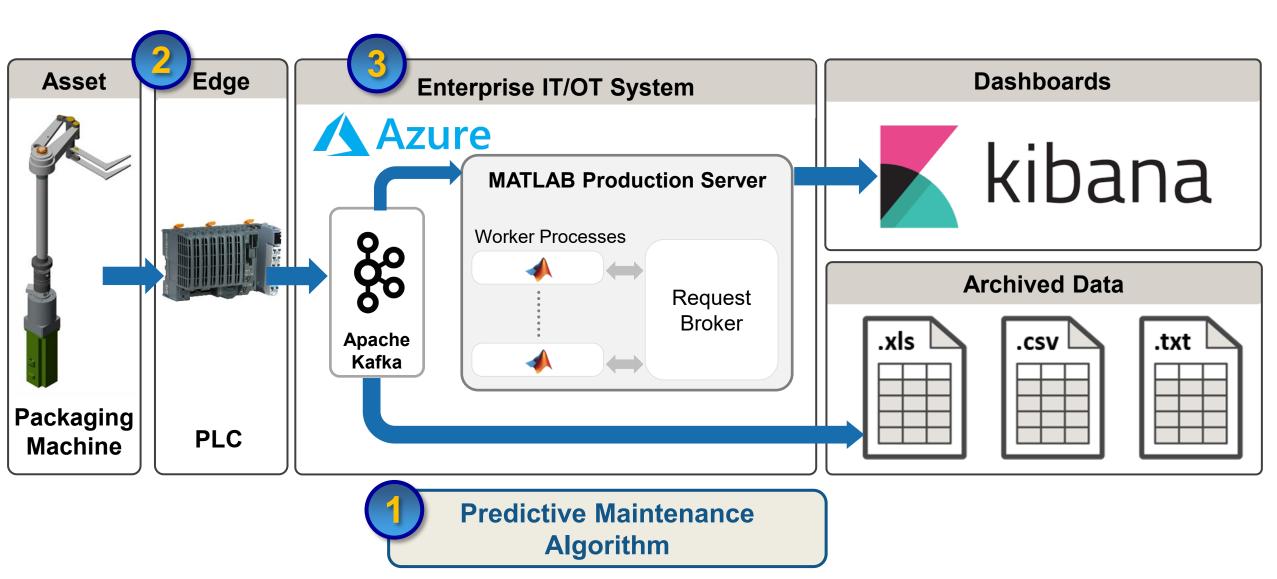
...But Deploying a Predictive Maintenance Algorithm Successfully Is Much More Complicated



Today, We Will Demonstrate How To Deploy A Predictive Maintenance Algorithm To The Edge & Enterprise IT/OT Systems



Today, We Will Demonstrate How To Deploy A Predictive Maintenance Algorithm To The Edge & Enterprise IT/OT Systems



Agenda For Today's Talk



Predictive Maintenance Algorithm Development

Predictive Maintenance Toolbox



Algorithm Test & Deployment To Edge Device

Simulink Real Time & Simulink Coder



Algorithm Deployment to Azure-based IT System

- MATLAB Compiler & MATLAB Production Server





Algorithm Development Includes Remaining Useful Life Estimation, Anomaly Detection, Fault Classification, & Condition Monitoring



Predictive Maintenance Algorithm Development

Predictive Maintenance Toolbox



Algorithm Test & Deployment To Edge Device

– Simulink Real Time & Simulink Coder



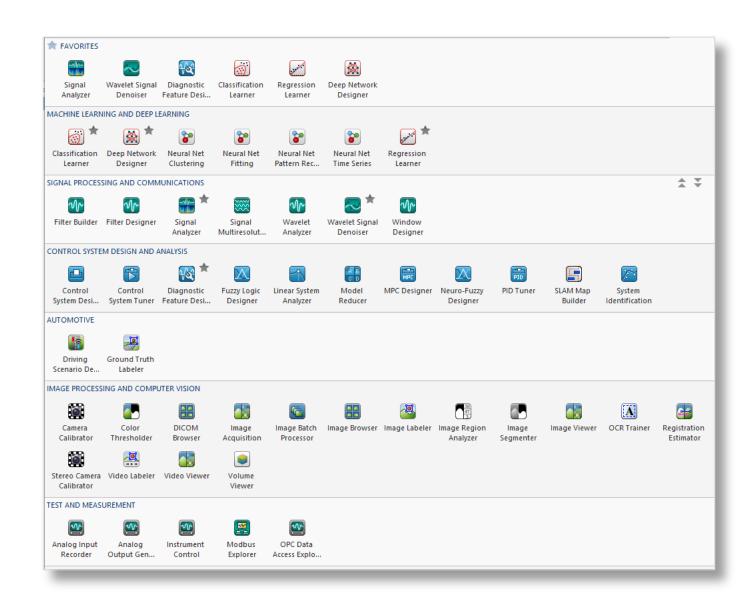
Algorithm Deployment to Azure-based IT System – MATLAB Compiler & MATLAB Production Server



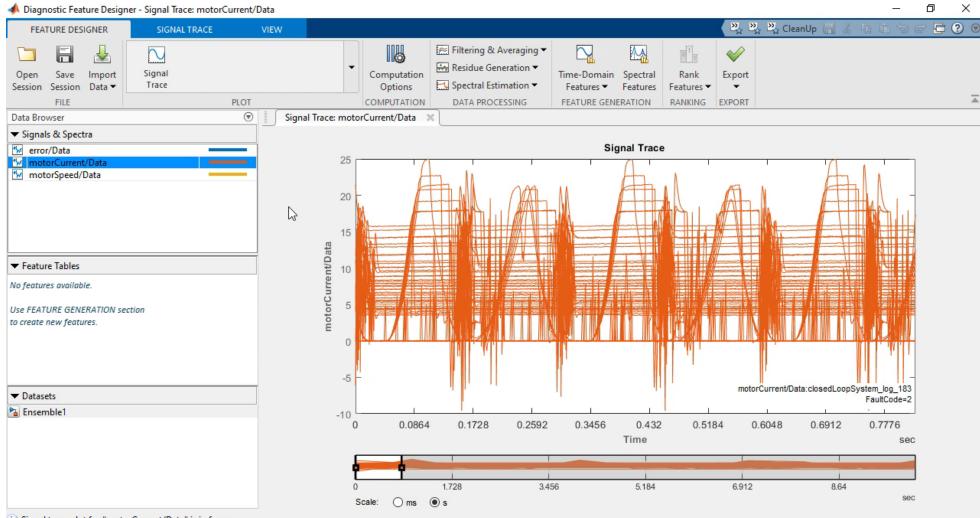


Explore and automate feature extraction & machine learning tasks using MATLAB Apps

- Signal Analyzer
- Wavelet Denoiser
- Diagnostic Feature Designer
- Classification Learner
- Regression Learner
- Deep Network Designer
- ...and many more



Visualize Data, Try Different Feature Extraction Methods & Compare Results Without Writing Any MATLAB Code



I Signal trace plot for "motorCurrent/Data" is in focus.

Metro de Madrid and IMA Are Using MATLAB For Developing Predictive Maintenance Algorithms

Metro de Madrid Adopts Machine Learning for Predictive Maintenance in Tunnels

Raúl Rico, Metro de Madrid

Every day, Metro de Madrid stores more than 10 GB of new data acquired from different sources. Many available tools can only analyze data from a single sensor, and such approaches lack domain expertise. In order to use all the data they acquire for predictive maintenance, Metro de Madrid needed to integrate the data from a wide variety of sensors and customize their signal analysis algorithms.

Metro de Madrid used MATLAB[®] and Statistics and Machine Learning Toolbox[™] to automate the data merging, signal analysis, and algorithm sharing, which enables people without MATLAB experience to perform advanced signal analysis.

Advantages of using MATLAB:

- · Save time in the data validation and analysis phase
- Integrate data from different sources
- Share algorithms with non-MATLAB users



CONCLUSIONS AND FUTURE ACTIVITIES

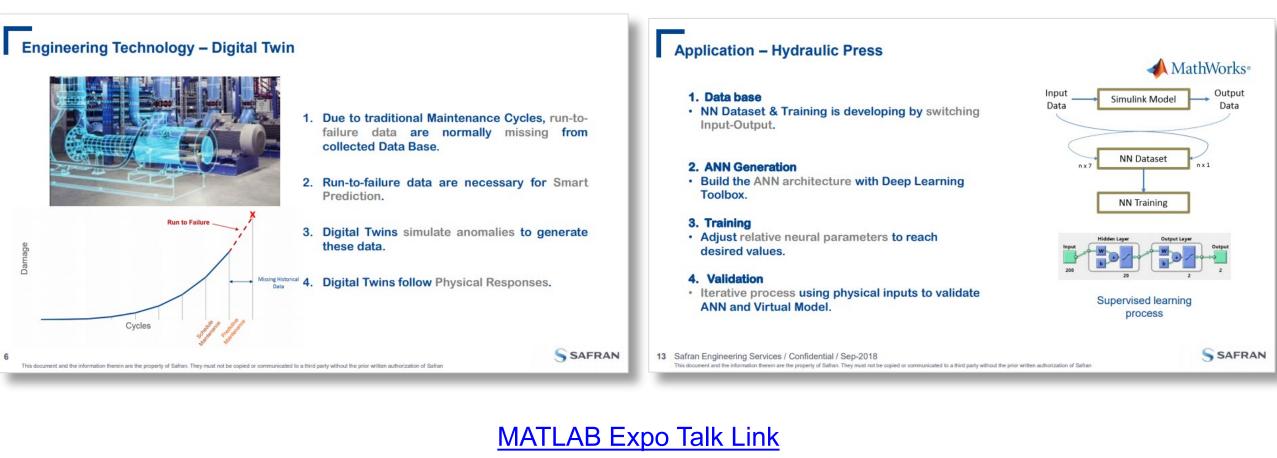
- Using MATLAB tools we managed to extract and select the best features to build a classification model
- The most promising algorithm uses 5 features and has an accuracy of 89%
- Ongoing: check capability of generalization using data of other parts that did not break
- In the future: acquire new data and test the model on-line



MATLAB EXPO

1

Safran Uses Simulink to Generate Failure Data To Train Neural Networks To Detect Anomalies and Predict Failures in Factories



MATLAB **EXPO**

A MathWorks

Edge Device Deployment Enables Data Reduction & Faster Results



Predictive Maintenance Algorithm Development – Predictive Maintenance Toolbox



Algorithm Test & Deployment To Edge Device

Simulink Real Time & Simulink Coder



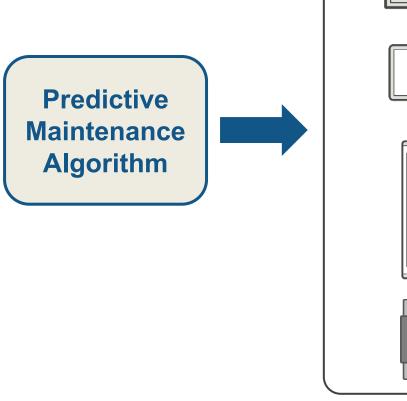
Algorithm Deployment to Azure-based IT System
MATLAB Compiler & MATLAB Production Server

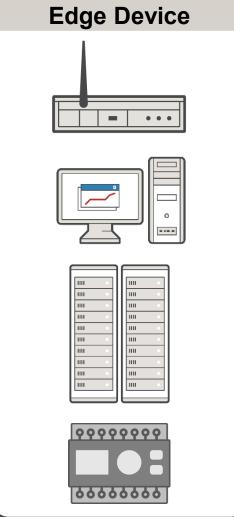




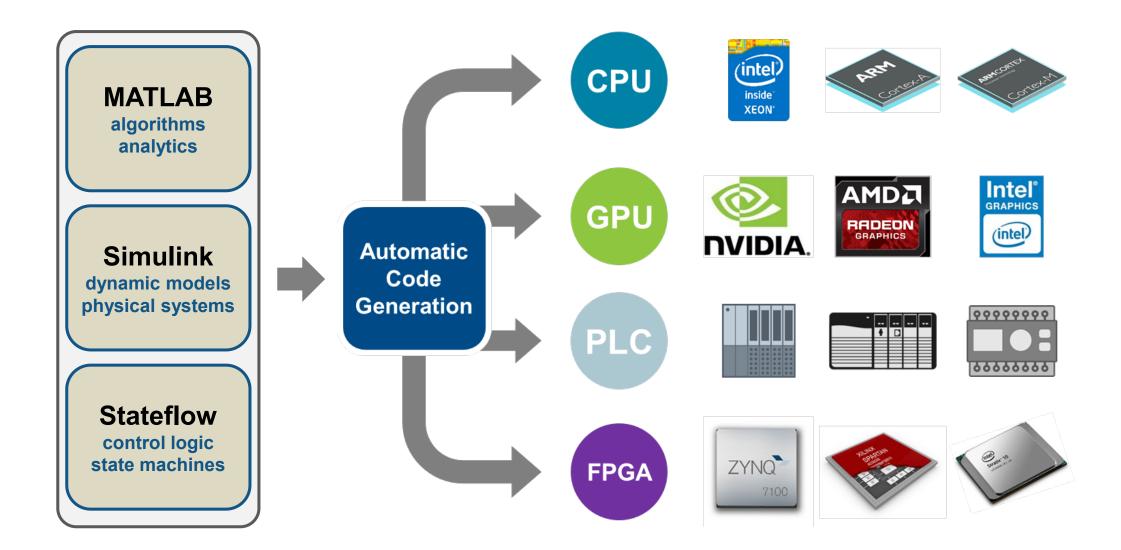
The Steps Associated With Deploying To Hardware Are Complex, But Model-Based Design Is Perfectly Suited To This Application

- 1. Develop algorithm that can run on a resource-constrained edge device
- 2. Test algorithm in simulation
- 3. Verify performance using real-time testing
- 4. Deploy to actual hardware

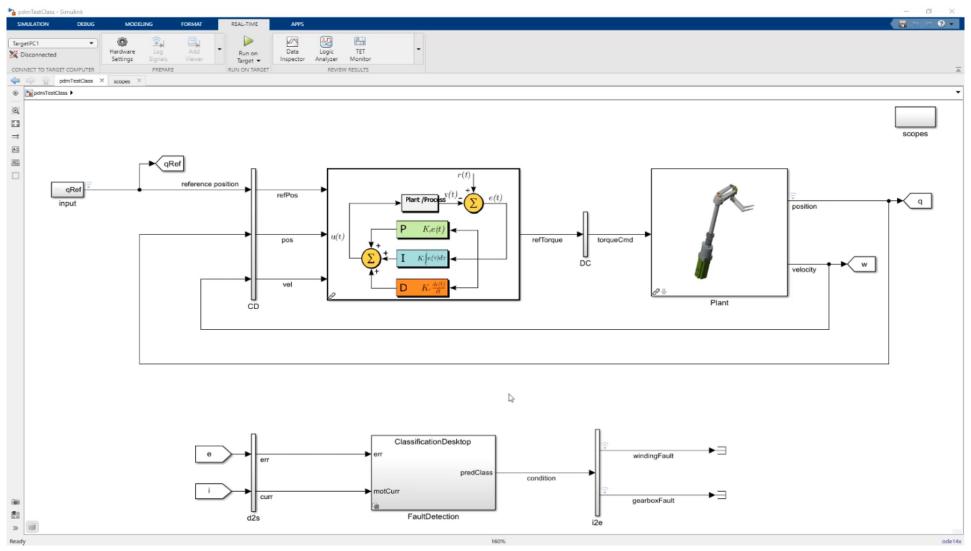




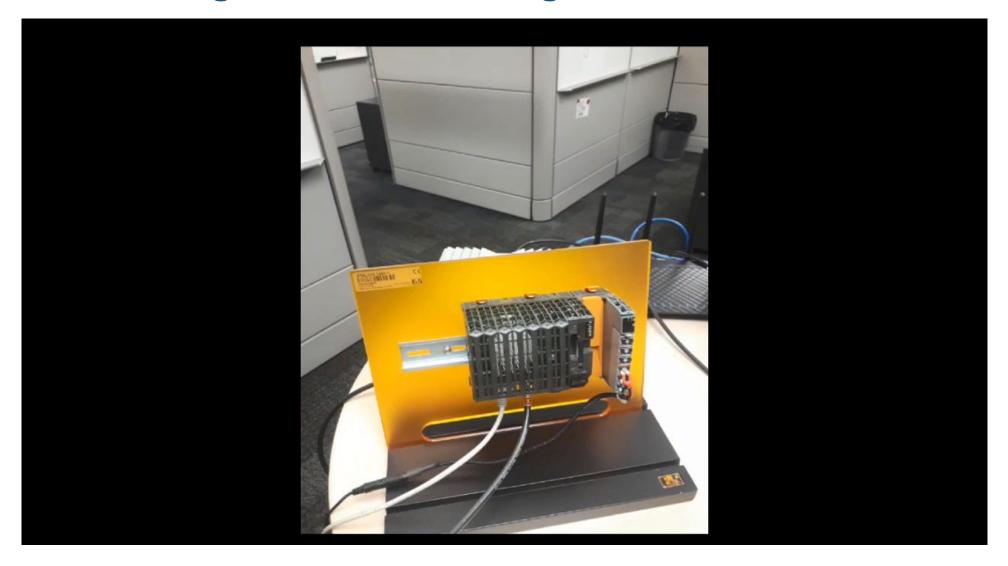
Automatic Code Generation From MATLAB & Simulink Simplifies This Process



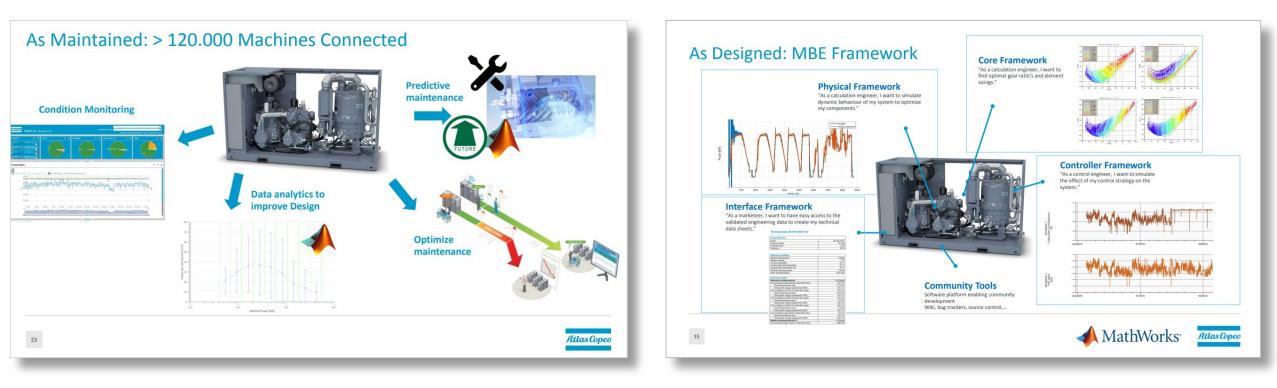
Check If Fault Classification Algorithm Behaves As Expected Using Simulation



Deploy Algorithm To PLC Using Automatic Code Generation & Verify Performance Using Real-Time Testing



Atlas Copco Is Using Model Based Engineering and Digital Twins For Minimizing Cost



MATLAB Expo Talk Link





End Users Require Easy Access To Actionable Information. Dashboards Integrated With IT & OT Systems Make This Possible



Predictive Maintenance Algorithm Development

Predictive Maintenance Toolbox



Algorithm Test & Deployment To Edge Device

– Simulink Real Time & Simulink Coder



Algorithm Deployment to Azure-based IT System

- MATLAB Compiler & MATLAB Production Server





Building Such A System Requires 3 Different Skill Sets: Algorithm Development, Data Visualization, & Data Management





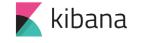
Engineer & Data Scientist

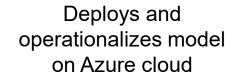
Develops algorithms in MATLAB and Simulink



Dashboard Builder

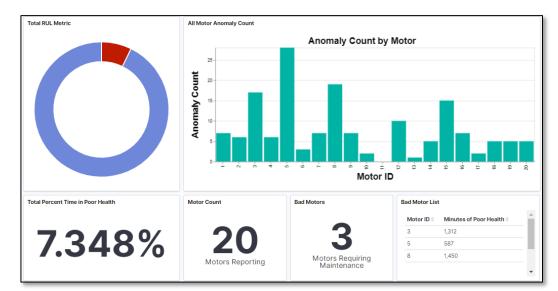
Designs visualization for plant operator





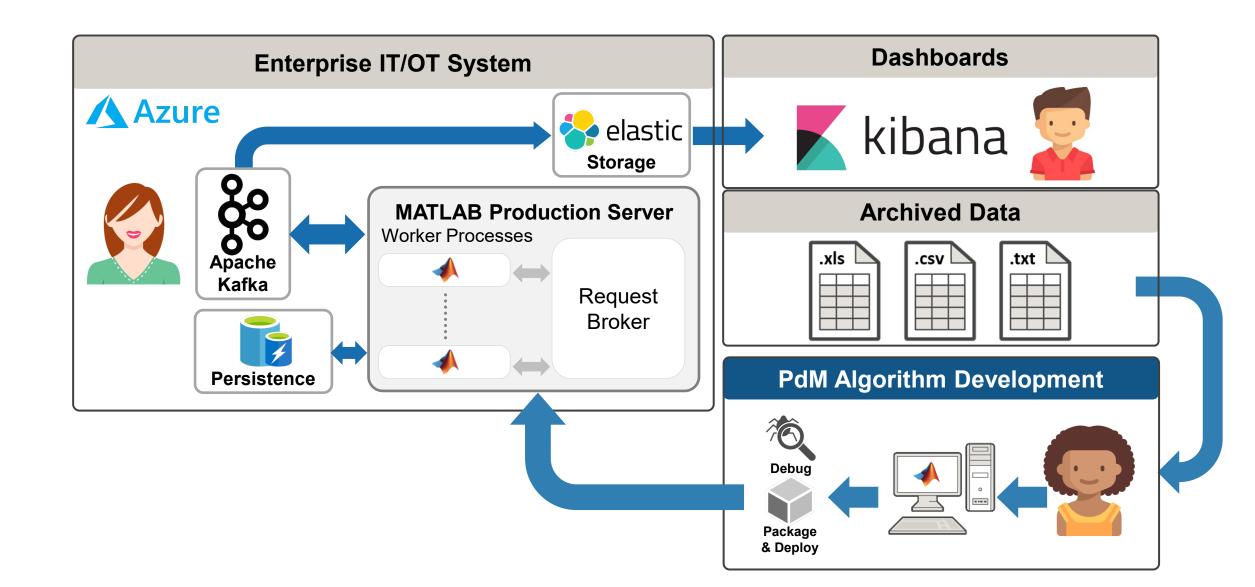
System Architect







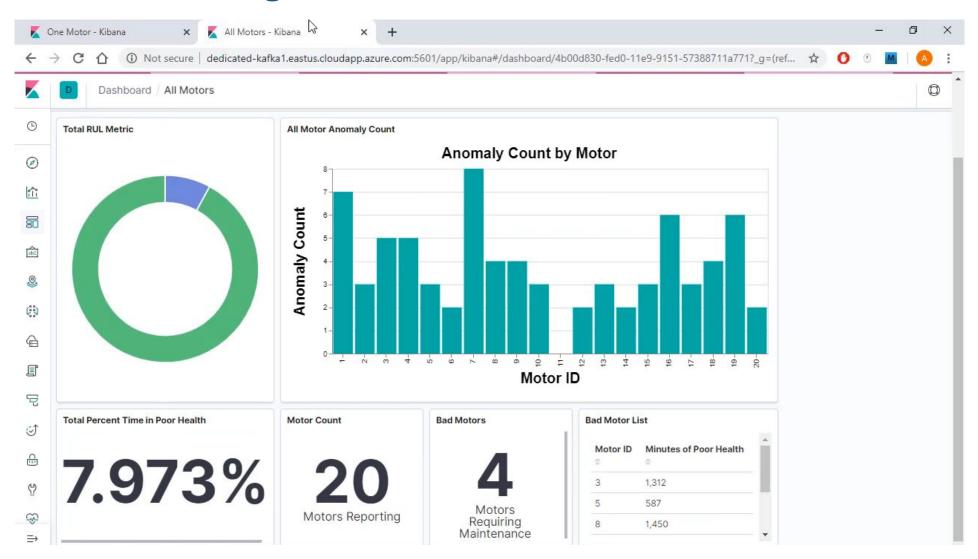
Engineers & Data Scientists Can Package Their Algorithms As Standalone Executables Or Shareable Libraries Using MATLAB



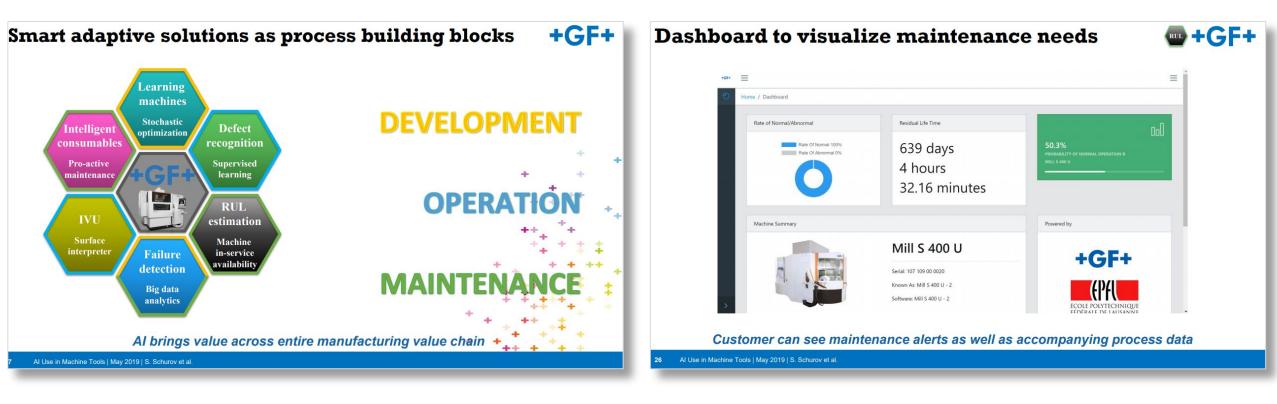
Well Defined Interfaces For Cloud Architectures & One-Click Creation of Cloud-Deployable Applications

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Name - 1 function [nextModel, results] = remainingMotorLife(tbl, prevModel]] 1 AnomalyDetection 2 %Compute remaining useful life for a given motor. Input 2019a 3 %data must be for a single motor. AnomalyDetection 5 % Load the information needed to transform the incoming data, Consume.m 6- transformData = load('rulHealthIndicatorTransformData'); detectAnomalyInputsjon % Copy the timestamp column, motor_ts, which we don't use for RUL detectAnomalyTemplate.m 10- ts = tbl.motor_ts; intAnomalyData 11 % Copy the timestamp columns to extract the features required for RUL. kafkaproperties 2 % Filter the table columns to extract the features named in the kafka-mps-client-stop.bat 14 % Remove all those variables that are not features named in the make_cache_name.m 15 % transformData. OneClassSVM.m 16- rmvars = setdiff(tbl.Properties.VariableNames, is transformData. 16- rmvars = setdiff(tbl.Properties.variableNames, is detection of gigson >> edgeDevice('PackageMachineMotor', 'producer.properties.azure', 1, 'ManyMotors/Motor1.mat')			Variables - dAll	• × 🗹				- Material ife	and the second second second					
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Integrate MATLAB Analytics For Predictive Maintenance With Your Dashboards & Existing IT/OT Infrastructure



GF Machining Solutions Built Condition Monitoring Dashboards To Visualize Maintenance Needs & Predict Failures



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Our Solution Addresses Every Challenge By Providing a Workflow That Spans Algorithm <u>Development & Deployment</u>



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- Integrate with on-premise or cloud-based Enterprise IT/OT systems



Generate failure data from Simulink & Simscape models of machines



- Get started using Reference Examples
- Work with our Consulting group to scope & define a project





Learn More

- Predictive Maintenance Toolbox
 - <u>https://www.mathworks.com/products/predictive-maintenance.html</u>
- Predictive Maintenance Solutions with MATLAB and Simulink
 - <u>https://www.mathworks.com/solutions/predictive-maintenance.html</u>
- Consulting for Predictive Maintenance
 - <u>https://www.mathworks.com/services/consulting/proven-solutions/predictive-maintenance.html</u>



