MATLAB EXPO

건전성 예측관리 시스템의 개발/운용 효율을 위한 DevOps 구축

엄준상 부장, 매스웍스코리아



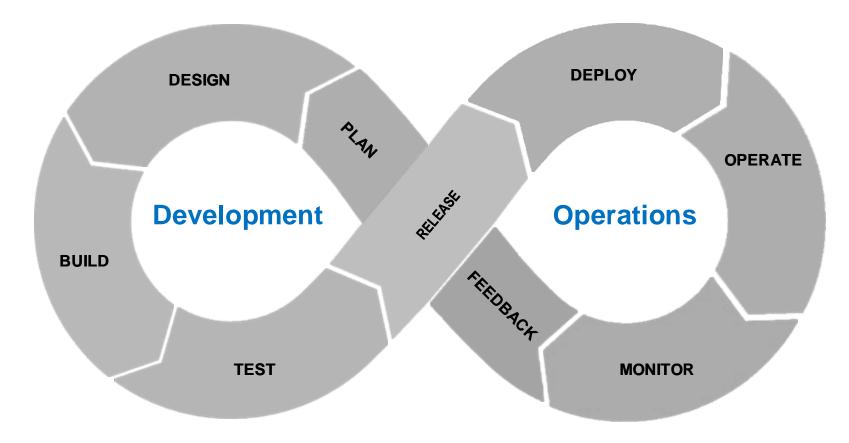


Key Takeaways

- Incorporate familiar MATLAB capabilities, including Predictive Maintenance and Drift Detection, in operations
- Integrate with production systems like data sources and dashboards, and translate those integrations from desktop to cloud servers
- Automatically build, test, package, and deploy MATLAB code and Simulink models with CI/CD

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DevOps: Develop and Operate Production Software

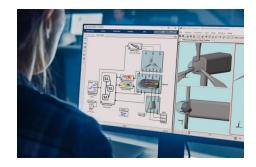




Predictive Maintenance

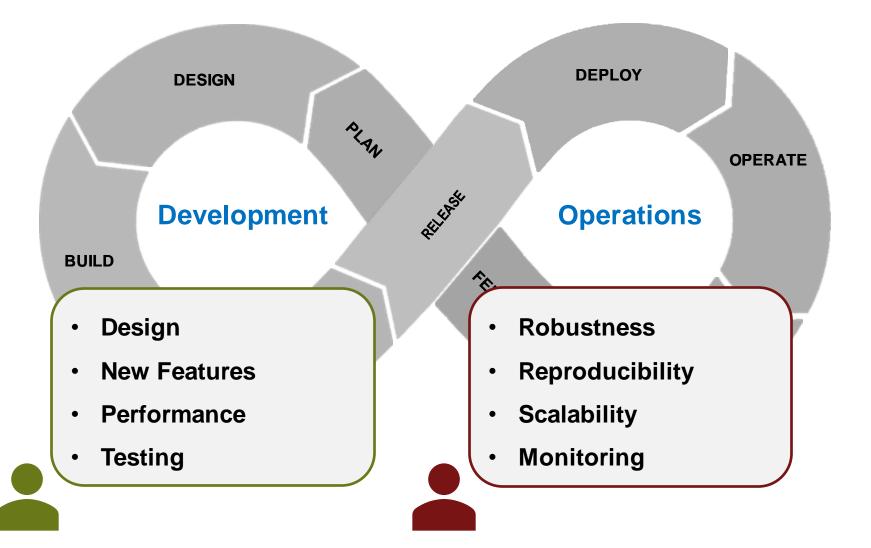


Financial Modeling



Embedded Controls

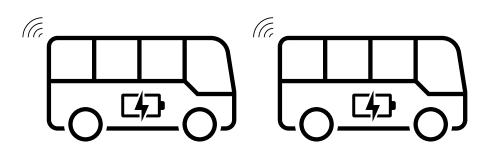
Not necessarily a conflict of interests, but certainly different interests

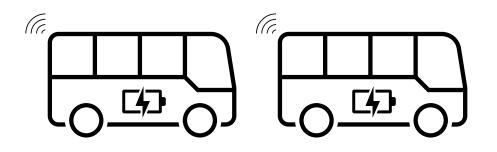


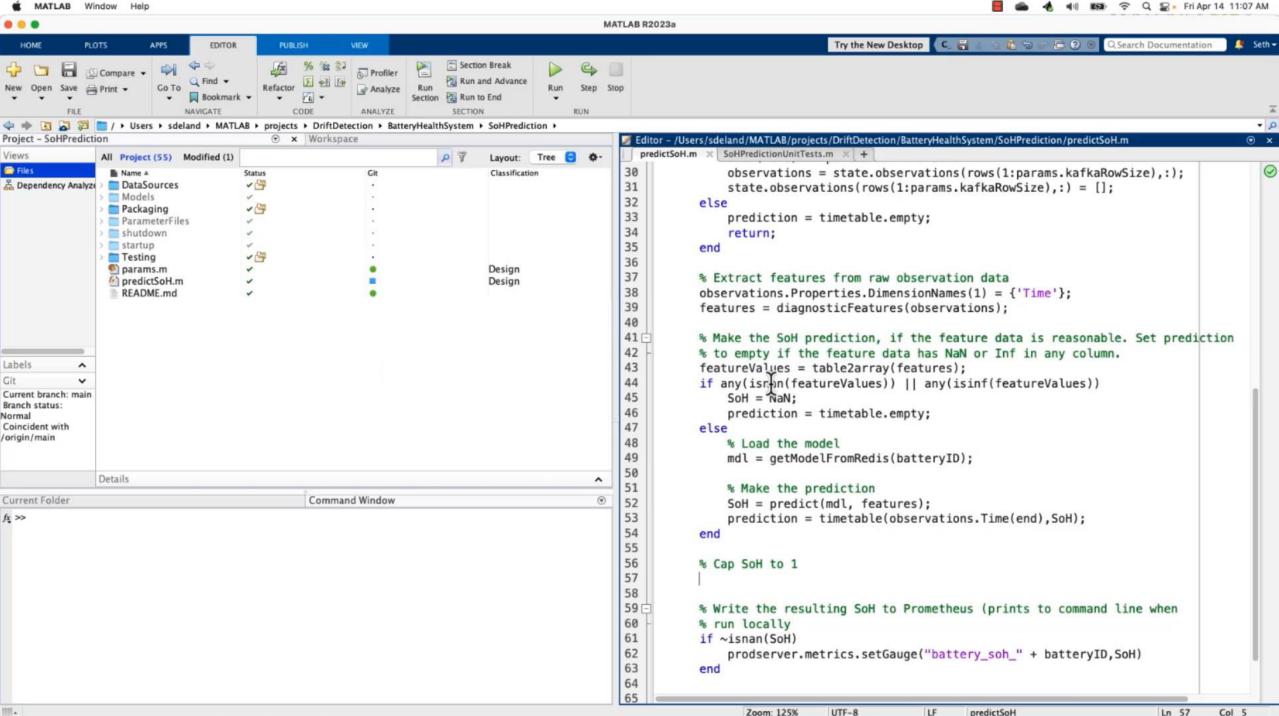
Example: Predicting Battery State-of-Health

- Fleet of electric buses
- Maintenance is expensive. Could we do a better job predicting when batteries need replacing?
- Started gathering telemetry data on batteries

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1800x7 timetable											
	timestamp	1 Current	2 Voltage	3 Temperature1	4 Temperature2	5 SoC_B1	6 SoC_B2	7 BatteryID			
1	01-Nov-2021 00:	2.6869	7.4436	333.1463	332.7619	0.4995	0.4995	1			
2	01-Nov-2021 00:	2.6872	7.4426	333.1317	332.3924	0.4990	0.4990	1			
3	01-Nov-2021 00:	2.6876	7.4417	333.1073	332.0405	0.4985	0.4985	1			
4	01-Nov-2021 00:	2.6879	7.4408	333.0740	331.7048	0.4980	0.4980	1			
5	01-Nov-2021 00:	2.6882	7.4399	333.0327	331.3844	0.4975	0.4975	1			
6	01-Nov-2021 00:	2.6885	7.4390	332.9843	331.0783	0.4970	0.4970	1			
7	01_Nov_2021_00+	2 6880	7 / 2 2 1	222 0205	220 7857	0 4965	0 4065	1			







Zoom: 125% UTF-8 LF Ln 57 Col 5 🖲 🔍 🧑 Batteries - Dashboards - Grata 🗙 🕂

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> Notes (1 panel)

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- Battery and Model Health

Battery Health									
81.0%	82 [°] 9%	83.7%	87.2%	81.4%	82.3%	⁷ 88.6%	96.7%	85.6%	¹⁰ 89.7%
84.4%	¹² 81.7%	¹³ 85.5%	¹⁴ 84.6%	¹⁵ 79.5%	¹⁶ 83.7%	¹⁷ 84.0%	¹⁸ 82.3%	¹⁹ 81.0%	²⁰ 93.5%

~ Battery 1 Health and Drift Status



② Last 30 minutes ~ Q Q

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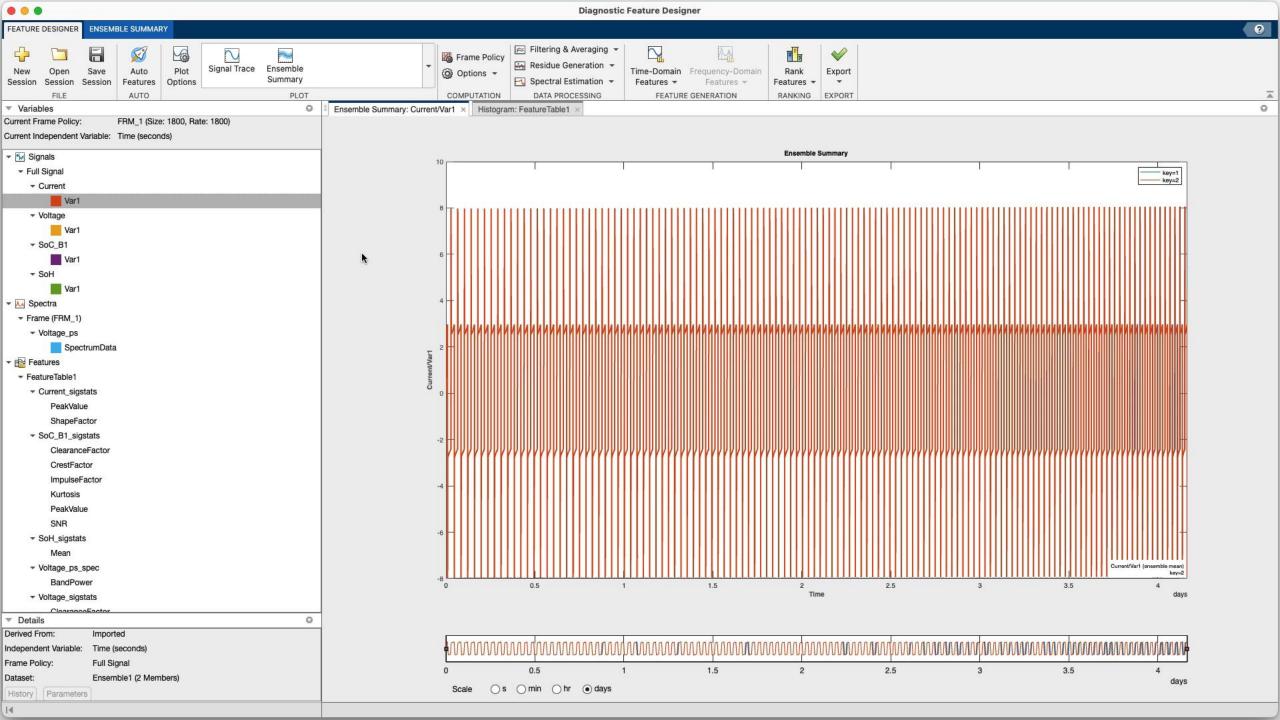
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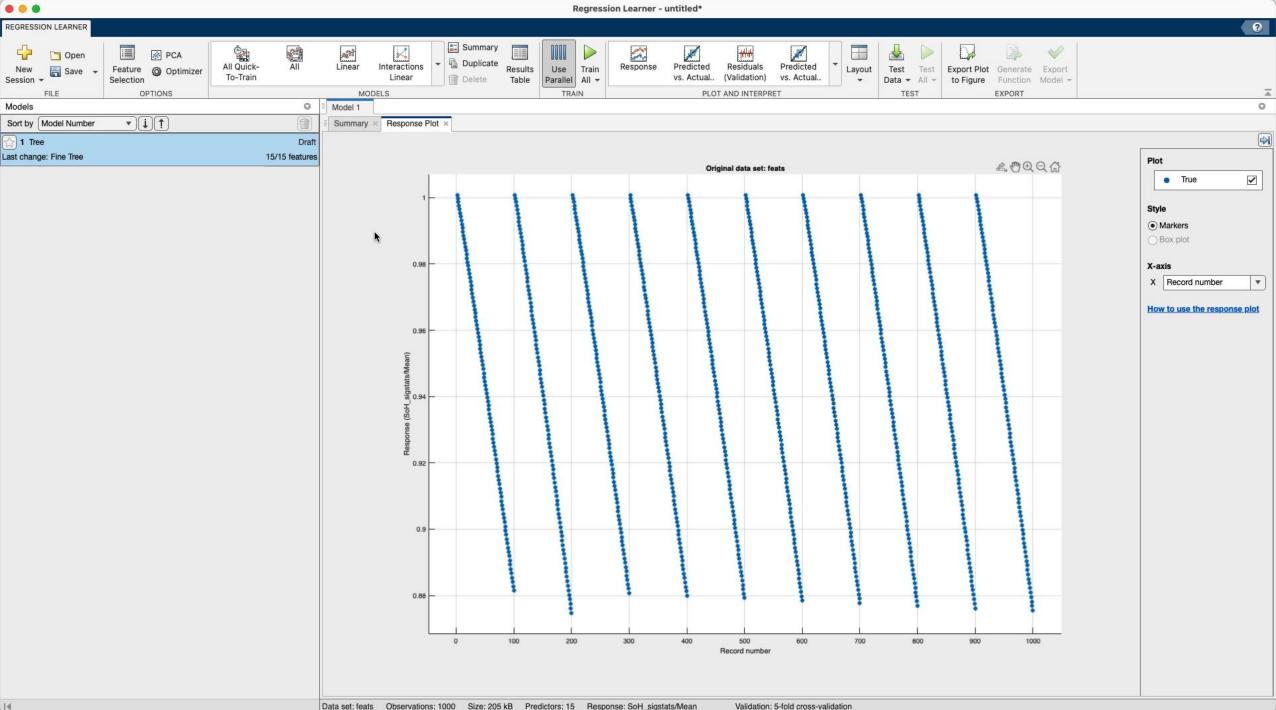
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Create a SoH prediction function using domain-specific tools for engineering data and predictive maintenance



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FILE VARIABLE CODE SIMULINK ENVIRONMENT RESOURCES Image: International control of the patternation of the patt	
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Name A Value BuildingSoHPredictionModel.mlx × + Battery State-of-Health Prediction Battery State-of-Health Prediction Battery State-of-Health (SoH) prediction is important for making sure batteries have not degraded beyond their useful range, and for estimating the capacity of the battery. Making accurate SoH predictions is challenging due to the complex , and difficult-to-anticipate	
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	-
Load historical data We use historical data from several batteries to build our model. We format the data as an "ensemble", which helps us keep the data organized by the battery that it came from.	
Current Folder Image v Name v Image v Image v Image v	
Look at SoH by battery	
<pre>8 figure 9 hold on 10 for i = 1:height(batteryData) 11 plot(batteryData.SoH{i}.Time, batteryData.SoH{i}.Var1, DisplayName=num2str(i)); 12 and</pre>	
12 end 13 legend show 14 title('State of Health by Battery')	
Zoom: 150% UTF-8 LF script Ln 1 Col 45	



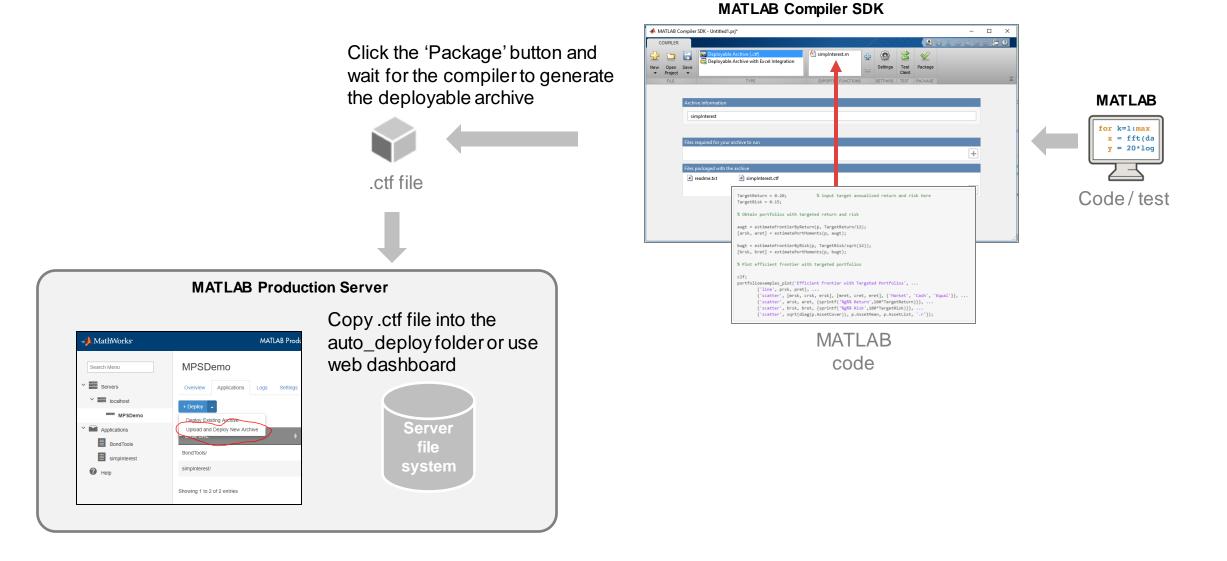


Data set: feats Observations: 1000 Size: 205 kB Predictors: 15 Response: SoH_sigstats/Mean

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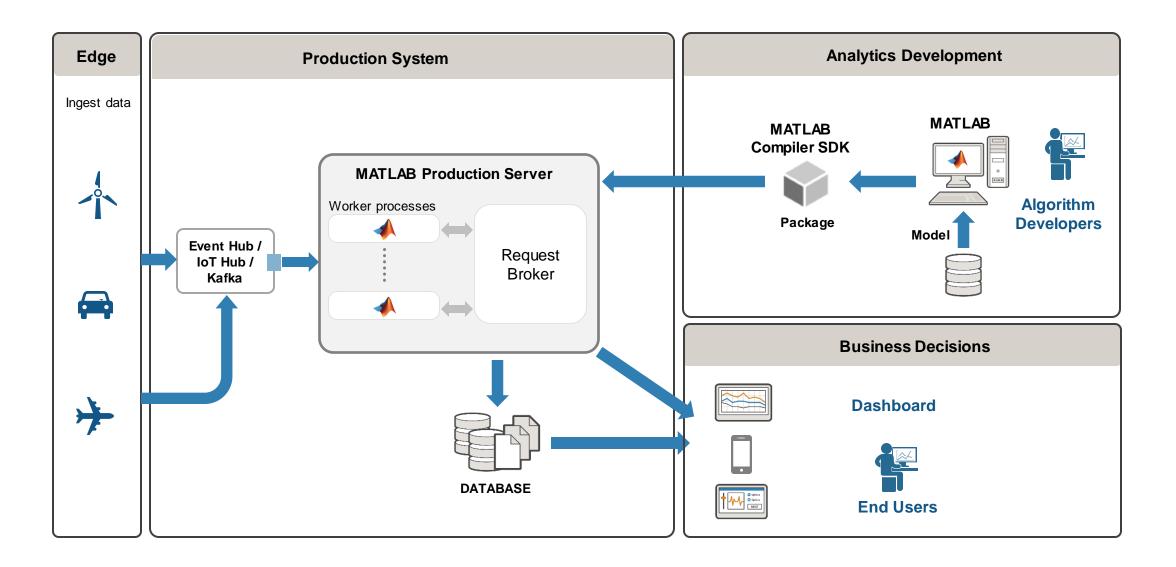
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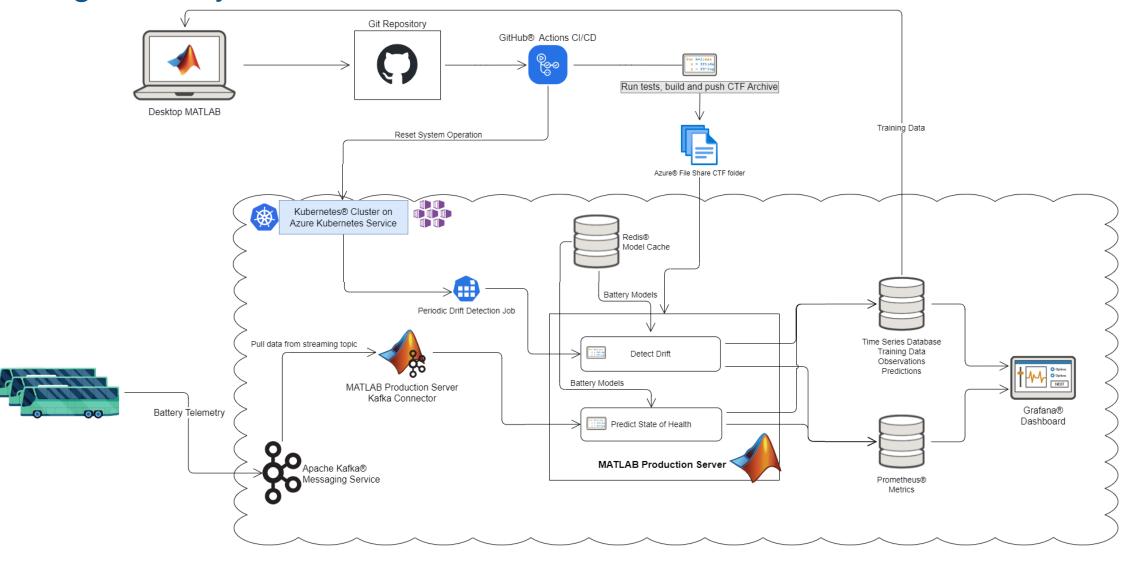
Typical authoring workflow

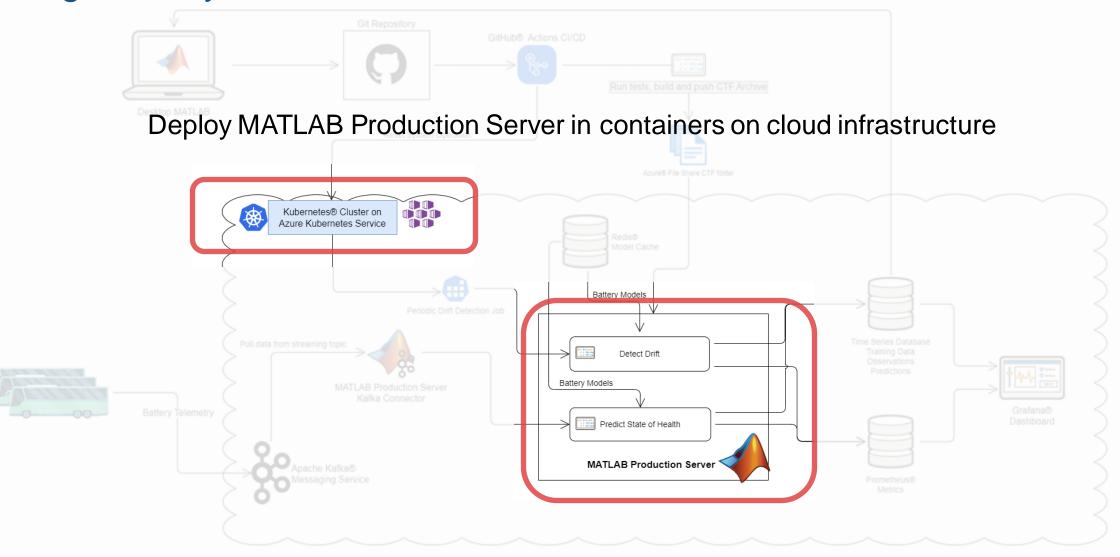


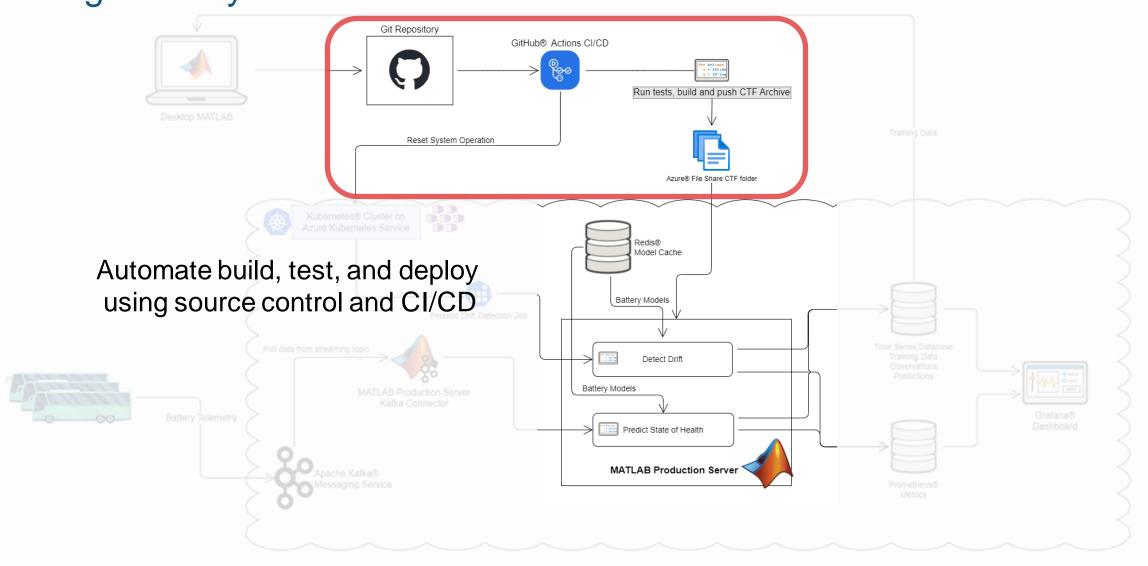
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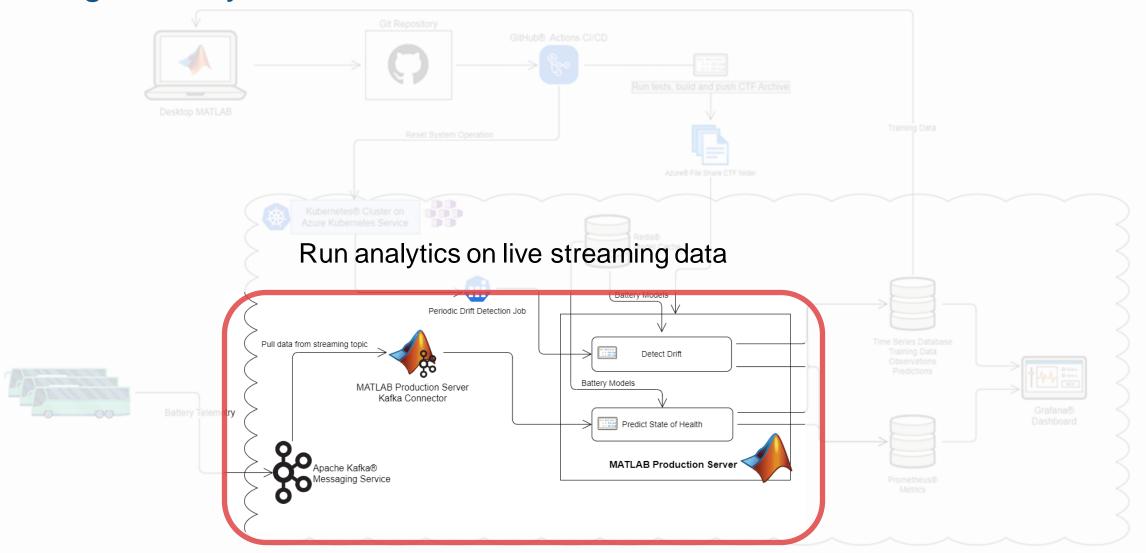
Example reference architecture

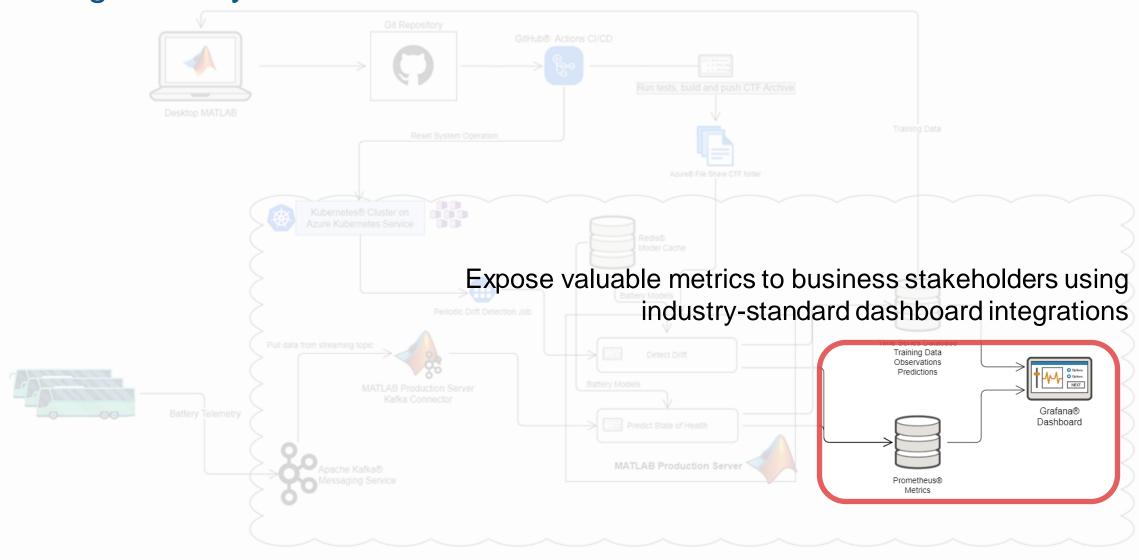


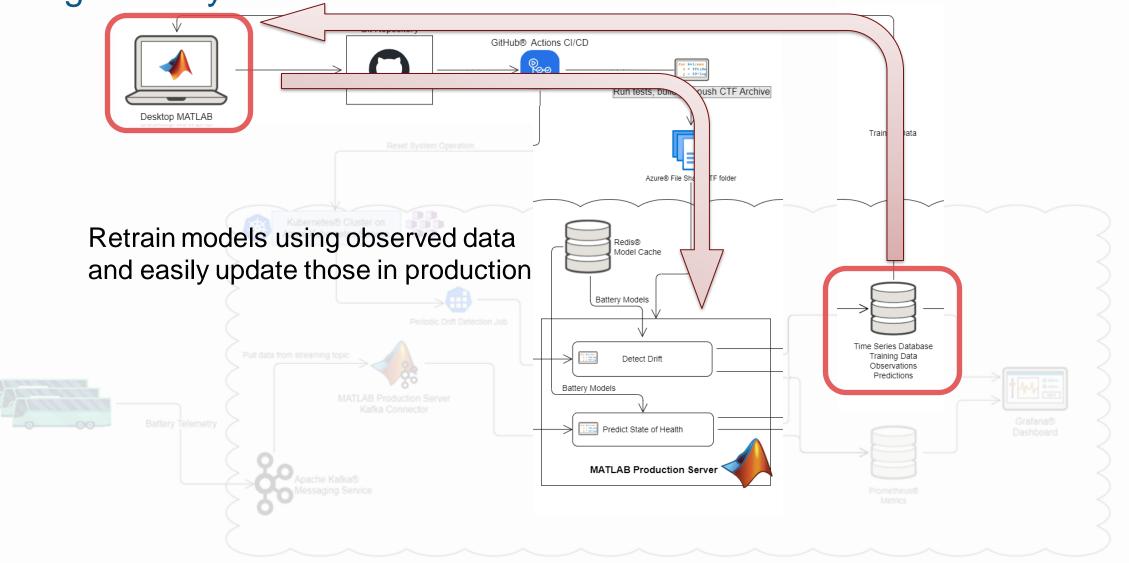












State of health algorithm in production

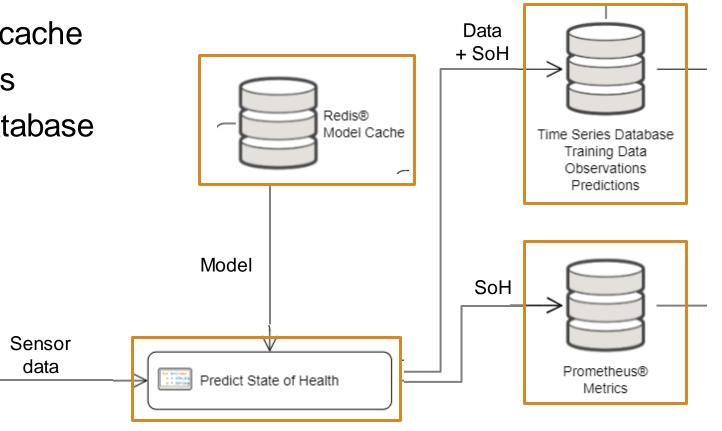
Production System

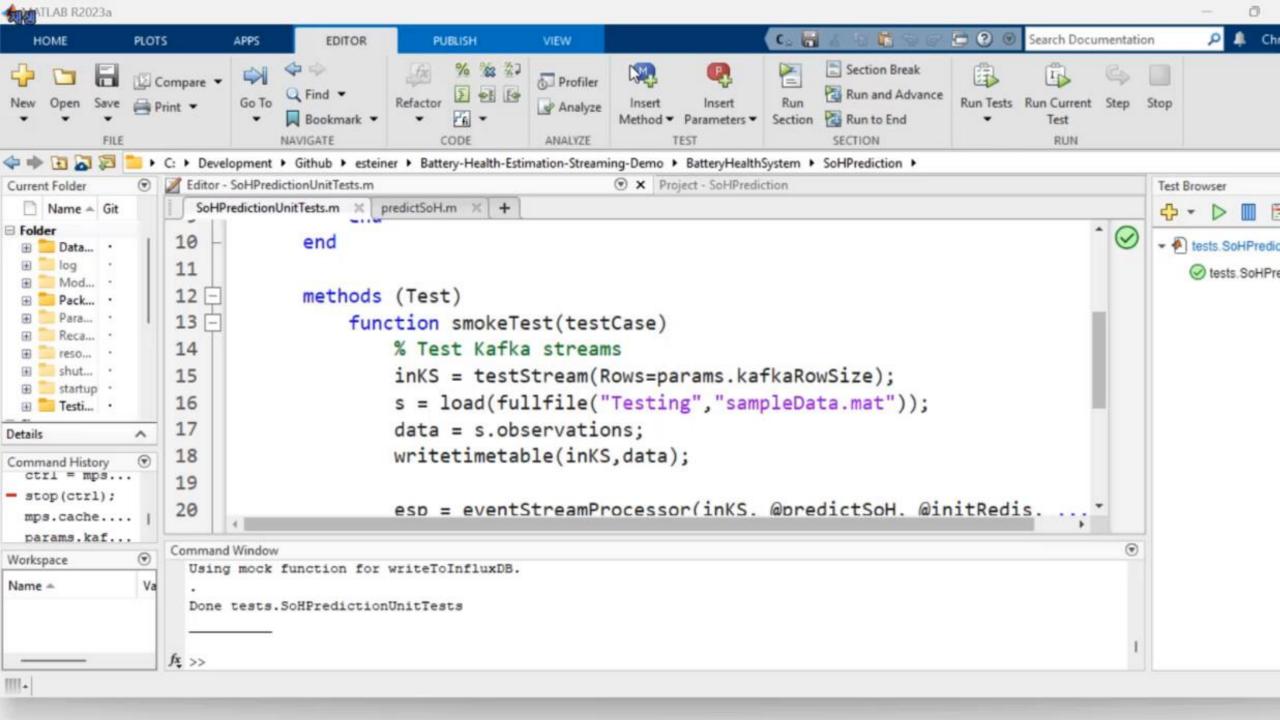
- Receive sensor data as kafka stream
- Load battery model from Redis cache
- Expose metrics with Prometheus
- Save data and predictions to database
 Local testing

Apache Kafka®

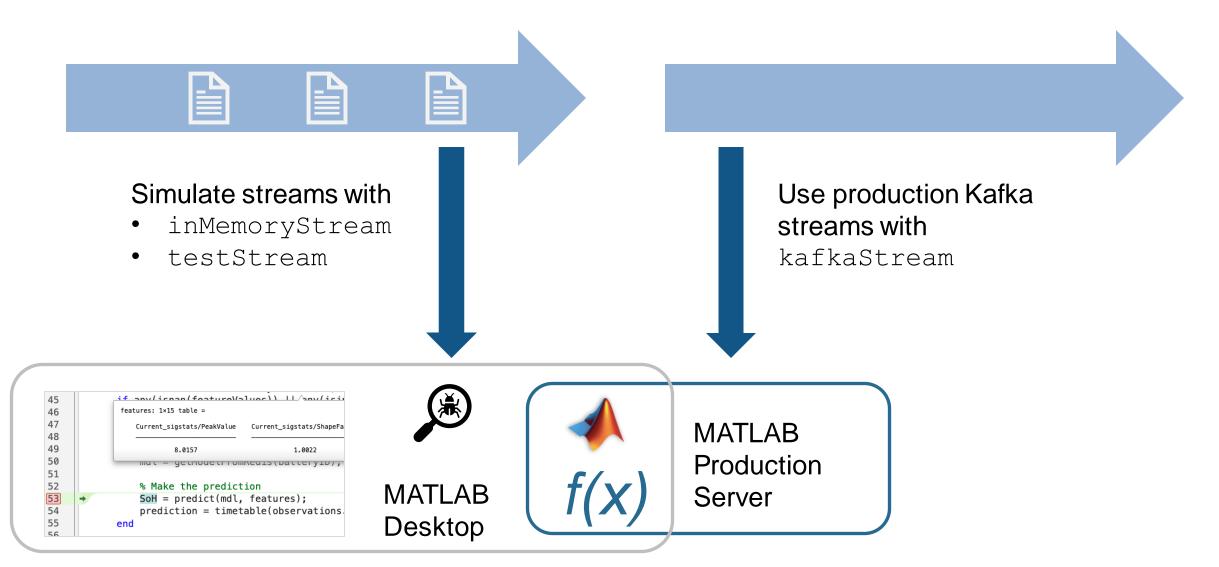
Messaging Service

Mock dependencies



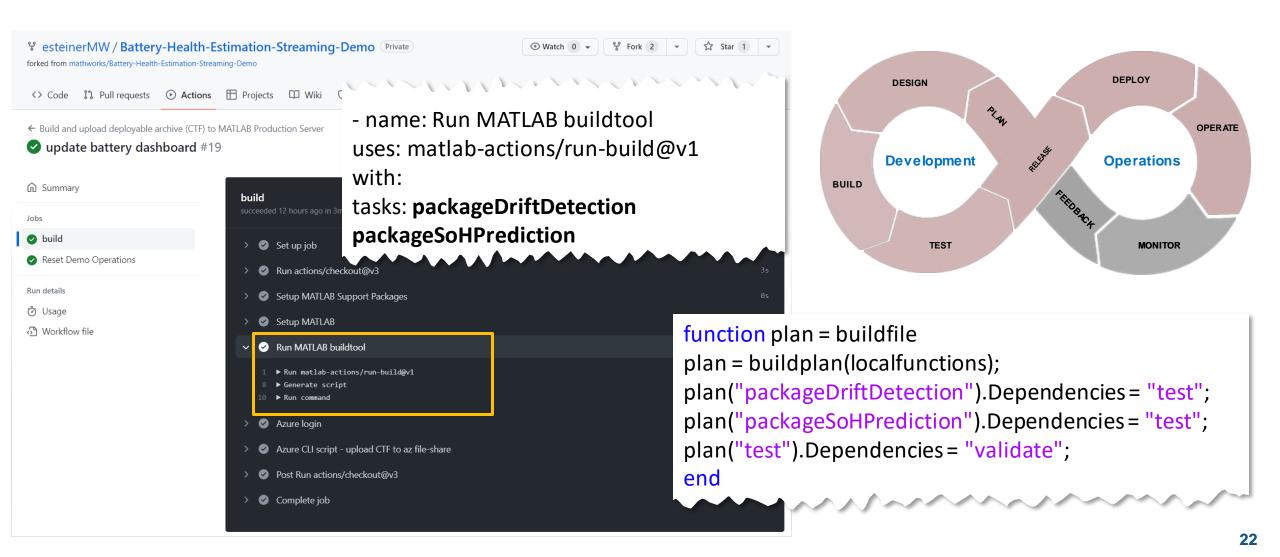


Write SoH prediction function to use kafka streams



Debug locally, then deploy the same MATLAB code to production.

Automatically build, test, package, and deploy MATLAB code



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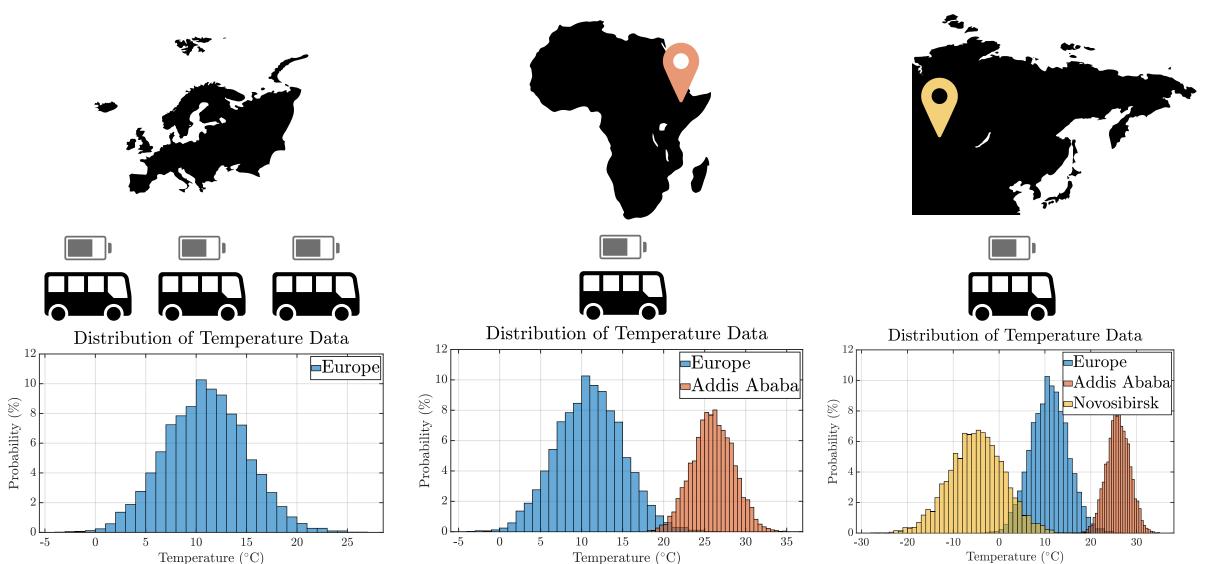
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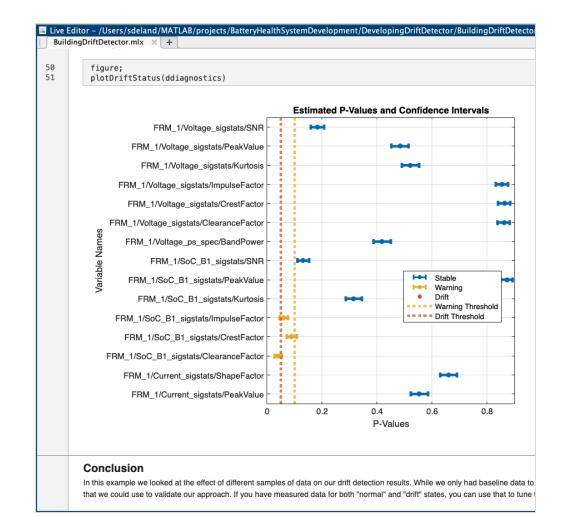
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The static data assumption rarely holds in the real world

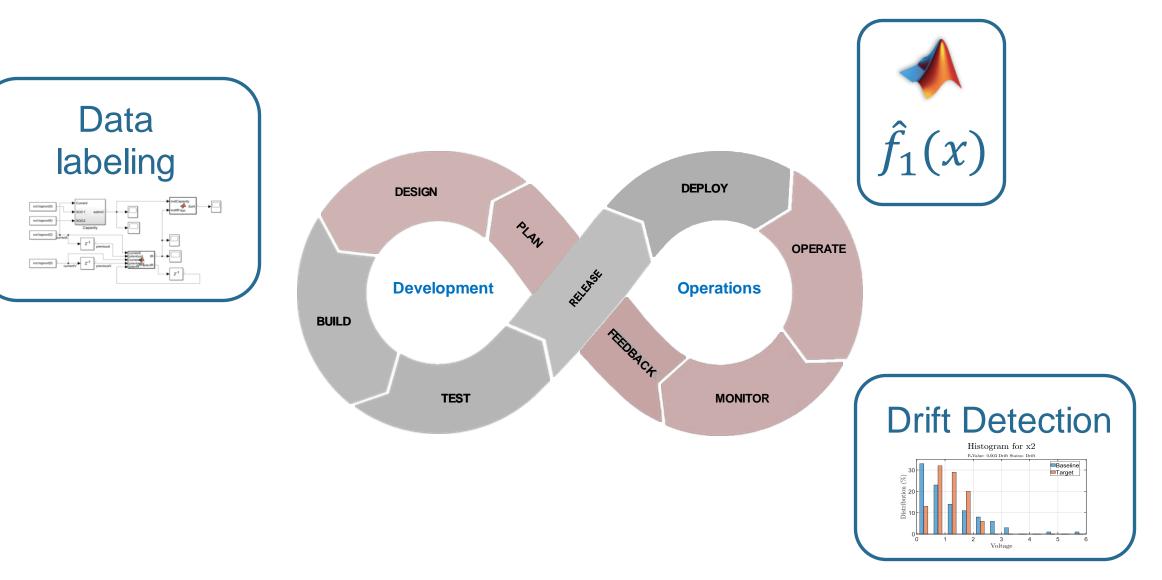


Developing drift detection with detectdrift

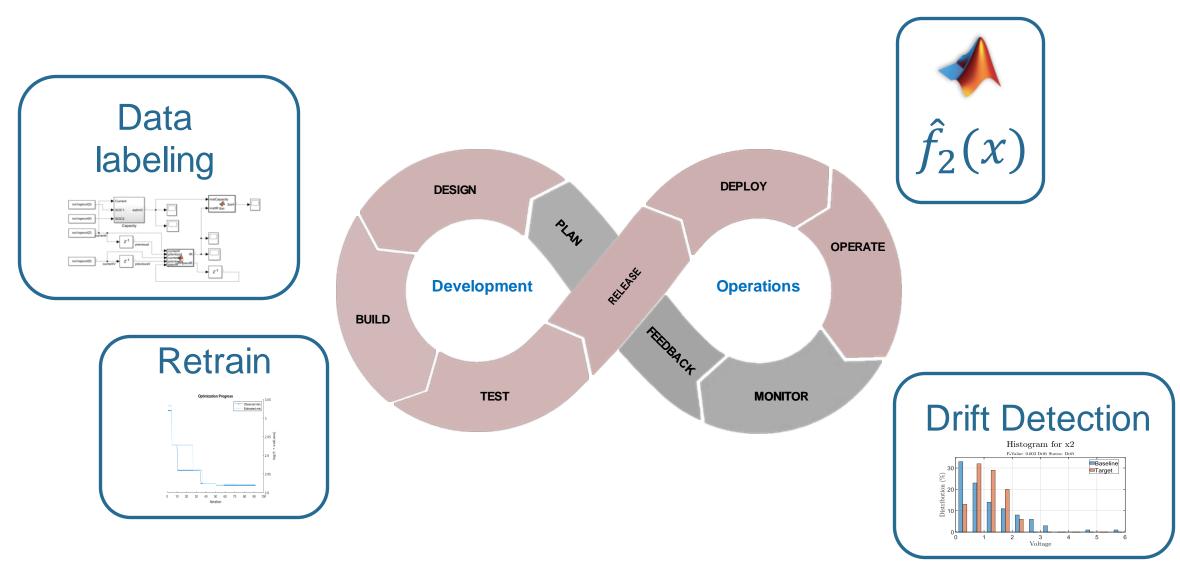
- Use historical data (training data) to create a baseline distribution
- Generate synthetic data to test for drift
 - This will be replaced by streaming data in the production system
- <u>Out-of-Distribution Detection for Deep</u> <u>Neural Networks</u>



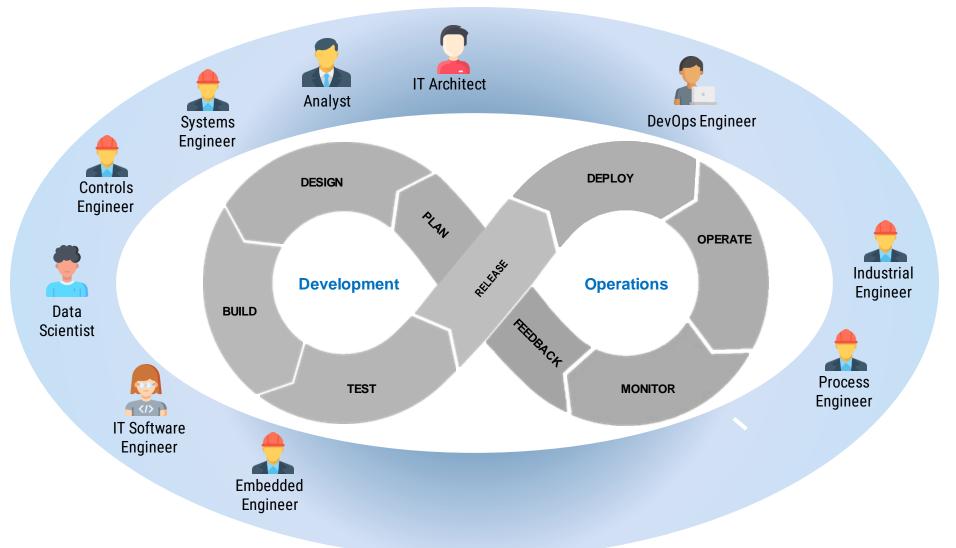
Update model when drift is detected



Update model when drift is detected



Diverse teams across Engineering, IT, Line of Business, and Operations must collaborate to achieve success in DevOps



Key Takeaways

- Incorporate familiar MATLAB capabilities, including Predictive Maintenance and Drift Detection, in operations
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Learn More





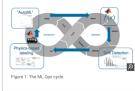


Automating Machine Learning with $\ensuremath{\mathsf{DevOps}}$ for MATLAB and Simulink

By Peter Webb and Gokhan Atinc, MathWorks

As more organizations rely on machine learning applications for core business functions, many are taking a closer look at the full lifecycle of those applications. The initial locus on development and deployment of machine learning models has expanded to encompass continuous monitoring and updates. Changes in the input data may decrease a model's predictive or classification accuracy. Promy training and model evaluation produces better models and more accurate decisions.

In machine learning operations, or ML Ops, the plan, design, build, and test activities of development are linked with the deploy, operate, and monitor activities of operations in a continuous feedback loop (Figure 1). Many data science teams have started to automate parts of the ML Ops cycle, such as deployment and operations.



How MATLAB and Simulink are used with Enterprise IT

https://www.mathworks.com/solutions /enterprise-it-systems.html

CI/CD Resources

https://www.mathworks.com/ solutions/enterprise-itsystems/ci-cd.html

MATLAB and Simulink in the Cloud

https://www.mathworks.com/ solutions/cloud.html

Automating Machine Learning with DevOps for MATLAB and Simulink

https://www.mathworks.com/company/n ewsletters/articles/automating-machinelearning-with-devops-for-matlab-andsimulink.html

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Thank you



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