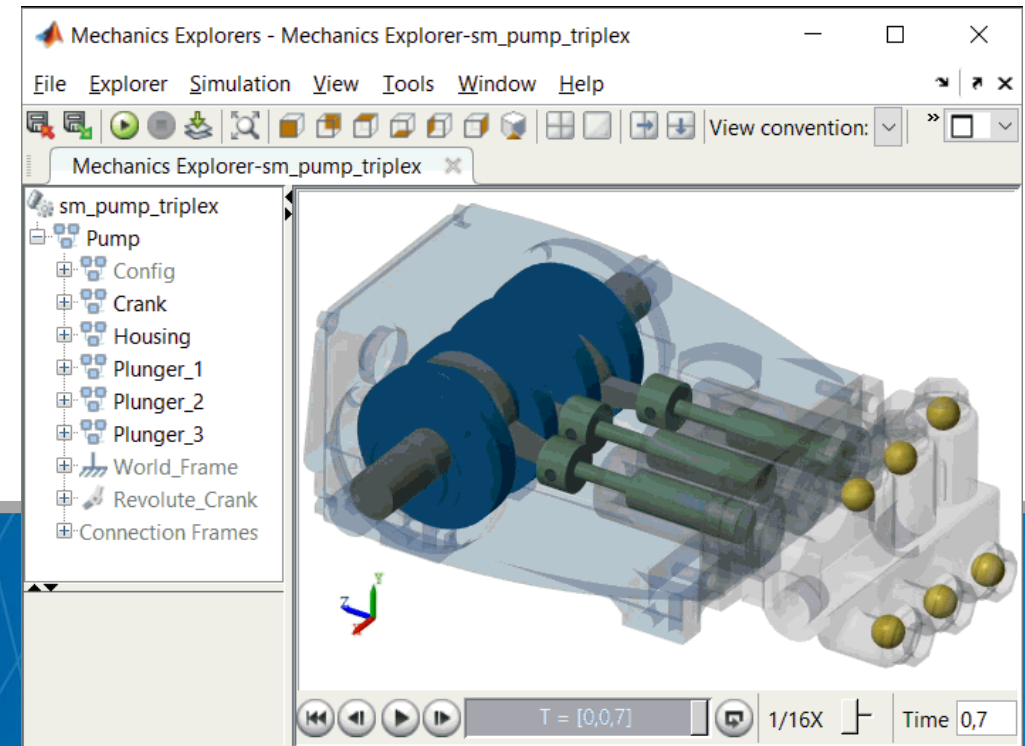
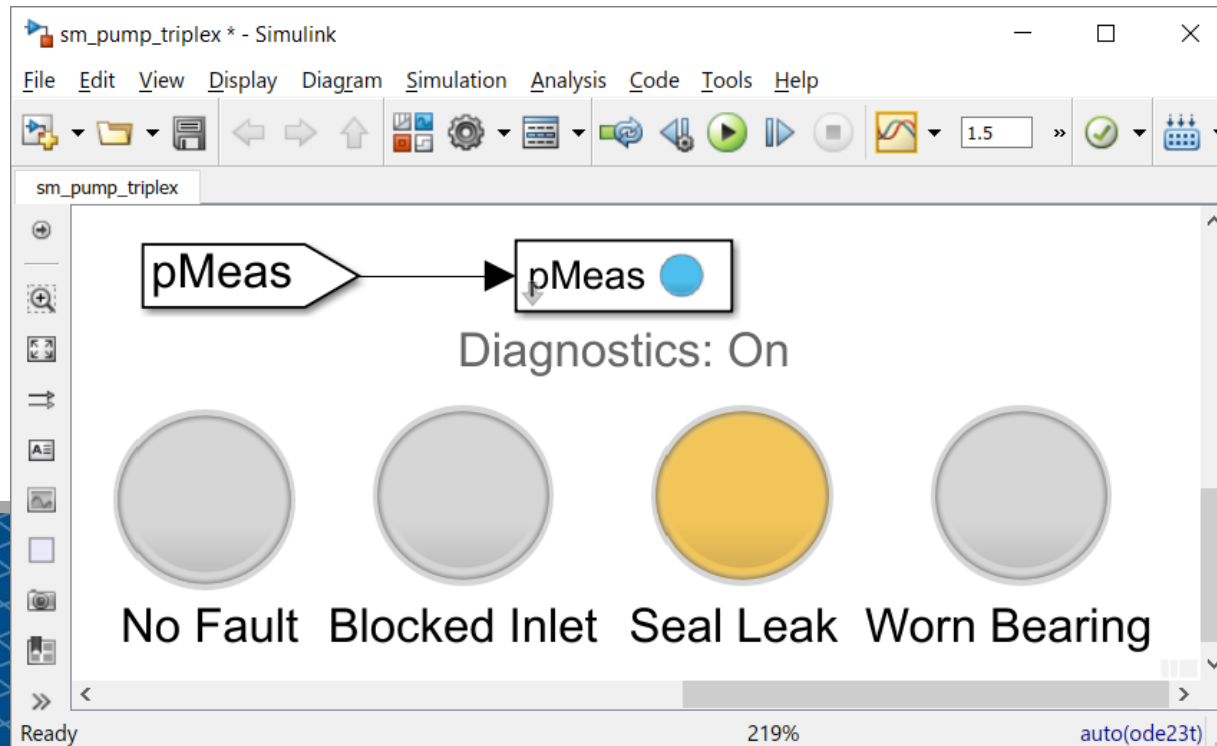


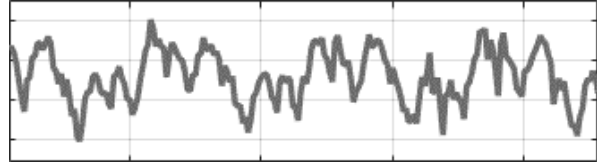
# Predictive Maintenance Using Digital Twins



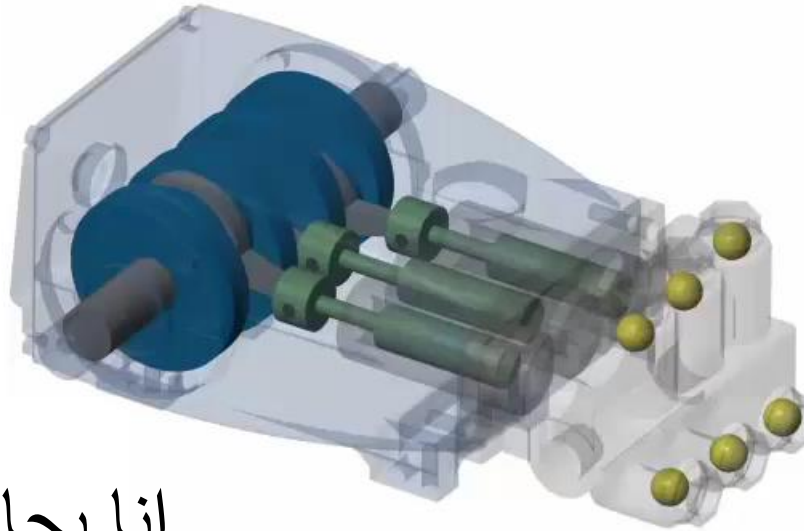
ฉันต้องการความช่วยเหลือ.

Ich brauche Hilfe.

Segítségre van szükségem.



Necesito ayuda.



Мне нужна помощь.

انا بحاجة الى مساعدة.

J'ai besoin d'aide.

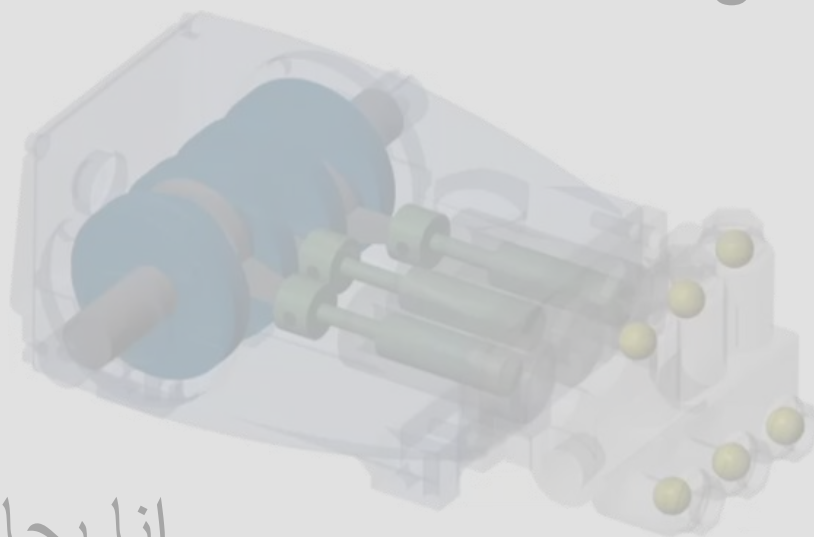
I need help.

Χρειάζομαι βοήθεια.

ฉันต้องการความช่วยเหลือ.

Ich brauche Hilfe.

Segítségre van szükségem.



Мне нужна помощь.

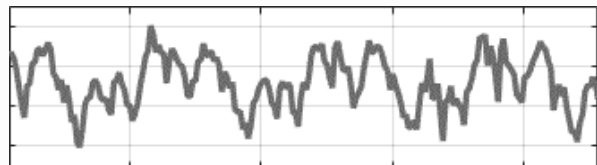
Necesito ayuda.

انا بحاجة الى مساعدة.

J'ai besoin d'aide.

I need help.

Χρειάζομαι βοήθεια.





English

Spanish

French

Pump - detected

▼



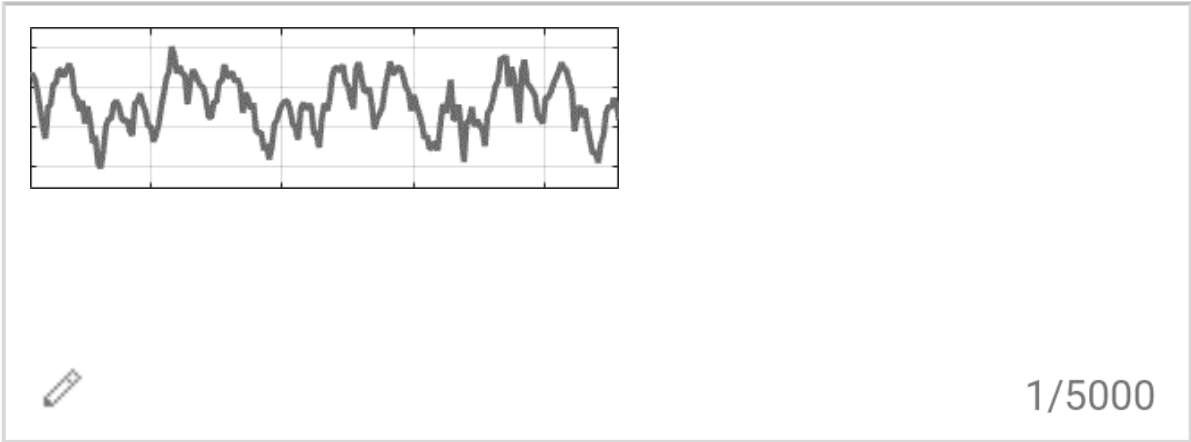
English

Russian

Greek

▼

Translate



I need help.

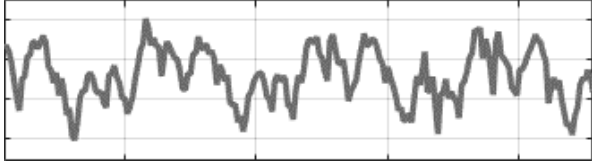


English Spanish French Pump - detected ▼



English Russian Greek ▼

Translate

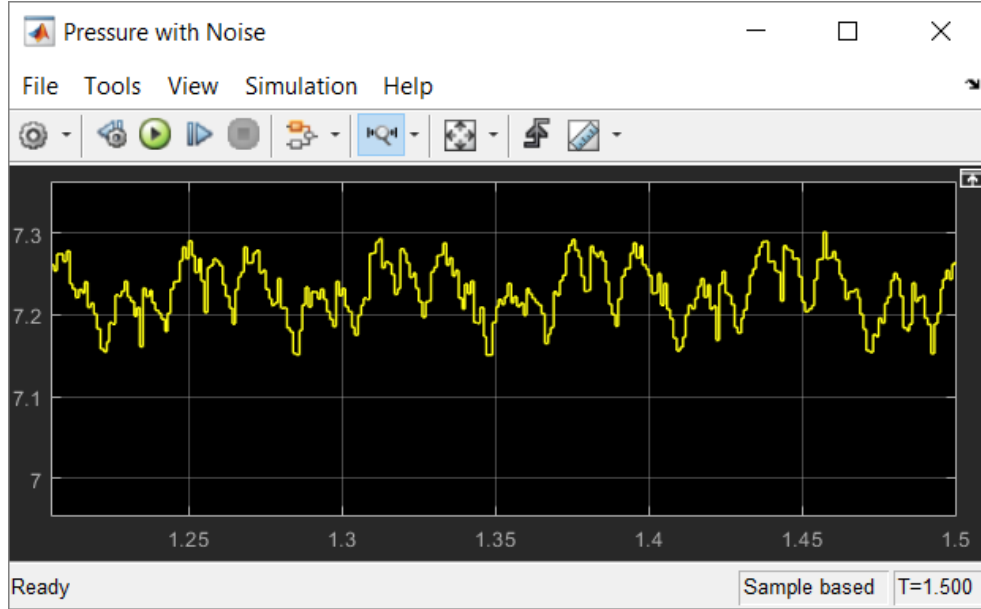


1/5000

**I need help. One of my seals is leaking. I will shut down your line in 5 days**

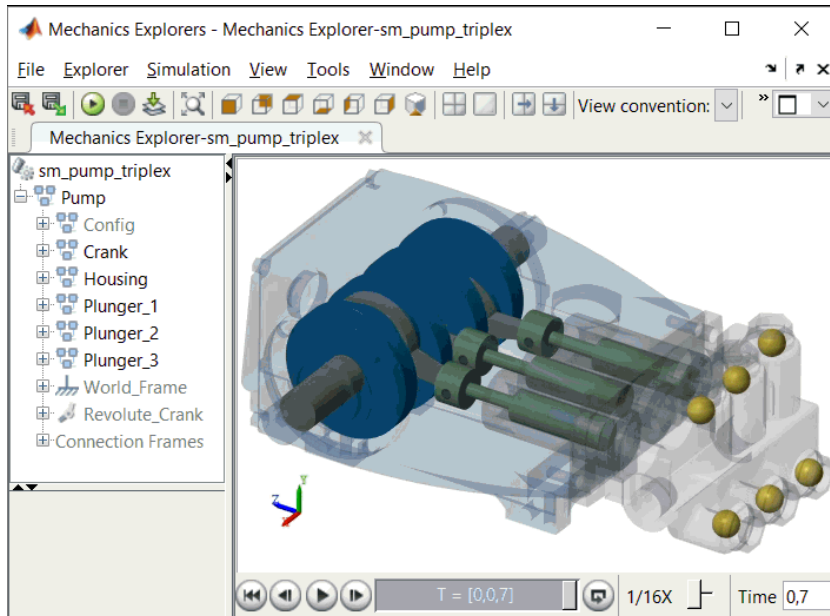
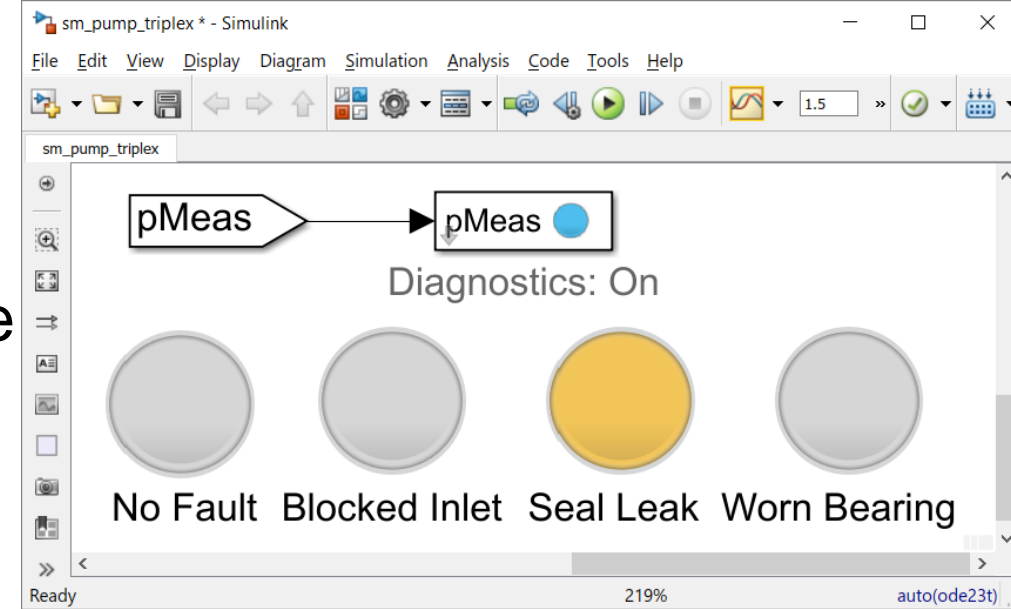


# Prevent system downtime



by sending  
sensor data

to a predictive  
maintenance  
algorithm



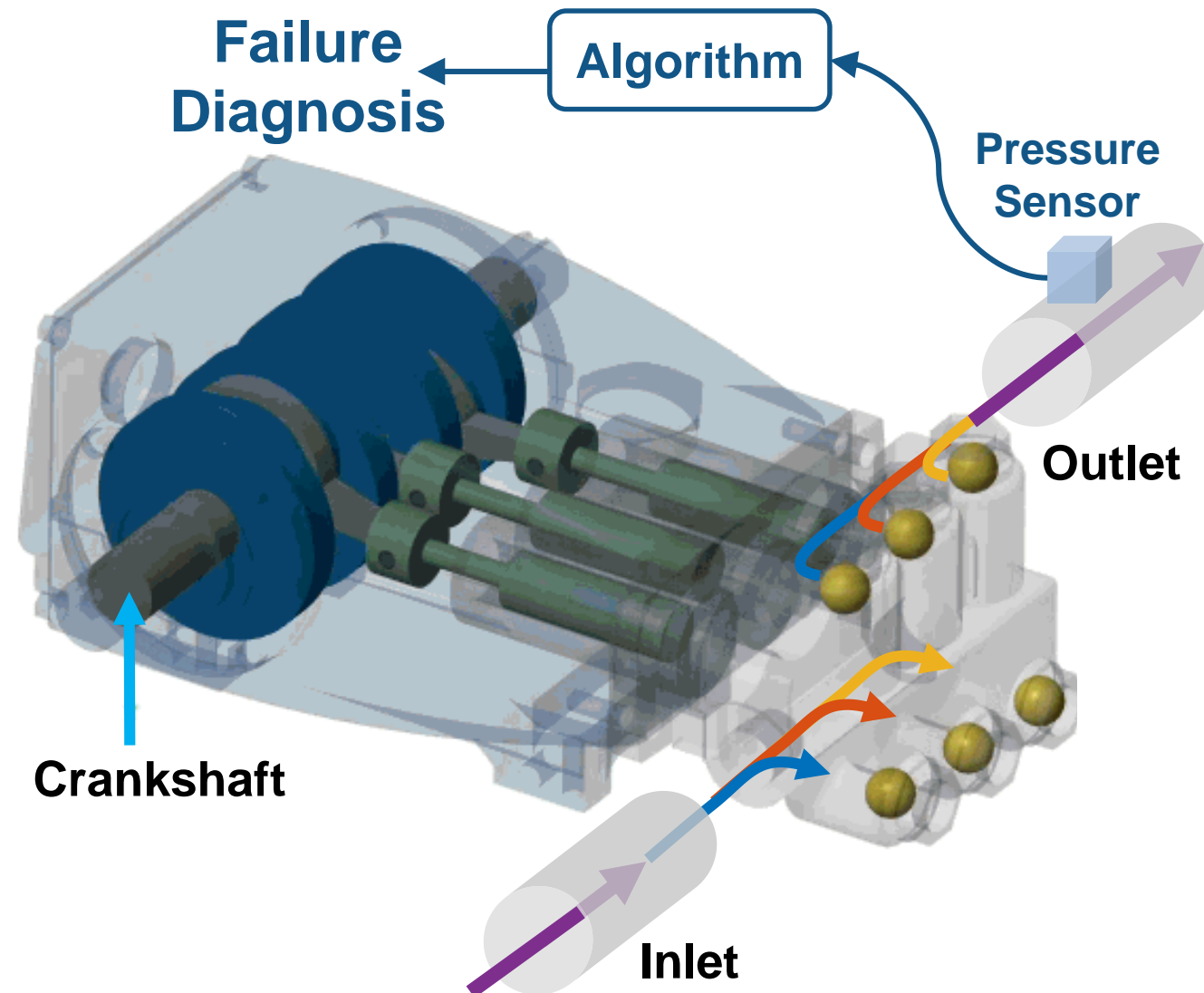
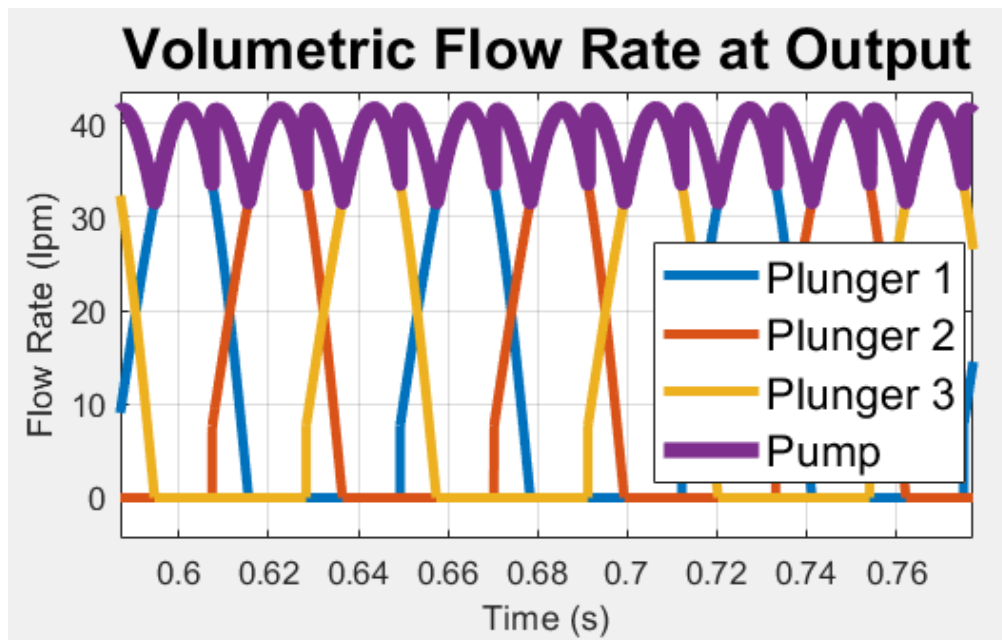
created using  
a Digital Twin

and machine  
learning  
in MATLAB.

		Model 1.18					
True class	Block P1	88%			12%		
	Block P1, Worn Bearing		100%				
	Leak P1			100%			
	Leak P1, Block P1	4%			96%		
	Leak P1, Worn Bearing					100%	
	Nominal						100%
	Worn Bearing						100%

# Triplex Pump

- Crankshaft drives three plungers
  - Each 120 degrees out of phase
  - One chamber always discharging
  - Smoother flow than single or duplex piston pumps









# Agenda

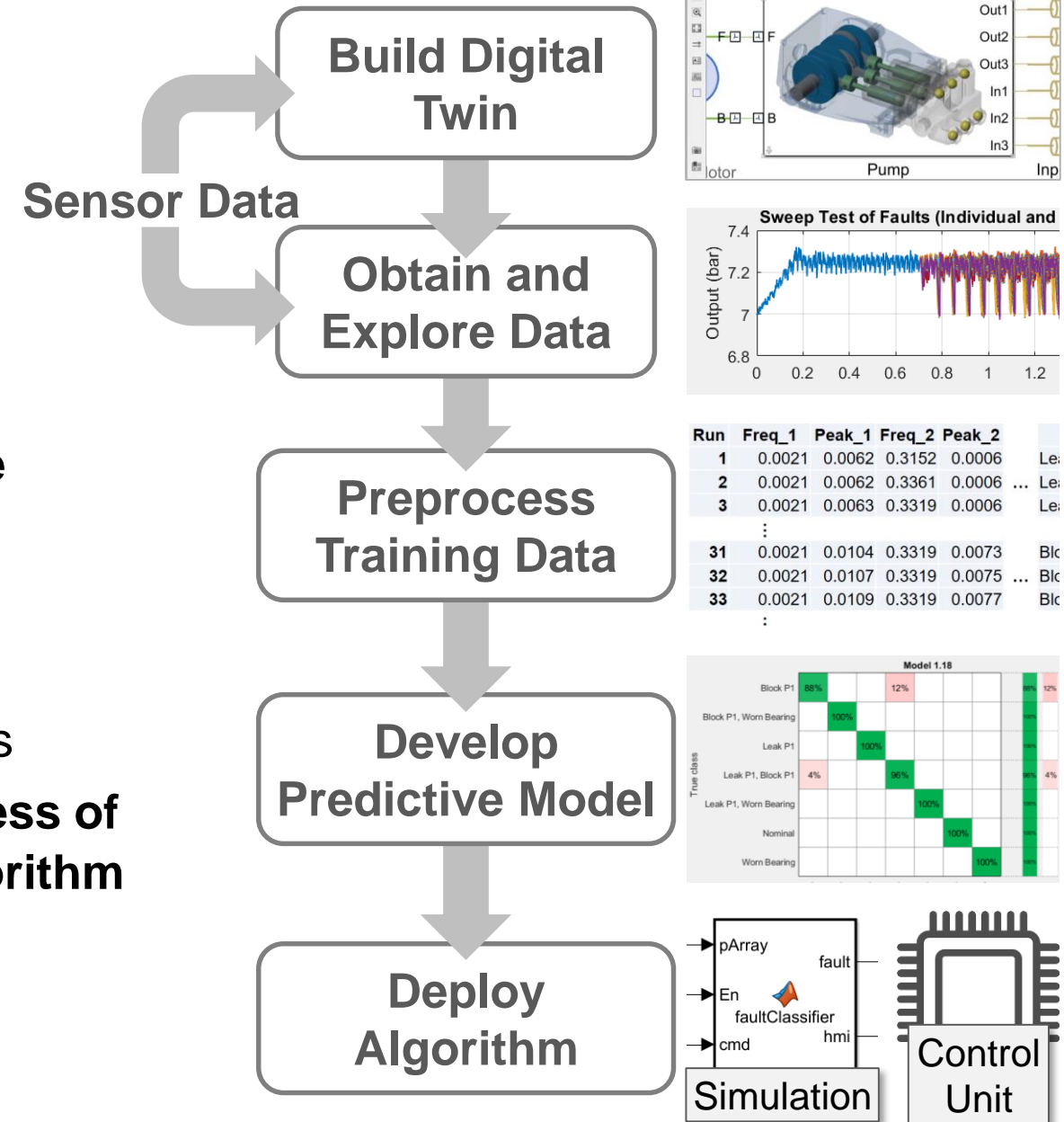
- Predictive Maintenance Workflow
- Build a Digital Twin
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  - Tune using measured data
- Create Predictive Model
  - Model component failure
  - Generate training data
  - Select and train classification model
- Deploy Fault Diagnostics Algorithm

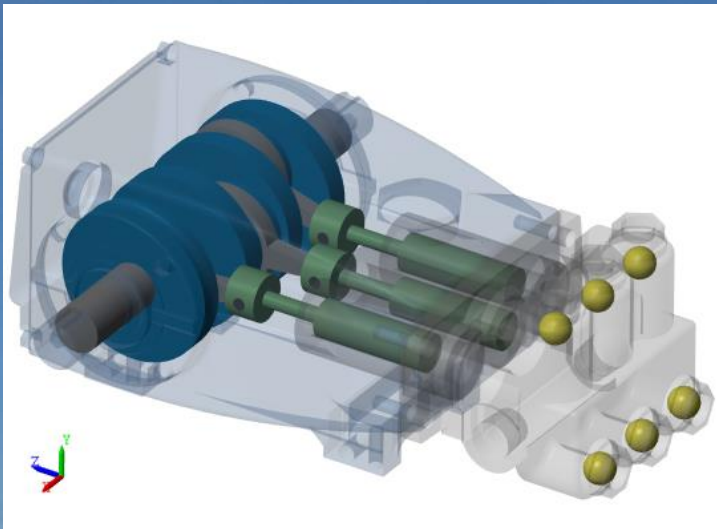
# Predictive Maintenance Workflow

- Sensor data isn't always available
  - Failure conditions difficult to reproduce
  - Time consuming or costly to generate

**Solution: Build digital twin and generate sensor data using simulation**
- Developing algorithm is complex
  - Requires complex concepts and analysis

**Solution: Use MATLAB to simplify process of developing and deploying algorithm**







# Baker Hughes Develops Predictive Maintenance Software for Gas and Oil Extraction Equipment Using Data Analytics and Machine Learning

## Challenge

Develop a predictive maintenance system to reduce pump equipment costs and downtime

## Solution

Use MATLAB to analyze nearly one terabyte of data and create a neural network that can predict machine failures before they occur

## Results

- Savings of more than \$10 million projected
- Development time reduced tenfold
- Multiple types of data easily accessed



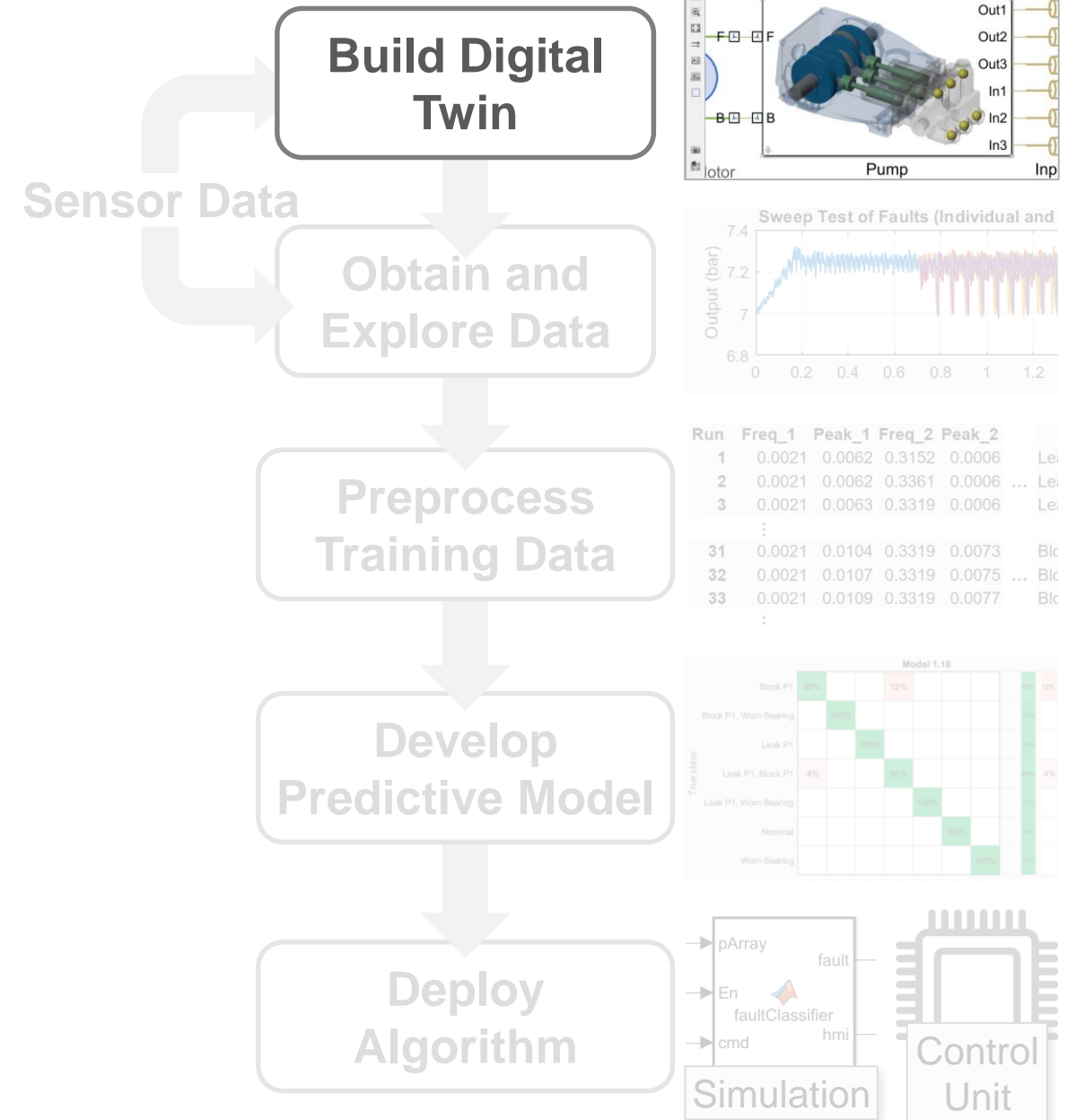
**Truck with positive displacement pump.**

*“MATLAB gave us the ability to convert previously unreadable data into a usable format; automate filtering, spectral analysis, and transform steps for multiple trucks and regions; and ultimately, apply machine learning techniques in real time to predict the ideal time to perform maintenance.”*

*- Gulshan Singh, Baker Hughes*

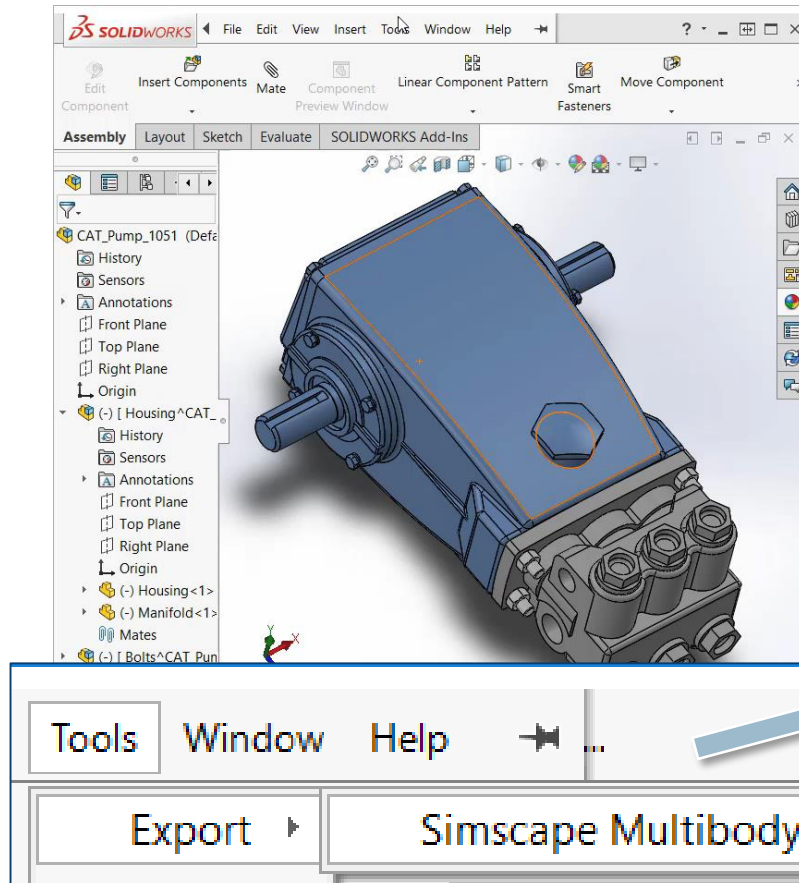
# Agenda

- Predictive Maintenance Workflow
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  - Tune using measured data
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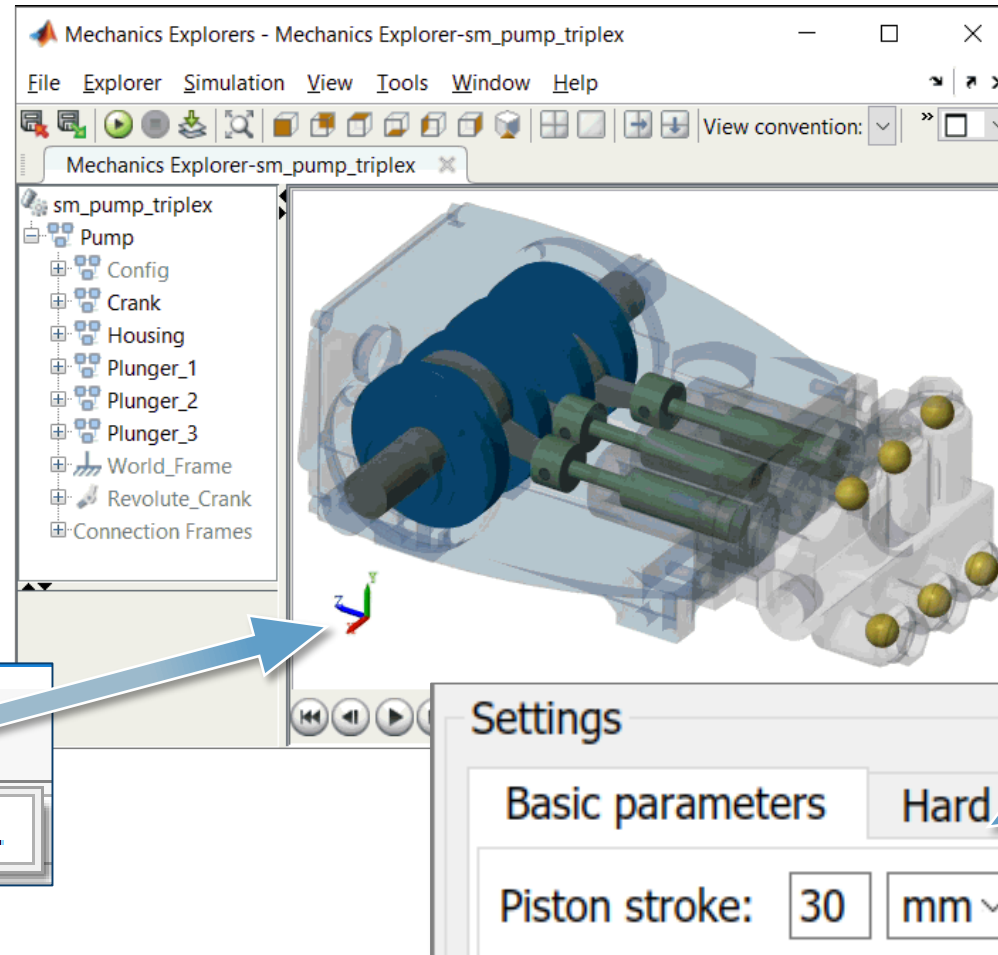


# Build Digital Twin of Hydraulic Pump

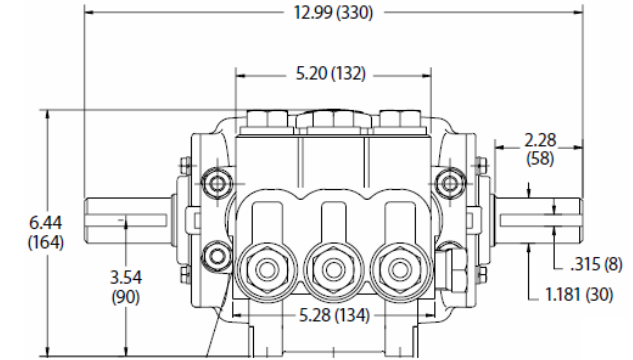
## Import CAD Data



## Digital Twin (Dynamic Model)



## Tune to Datasheet

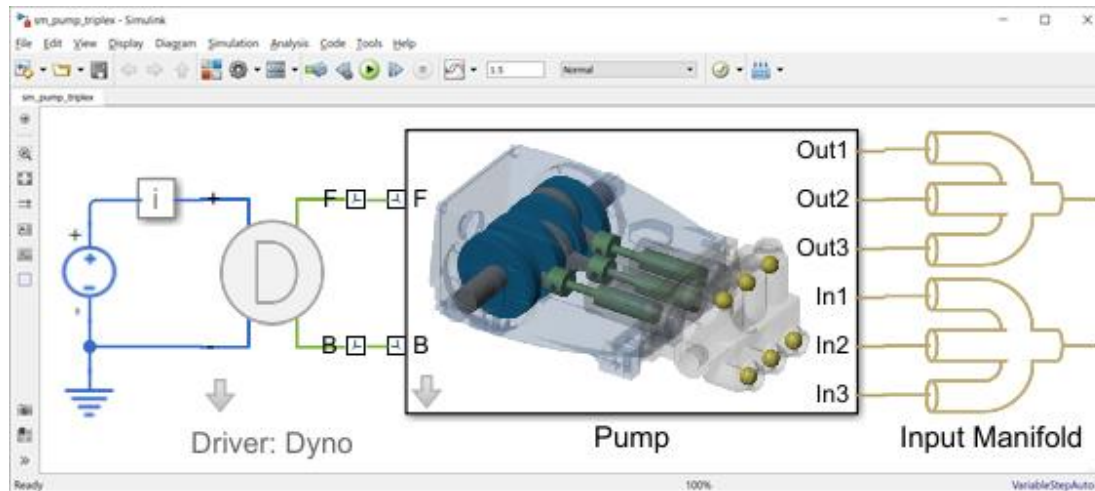


COMMON SPECIFICATIONS	U.S.	Metric
Bore	0.945"	24 mm
Stroke	1.18"	30 mm
Crankcase Capacity	42 oz.	1.26 l
Shaft Diameter	1.181"	30 mm



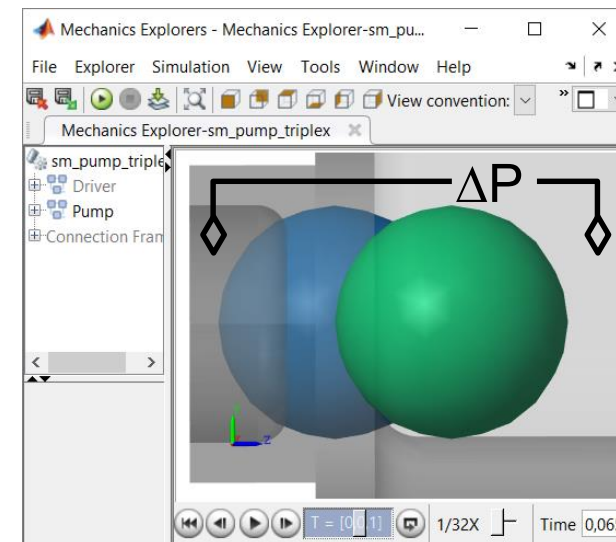
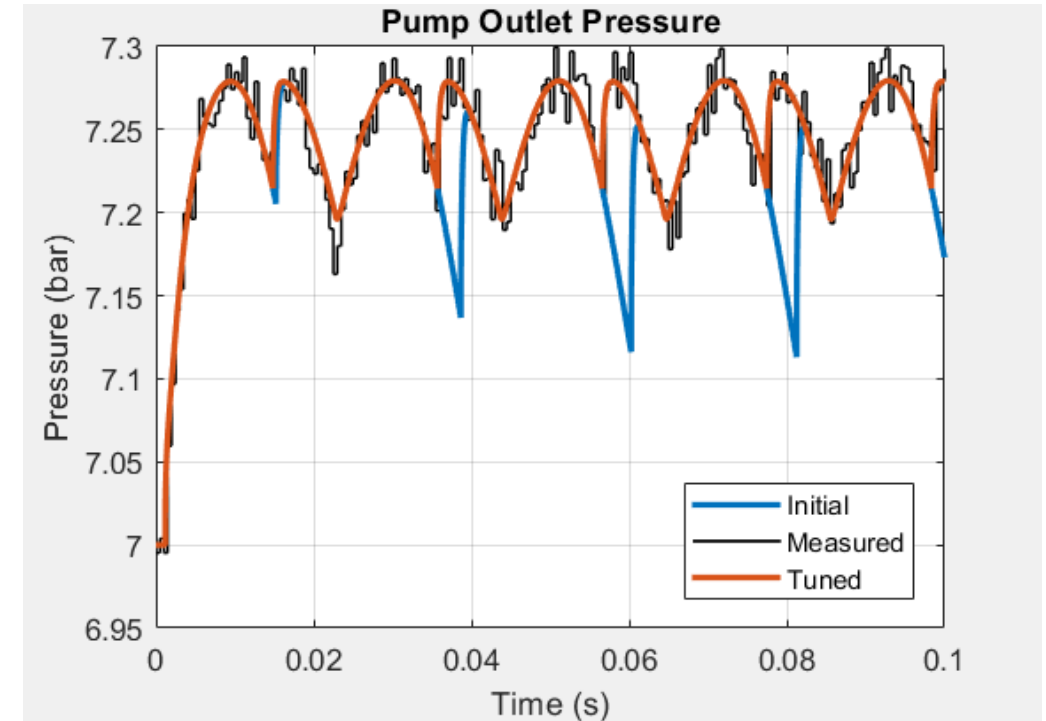
# Estimate Parameters Using Measured Data

## Model:

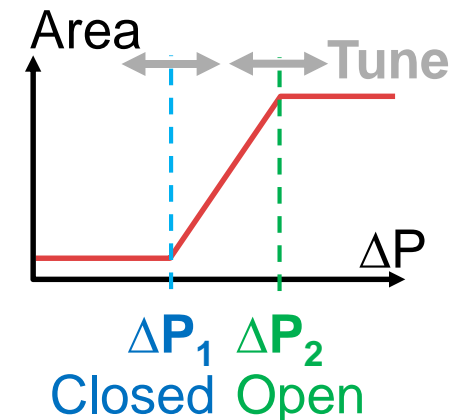


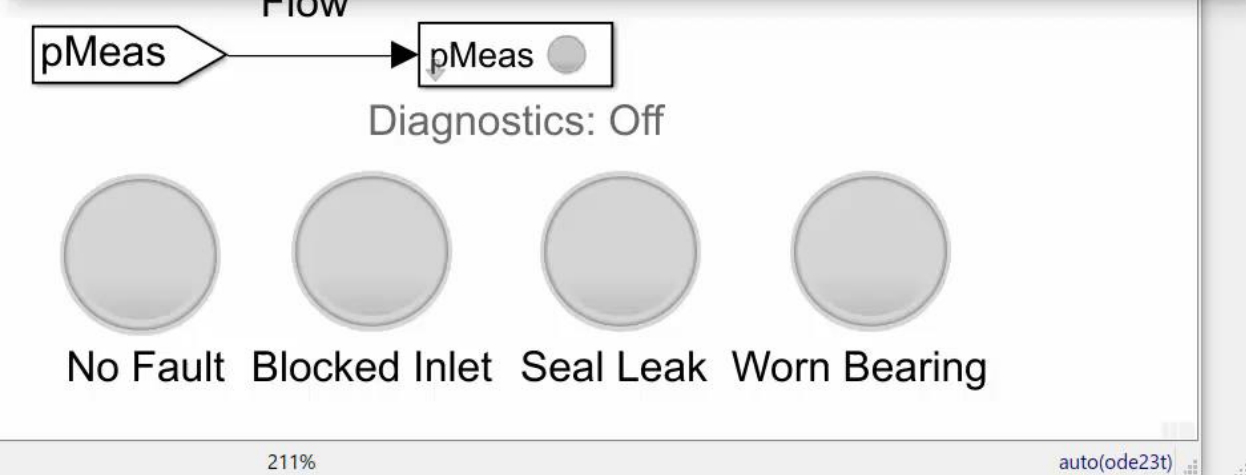
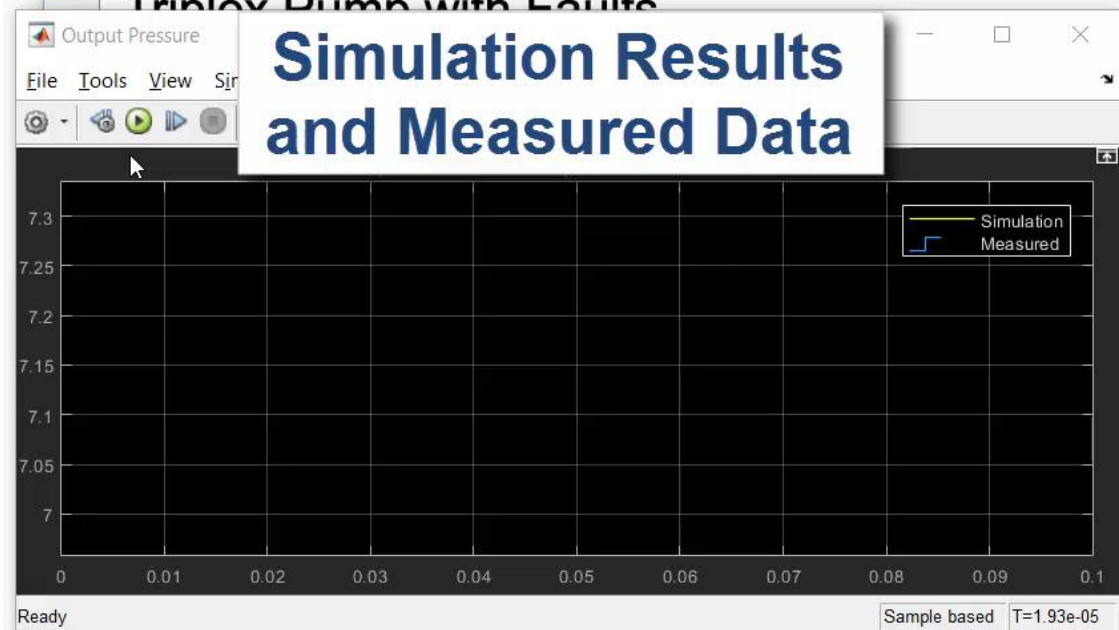
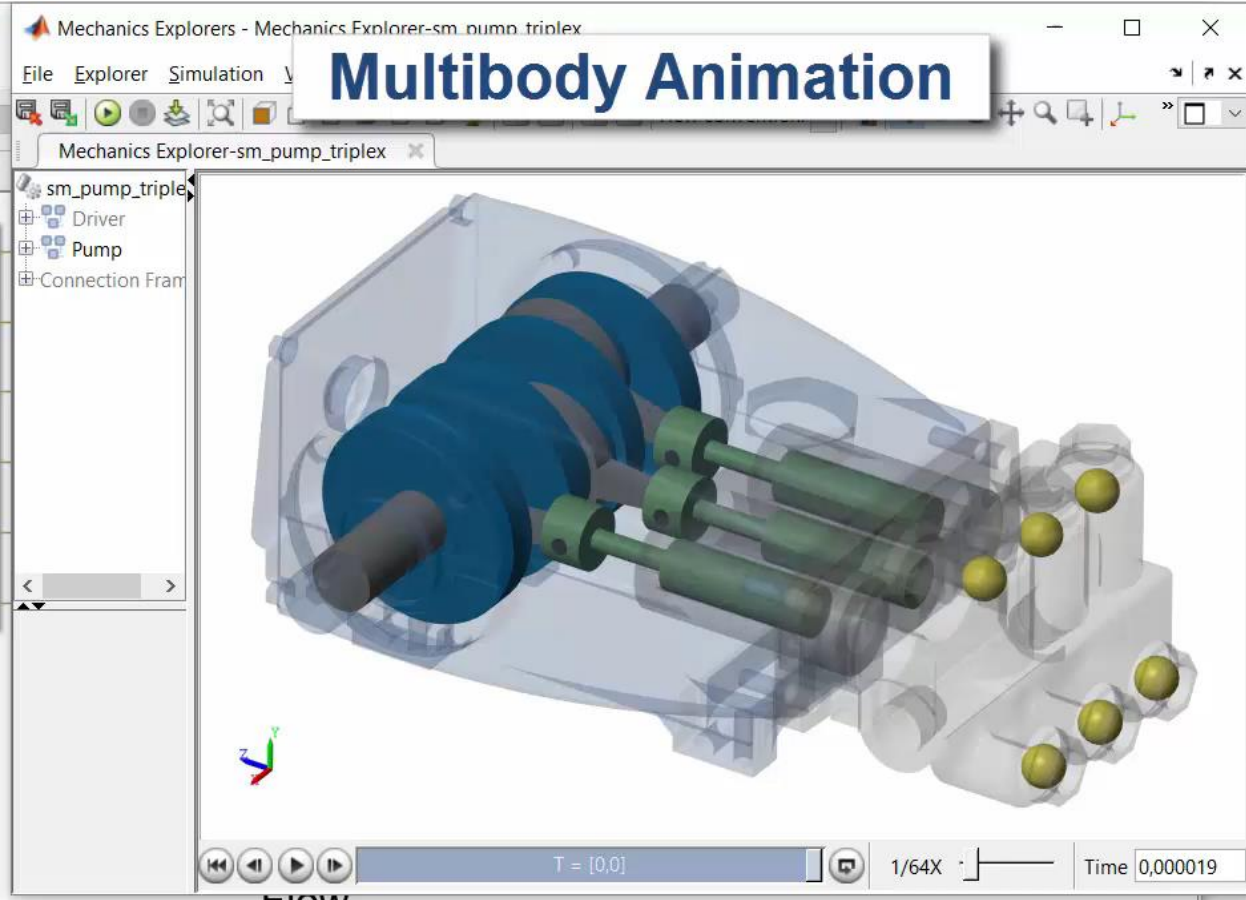
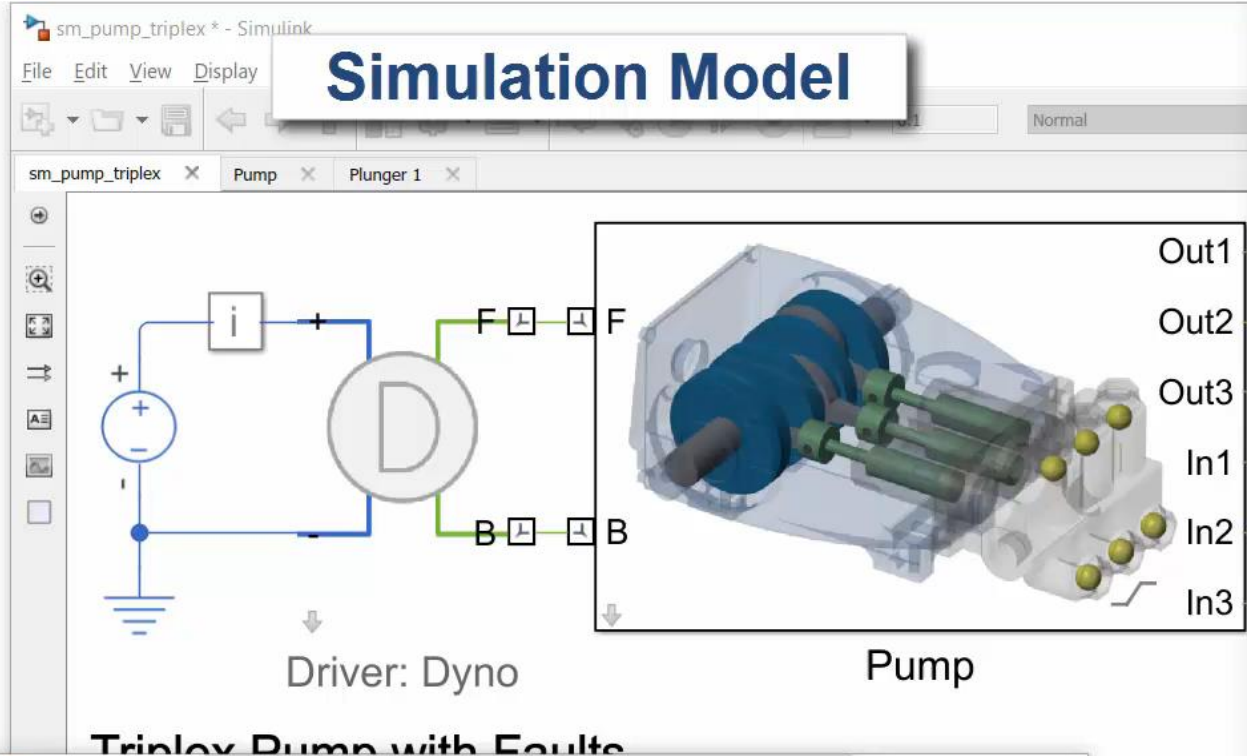
**Challenge:** Simulation results do not match behavior of real system

**Solution:** Use [Simulink Design Optimization](#) to automatically tune model parameters



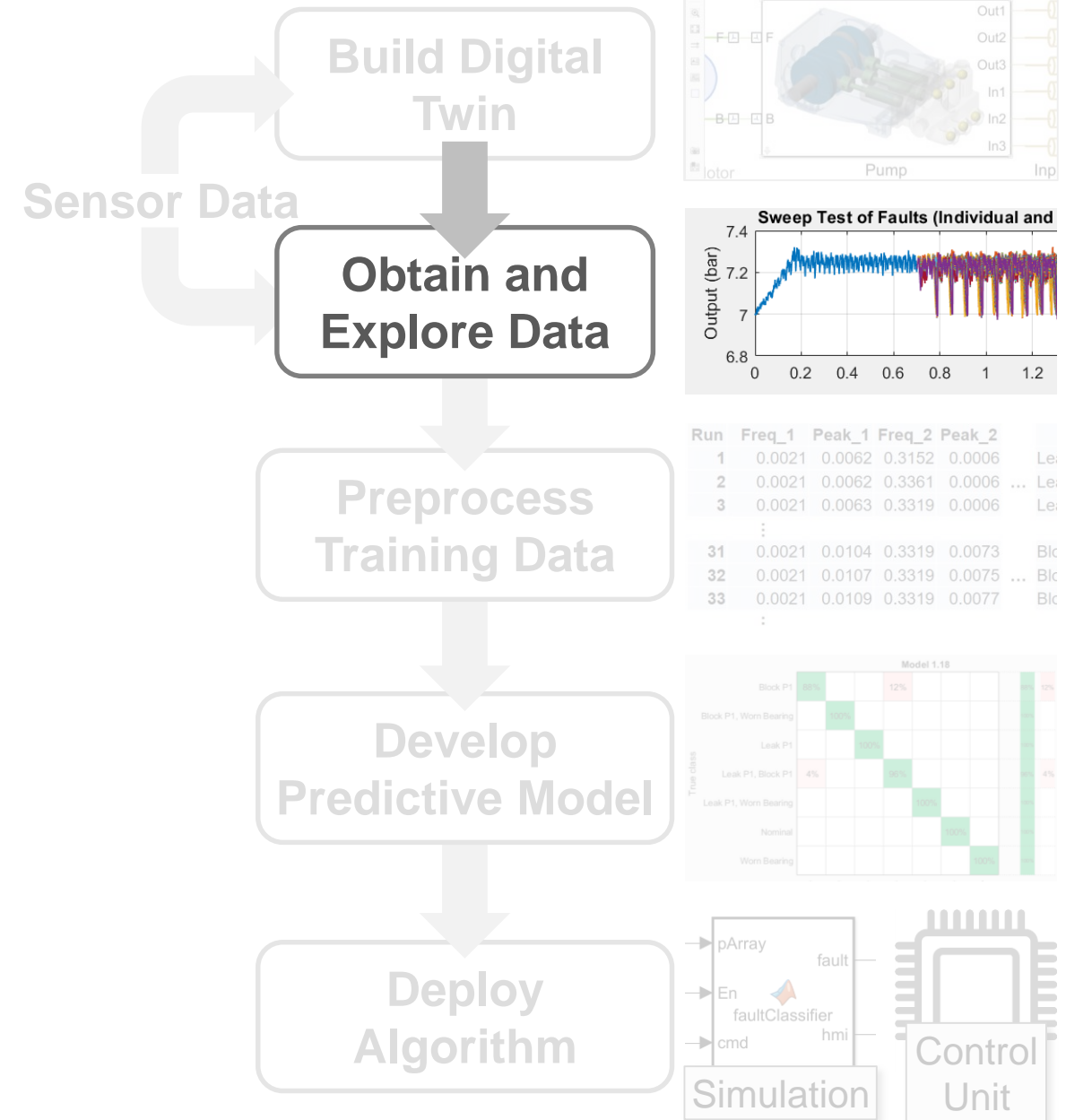
## Check Valve Characteristic





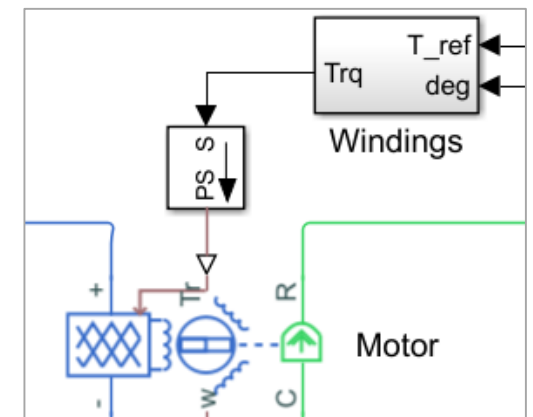
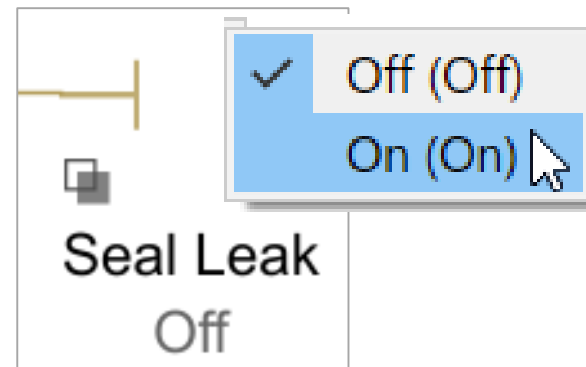
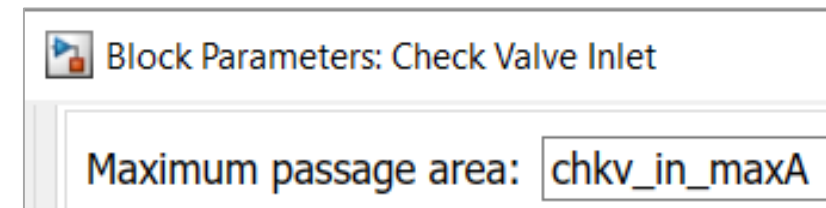
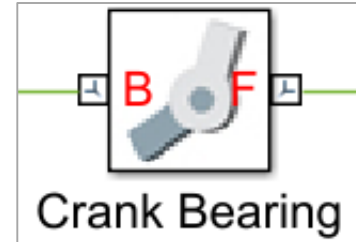
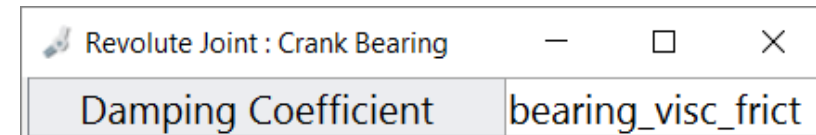
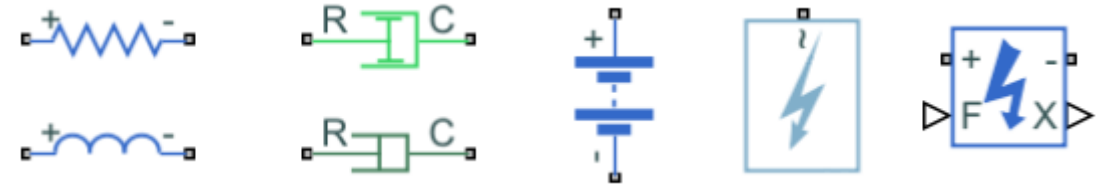
# Agenda

- Predictive Maintenance Workflow
- Build a Digital Twin
  - Model physical system
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- Create Predictive Model
  - Model component failure
  - Generate training data
  - Select and train classification model
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# Model Component Failure

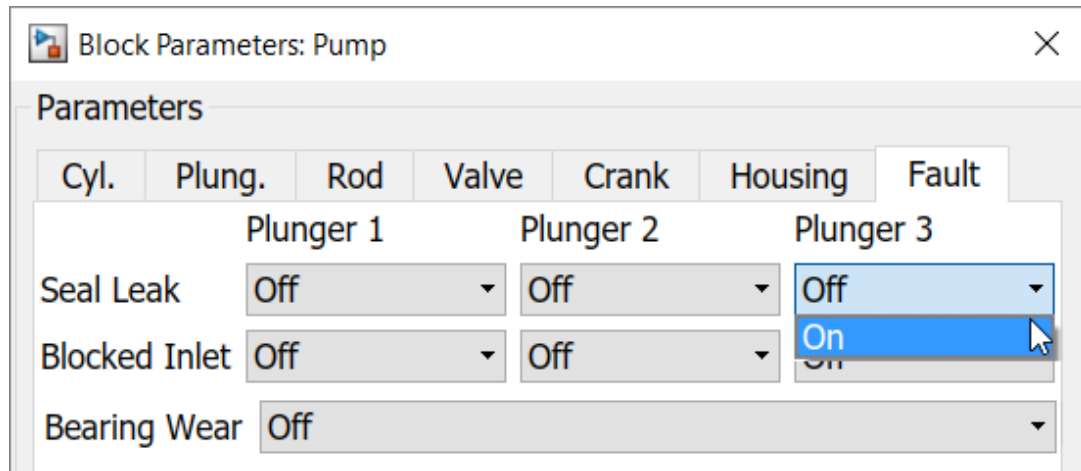
- Generic faults in many components
  - Short circuit, open circuit, friction, fade, etc.
  - Trigger based on time or conditions
- Adjust parameter values
  - Worn bearing adds friction
  - Blocked inlet has reduced passage area
- Adjust network
  - Seal leakage adds flow path
- Custom effects in Simulink
  - Broken winding applies no torque for 1/3 of every revolution





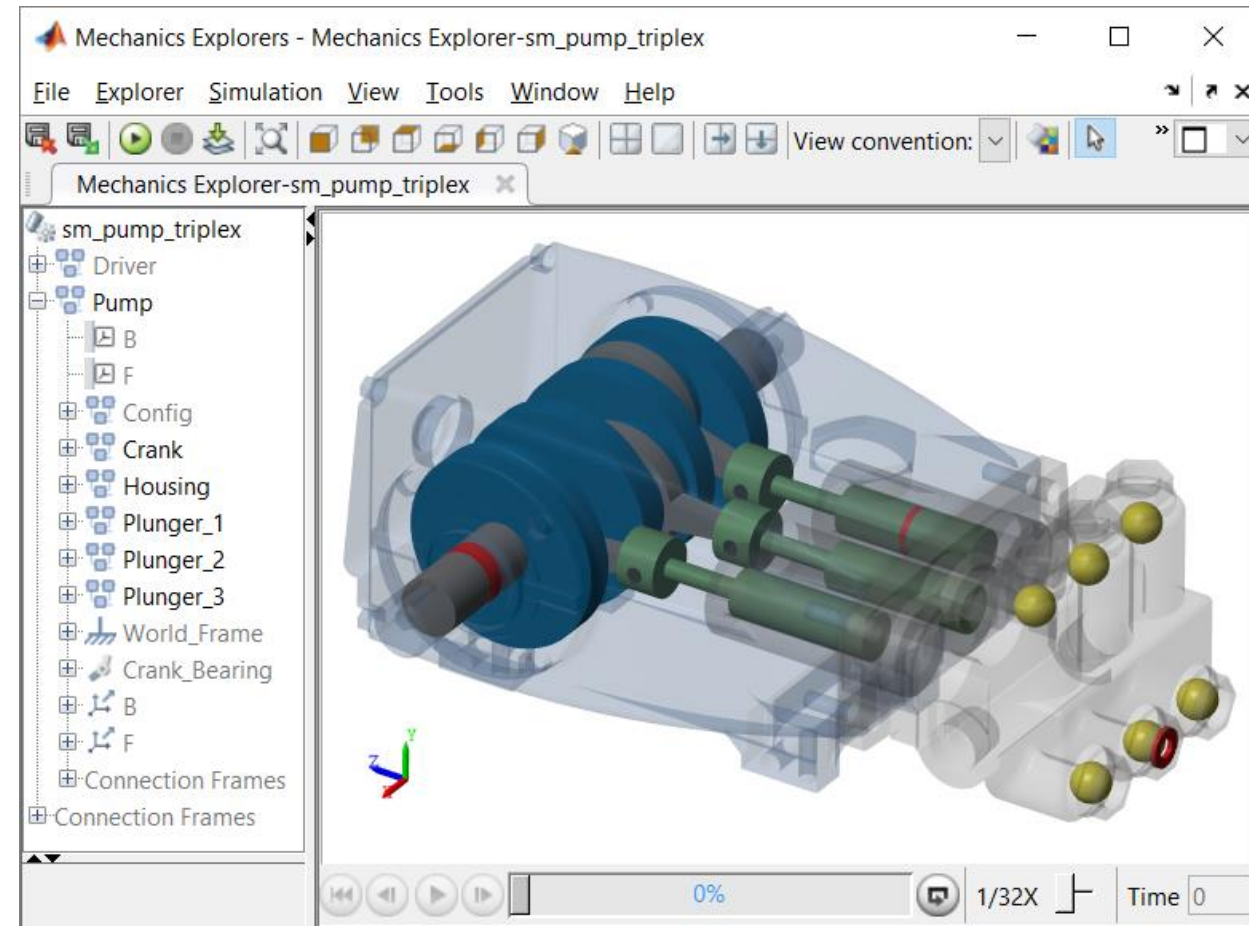
# Model Component Failure in Pump

- Enable from UI or MATLAB

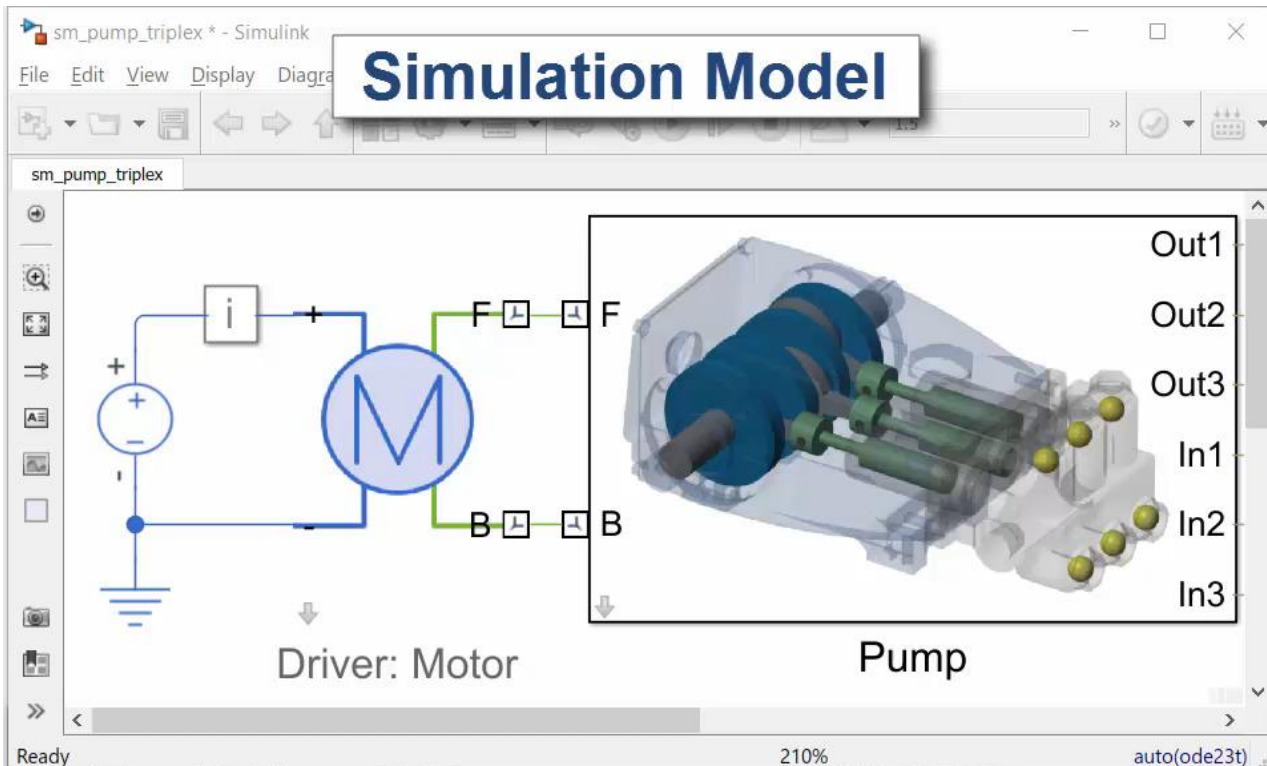


```
fx >> sm_pump_triplex_config_model...  
      ('sm_pump_triplex','Seal Leak','Off',1);
```

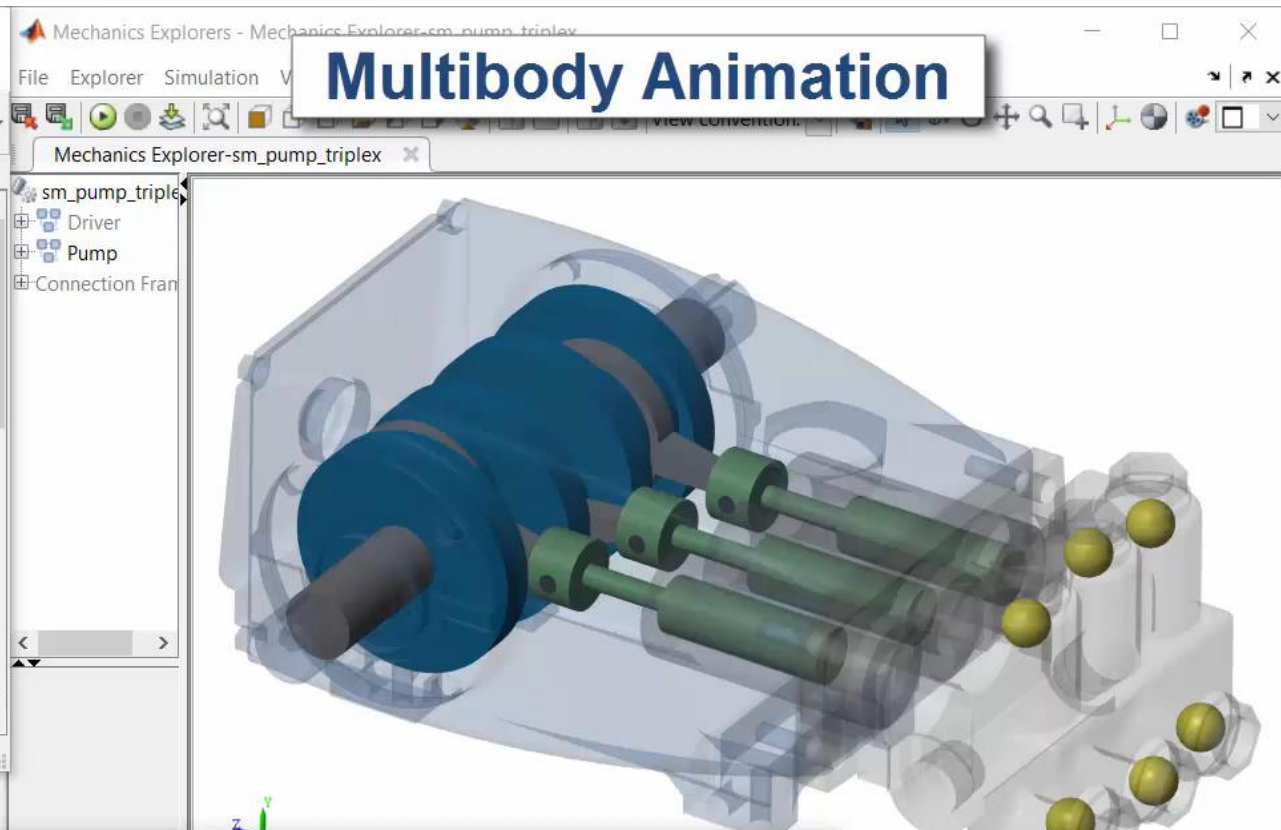
- Visual indication of fault



## Simulation Model



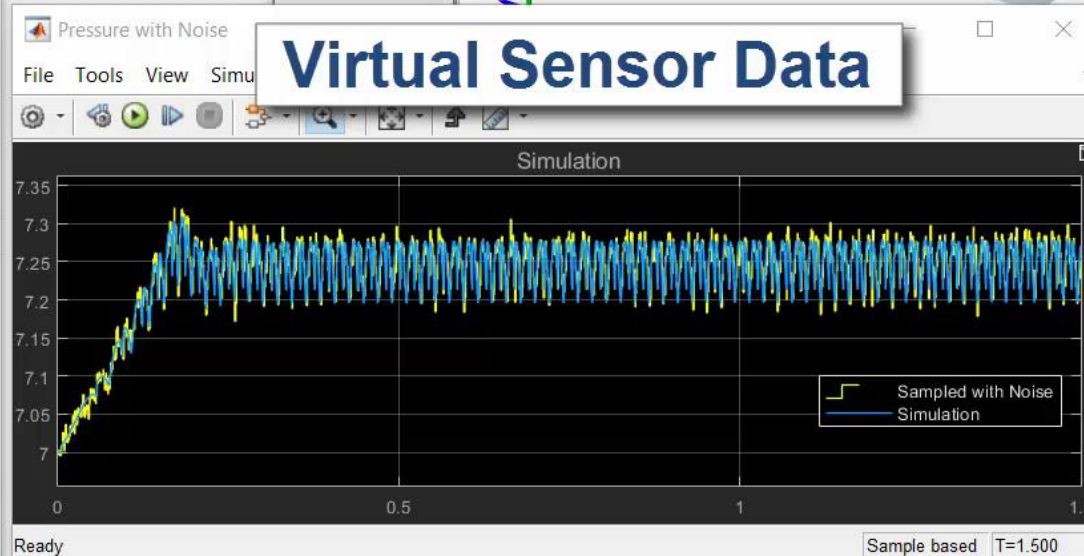
## Multibody Animation



## User Interface

All Faults Off	Plunger 1	Plunger 2	Plunger 3
Seal Leak	Off, On	Off, On	Off, On
Blocked Inlet	Off, On	Off, On	Off, On
Worn Bearing	Off, On		
Broken Winding	Off, On		

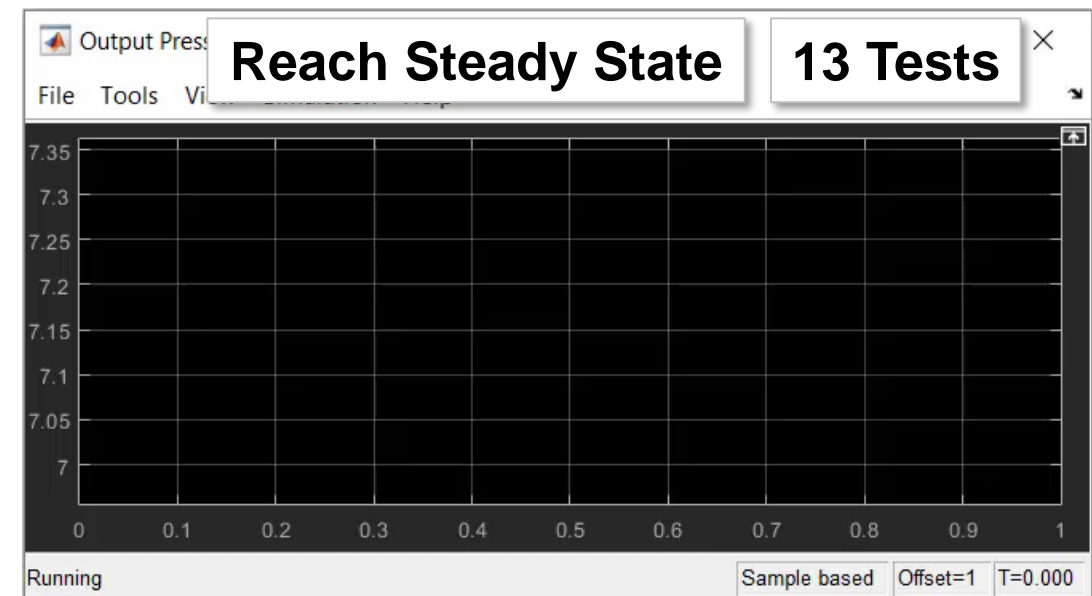
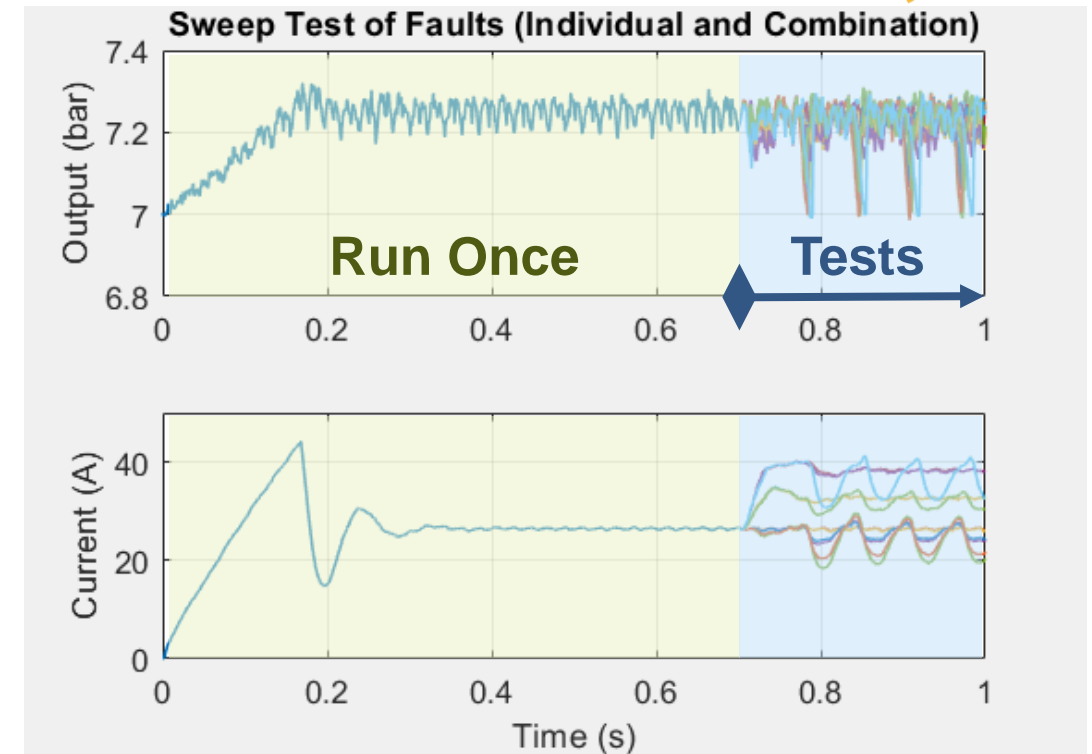
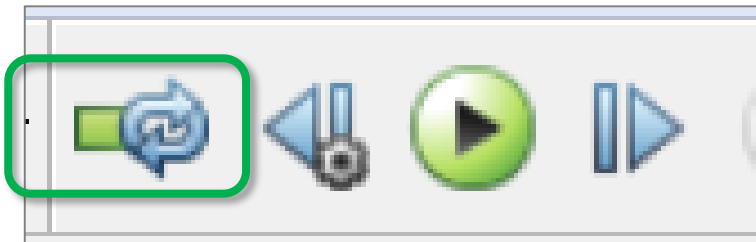
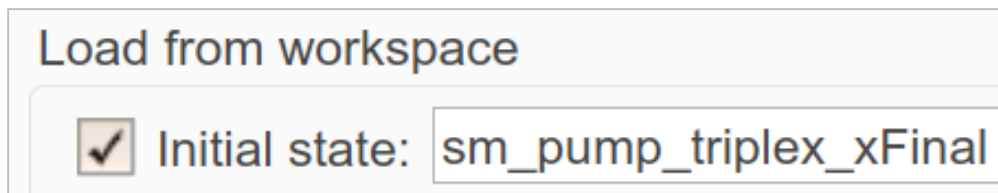
## Virtual Sensor Data





# Quickly Create Sensor Data Using Parallel Computing and Initial State

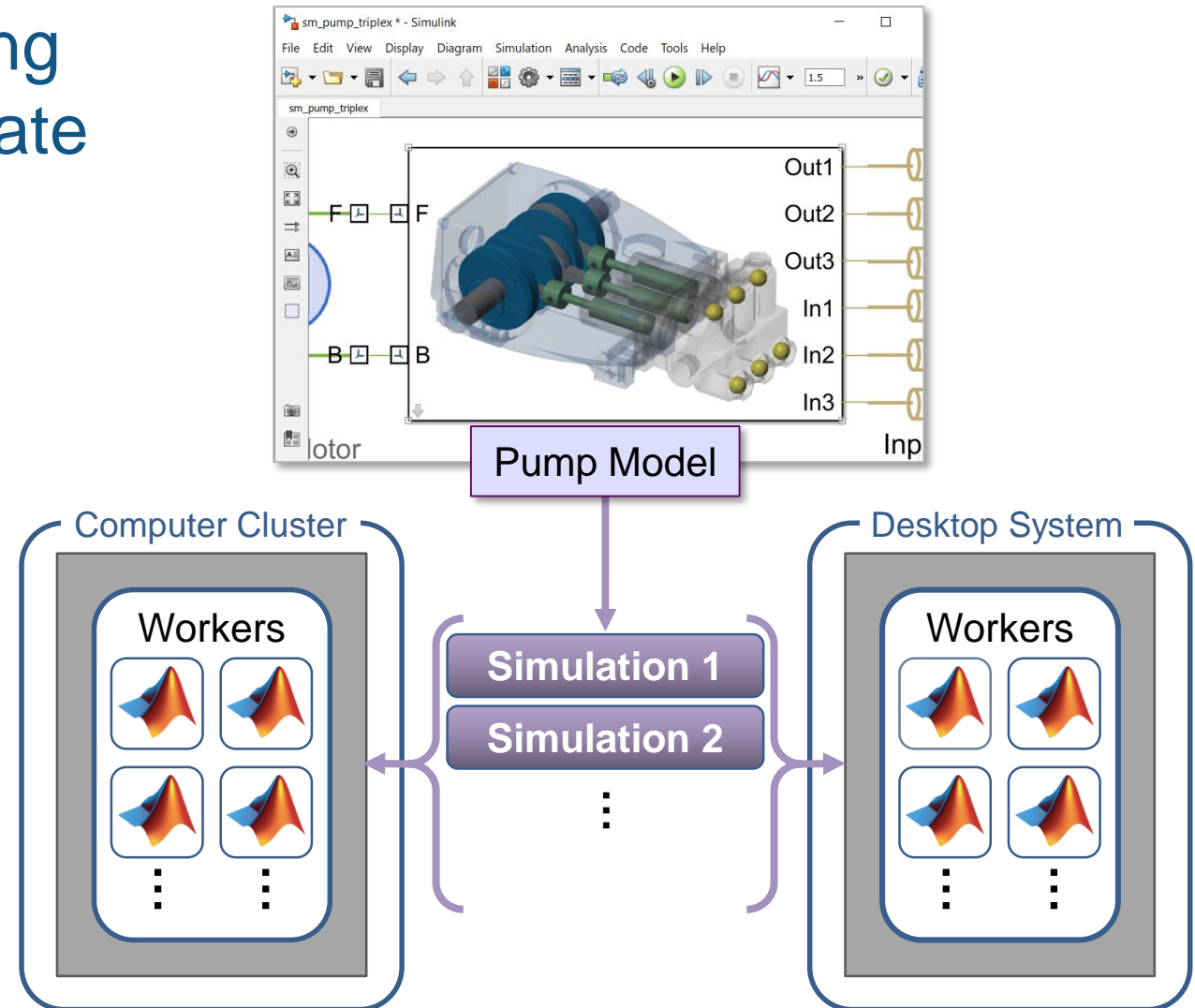
- Classification model requires data at various levels of failure for all fault combinations of interest
  - Many tests required
- Speed up tests
  - Start from steady state



# Quickly Create Sensor Data Using Parallel Computing and Initial State

- Classification model requires data at various levels of failure for all fault combinations of interest
  - Many tests required
- Speed up tests
  - Start from steady state
  - Run tests in parallel

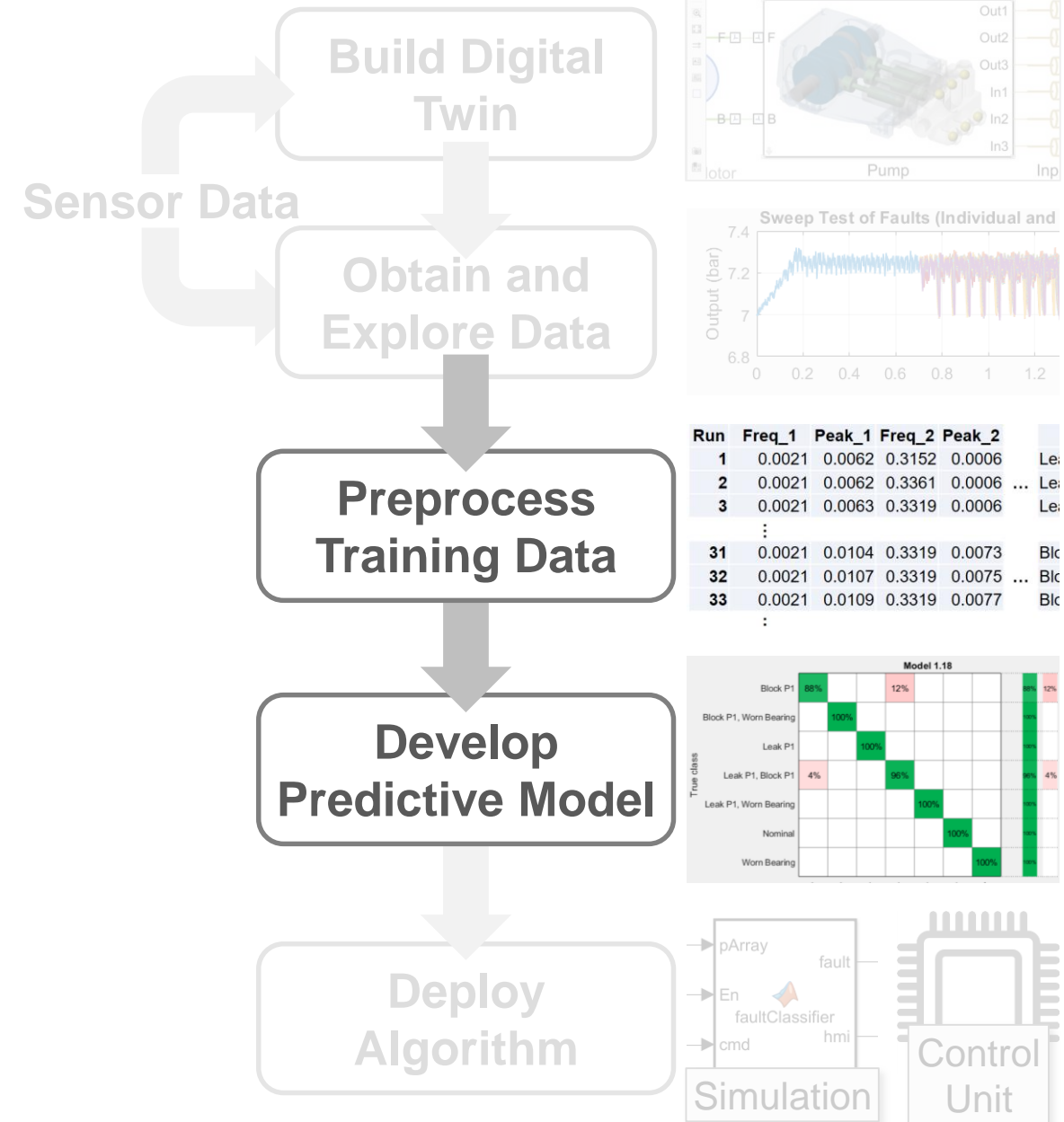
Distribute to multiple desktop workers or across a computing cluster



Running simulations in parallel speeds up your testing process.

# Agenda

- Predictive Maintenance Workflow
- Build a Digital Twin
  - Model physical system
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- Create Predictive Model
  - Model component failure
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  - Select and train classification model
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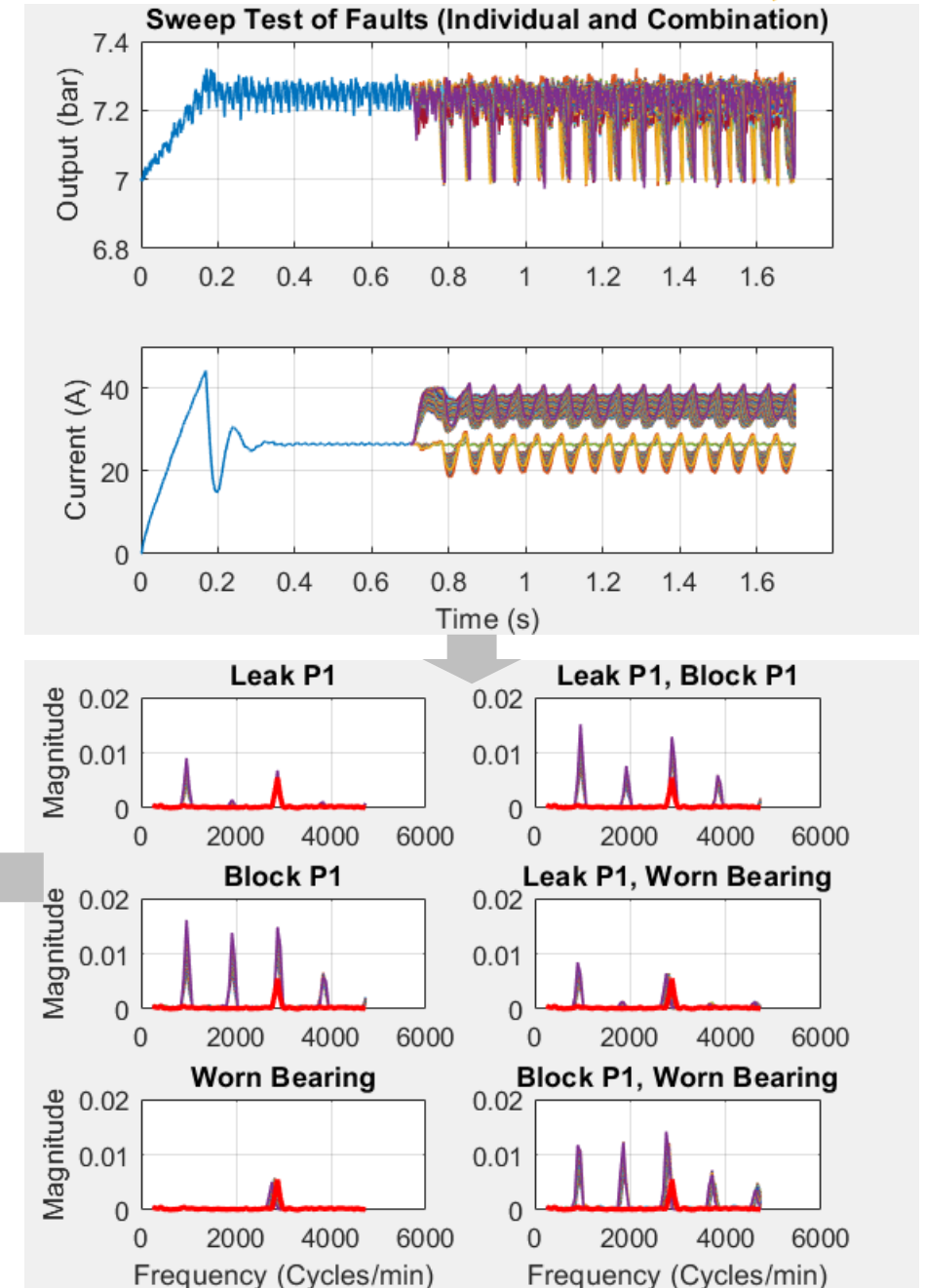


# Extract Training Data from Sensor Measurements

- Perform FFT on results
  - Save frequencies, magnitudes, fault type

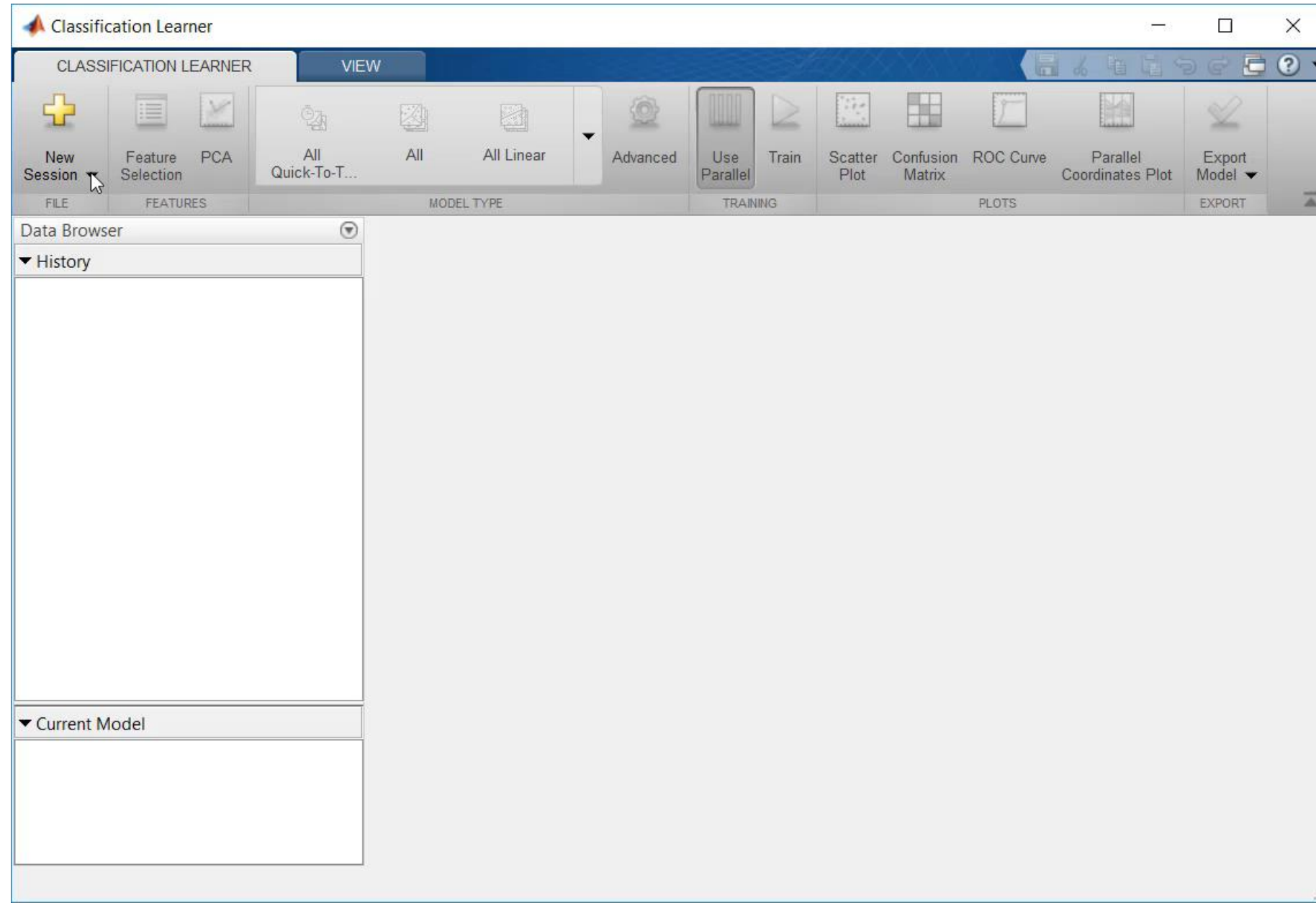
## Classification Model Training Data

Run	Freq_1	Peak_1	Freq_2	Peak_2	Fault
1	0.0021	0.0062	0.3152	0.0006	Leak_P1
2	0.0021	0.0062	0.3361	0.0006	...
3	0.0021	0.0063	0.3319	0.0006	Leak_P1
⋮	⋮	⋮	⋮	⋮	⋮
31	0.0021	0.0104	0.3319	0.0073	Block_P1
32	0.0021	0.0107	0.3319	0.0075	...
33	0.0021	0.0109	0.3319	0.0077	Block_P1
⋮	⋮	⋮	⋮	⋮	⋮
91	0.0021	0.0092	0.3319	0.0042	Leak P1, Block P1
92	0.0021	0.0095	0.3319	0.0044	...
93	0.0021	0.0097	0.3319	0.0045	Leak P1, Block P1
⋮	⋮	⋮	⋮	⋮	⋮
181	0.0021	0.0055			Nominal



# Evaluate all Classification Models

- Select data for training
- Train classifiers
- Evaluate results
- Export trained classifier for testing in Digital Twin



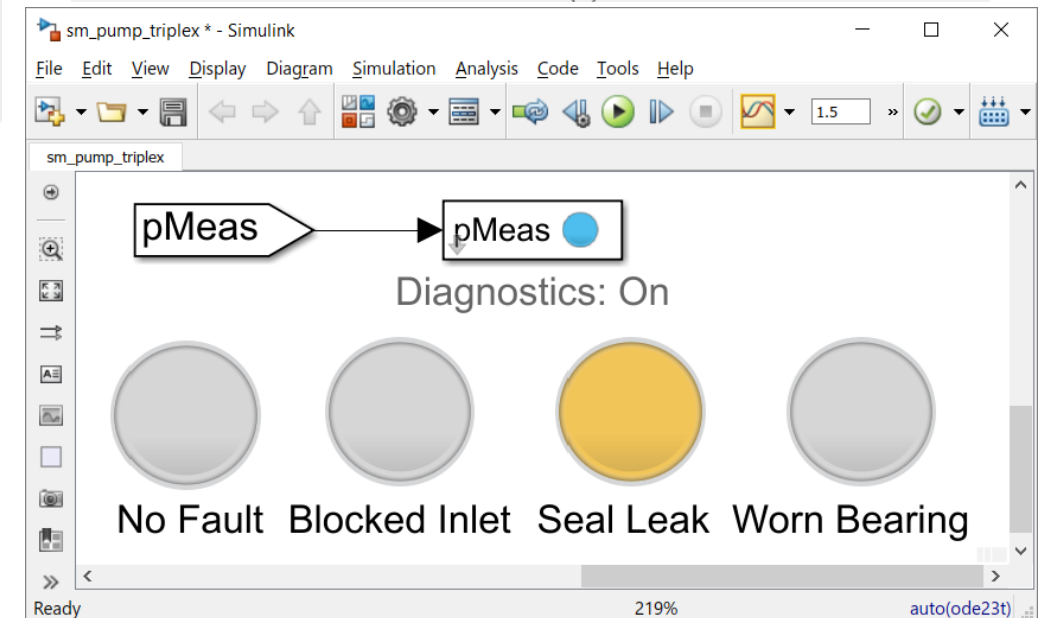
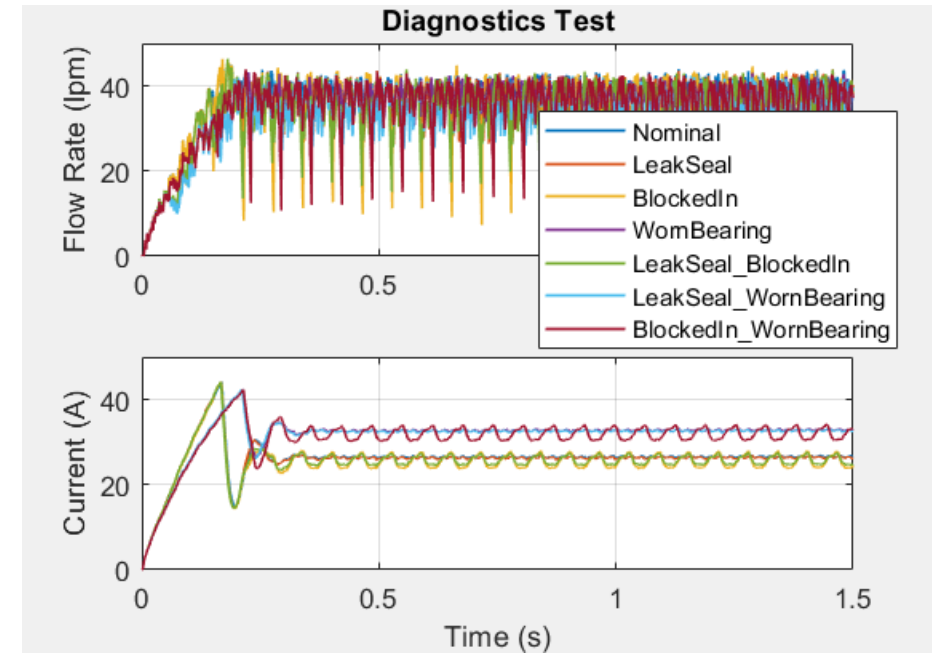
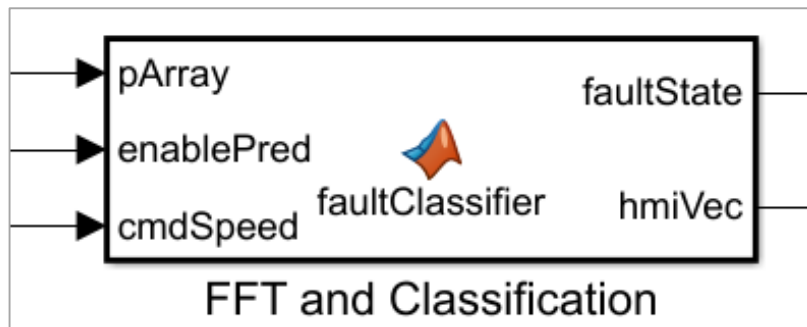
# Test Algorithm in Simulation

- Connect trained algorithm to digital twin
  - Verify behavior on new scenarios before deploying in embedded code



## Export Model

Export the currently selected model in the History list to the workspace to make predictions with new data





- Test and update algorithm when any aspect of system changes
  - Environment: temperature, fluid, power source
  - Supplier: Seals, valves, tolerances, material
  - Design: Larger, smaller, new markets
- Improve algorithm with new data
  - Tune digital twin with field data, automatically update algorithm



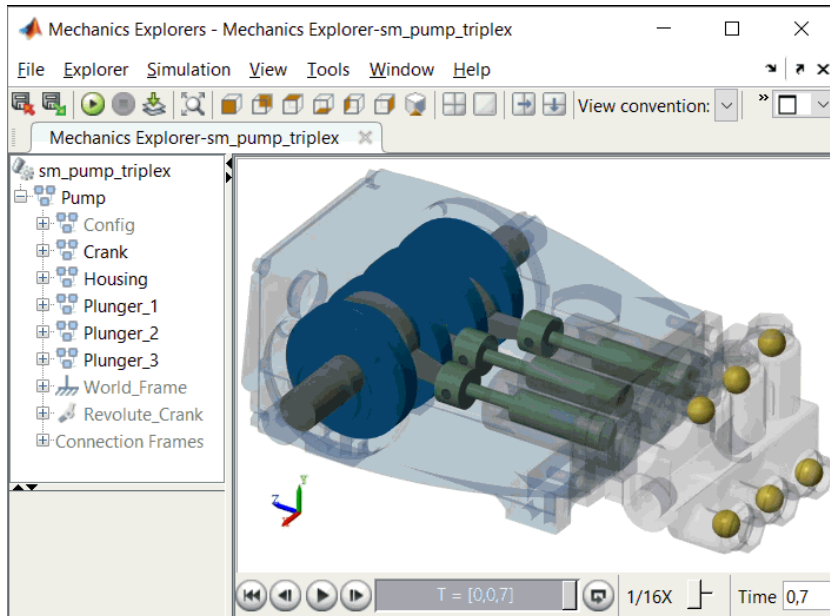
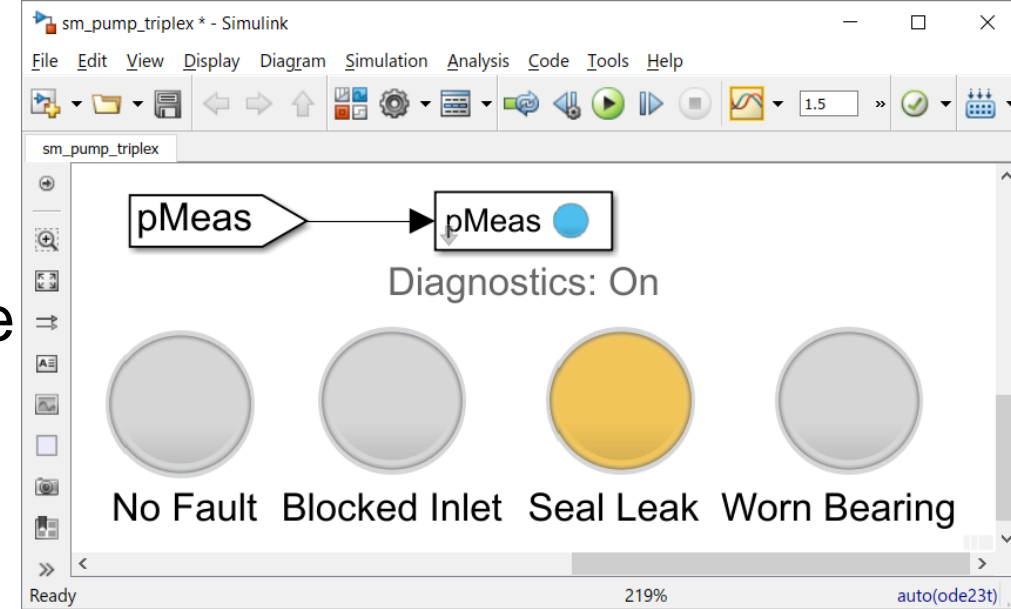
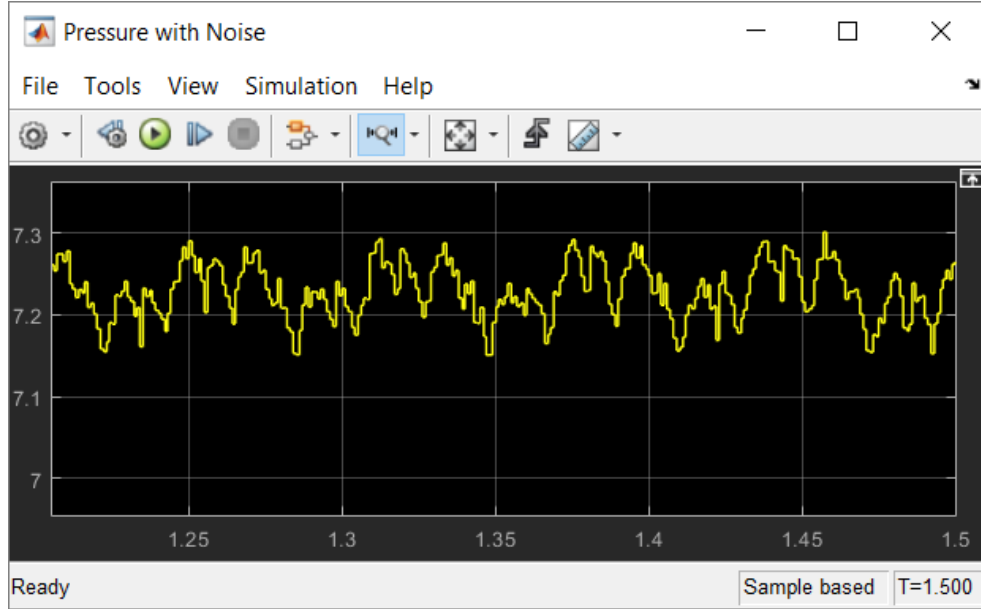
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Model 1.18

Block P1	88%			12%		
Block P1, Worn Bearing		100%				
Leak P1			100%			
Leak P1, Block P1	4%			96%		
Leak P1, Worn Bearing					100%	
Nominal						100%
Worn Bearing						100%

True class