

# Accurate Robotic Machining: Challenges and Solution Approaches

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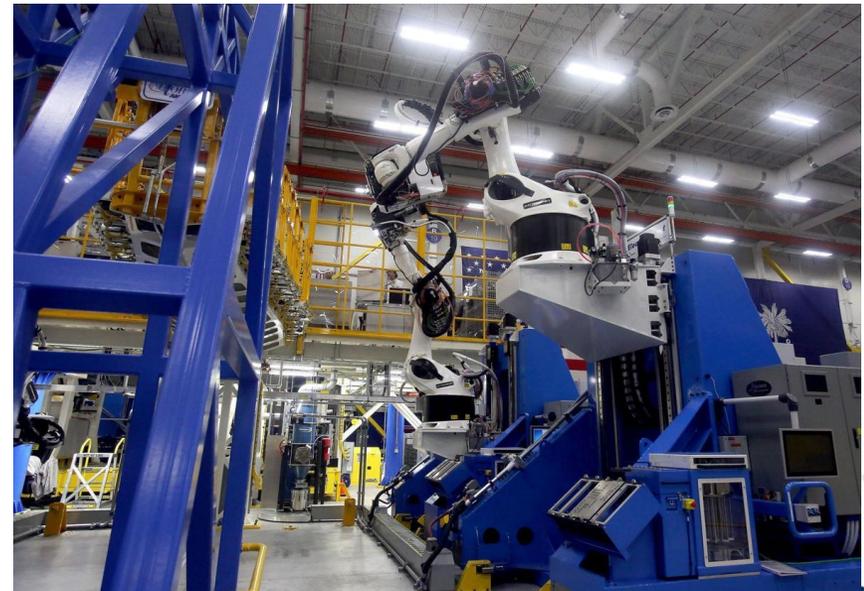
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# Accurate Industrial Robotics for Manufacturing

- Automation of large parts manufacturing dominated by monumental gantry systems
- Serial manipulators present an attractive alternative:
  - Cheaper
  - More flexible (reconfigurable)
  - More compliant
  - Less accurate
- <2% of industrial robots used for material removal (Verl et al. 2019)



Source: <https://www.electroimpact.com/Products/Composites/Overview.aspx>



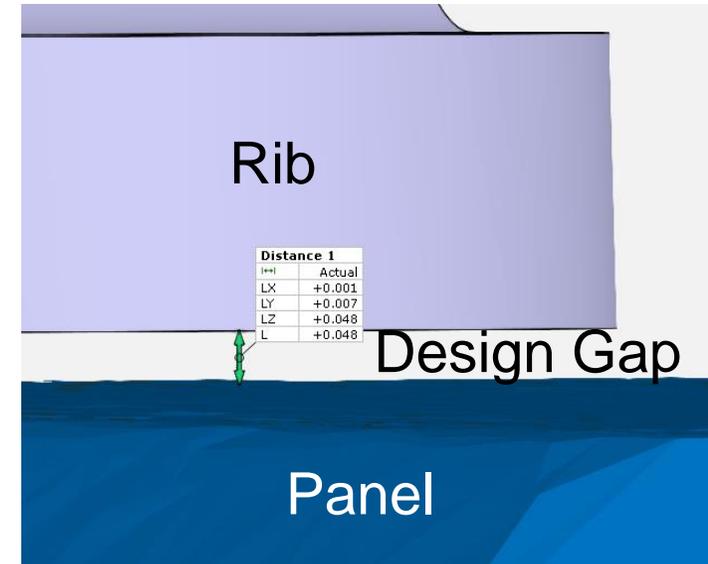
Source: [https://www.postandcourier.com/business/new-robots-to-help-boeing-s-c-meet-production-hike/article\\_b42c9b5b-3a40-5326-bcf4-ed3bacb83a06.html](https://www.postandcourier.com/business/new-robots-to-help-boeing-s-c-meet-production-hike/article_b42c9b5b-3a40-5326-bcf4-ed3bacb83a06.html)

# Example Application

## Aircraft Assembly

Major structural joints for aerospace have tolerances on the order of  $<.125$  mm

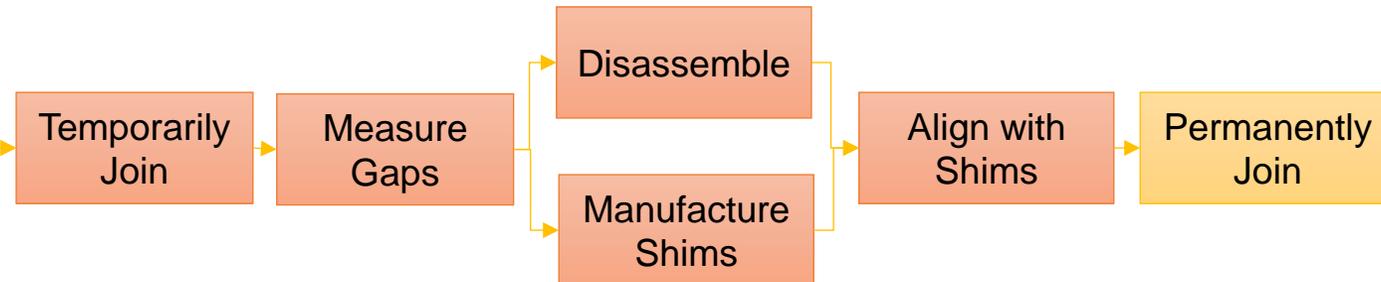
Current process is performed by hand which is time consuming, iterative, and causes inefficient production flow



Wing Skin



Wing Rib



# Potential Solution

## Shim-Less Aircraft Assembly

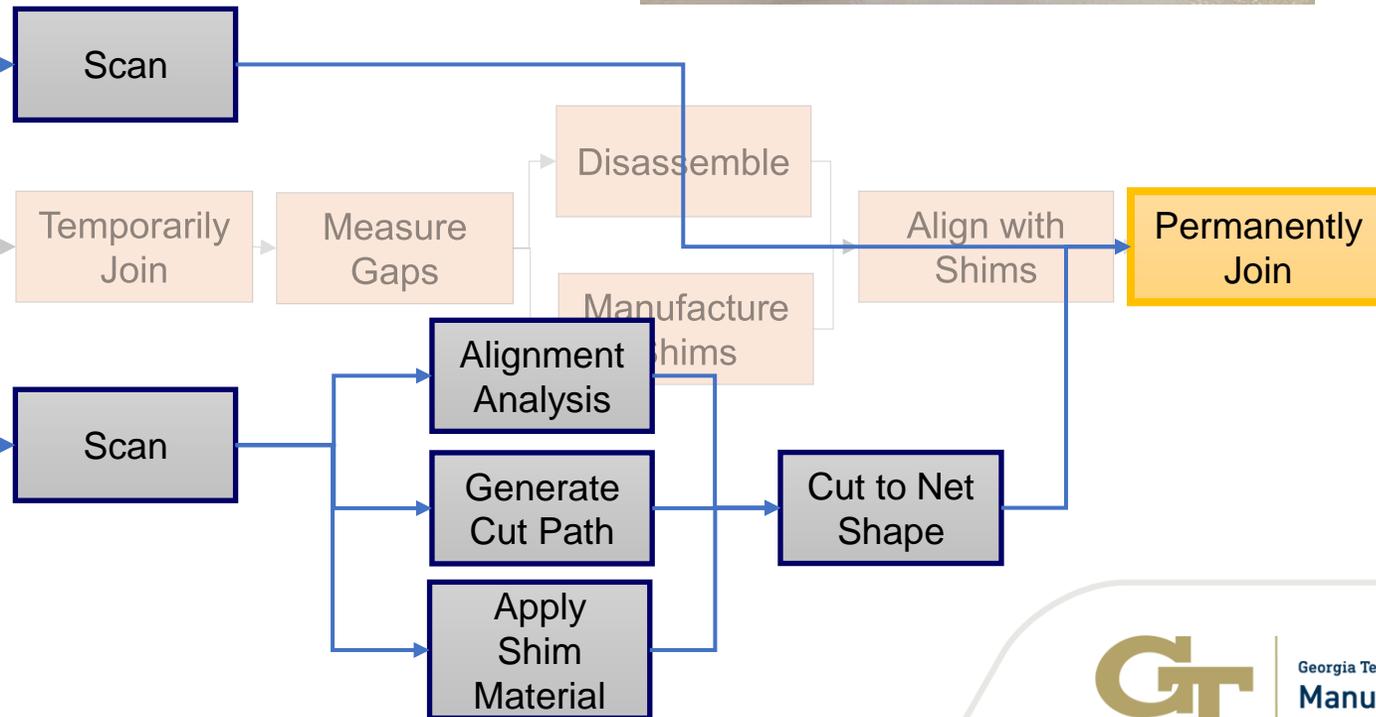
Automate shim assembly using low-cost, accurate articulated arms to scan mating surfaces and machine joining surface to net shape



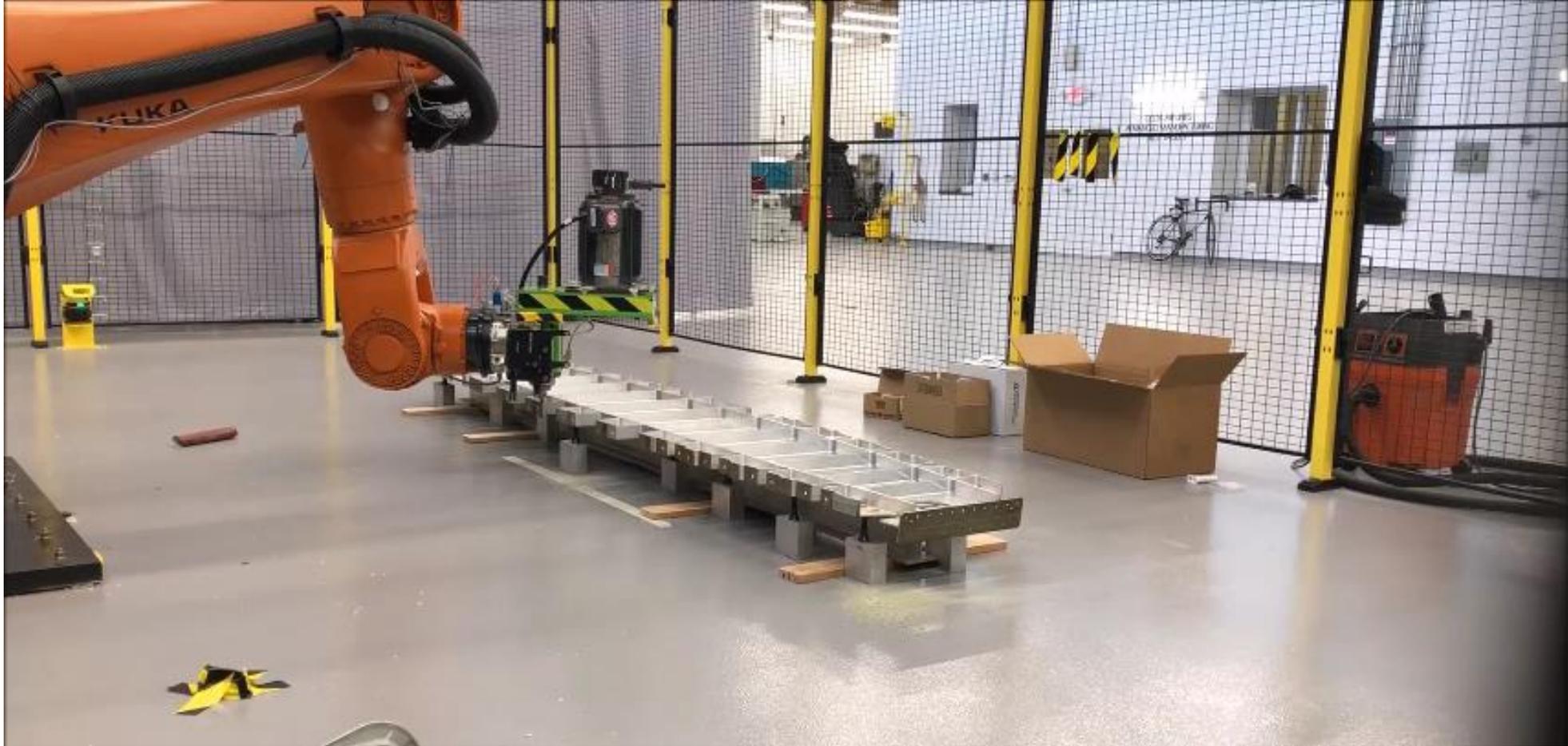
Wing Skin



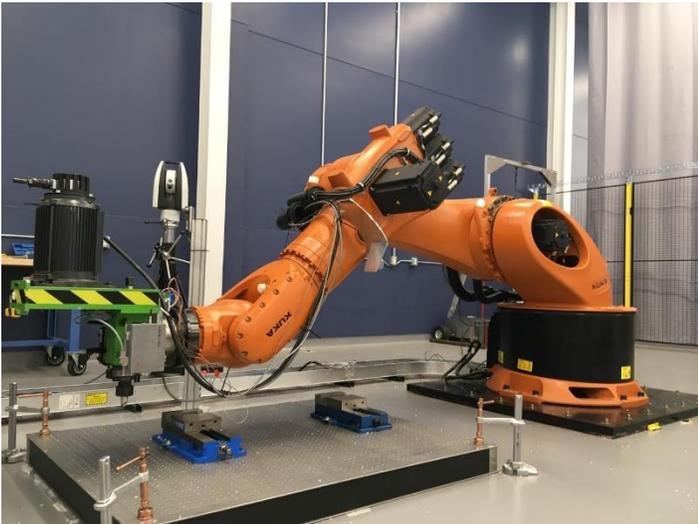
Wing Rib



# Wing Rib Shim Machining



# Factors Contributing to Machining Accuracy



Robots have much lower kinematic accuracy than CNC machines

Low kinematic accuracy → Part inaccuracy

Robots are more compliant than CNC machines

Large compliance → Part inaccuracy

Modal properties of robot arms change with configuration

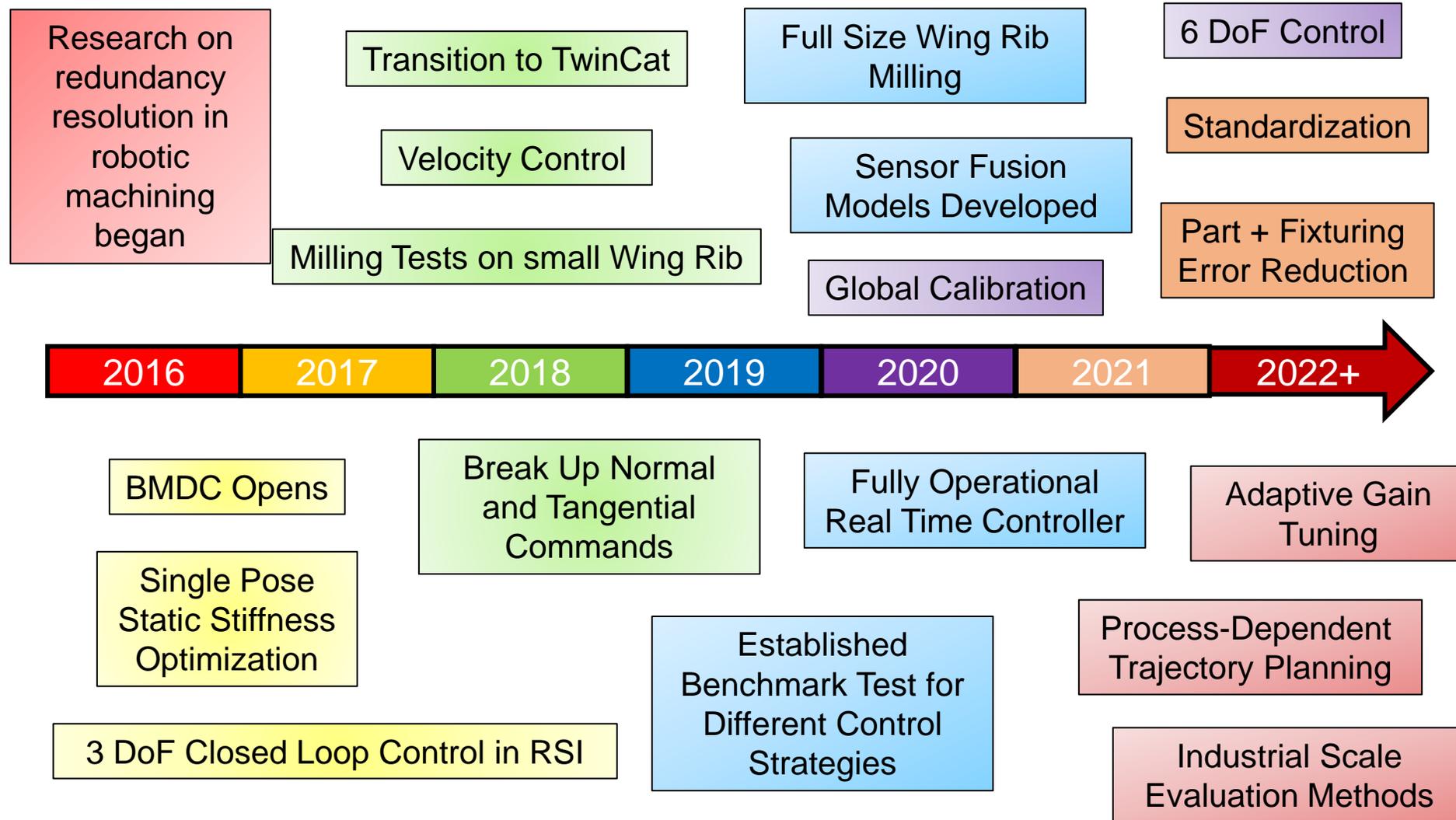
Pose dependent FRF → Dynamic path errors

Interaction between robot and workpiece

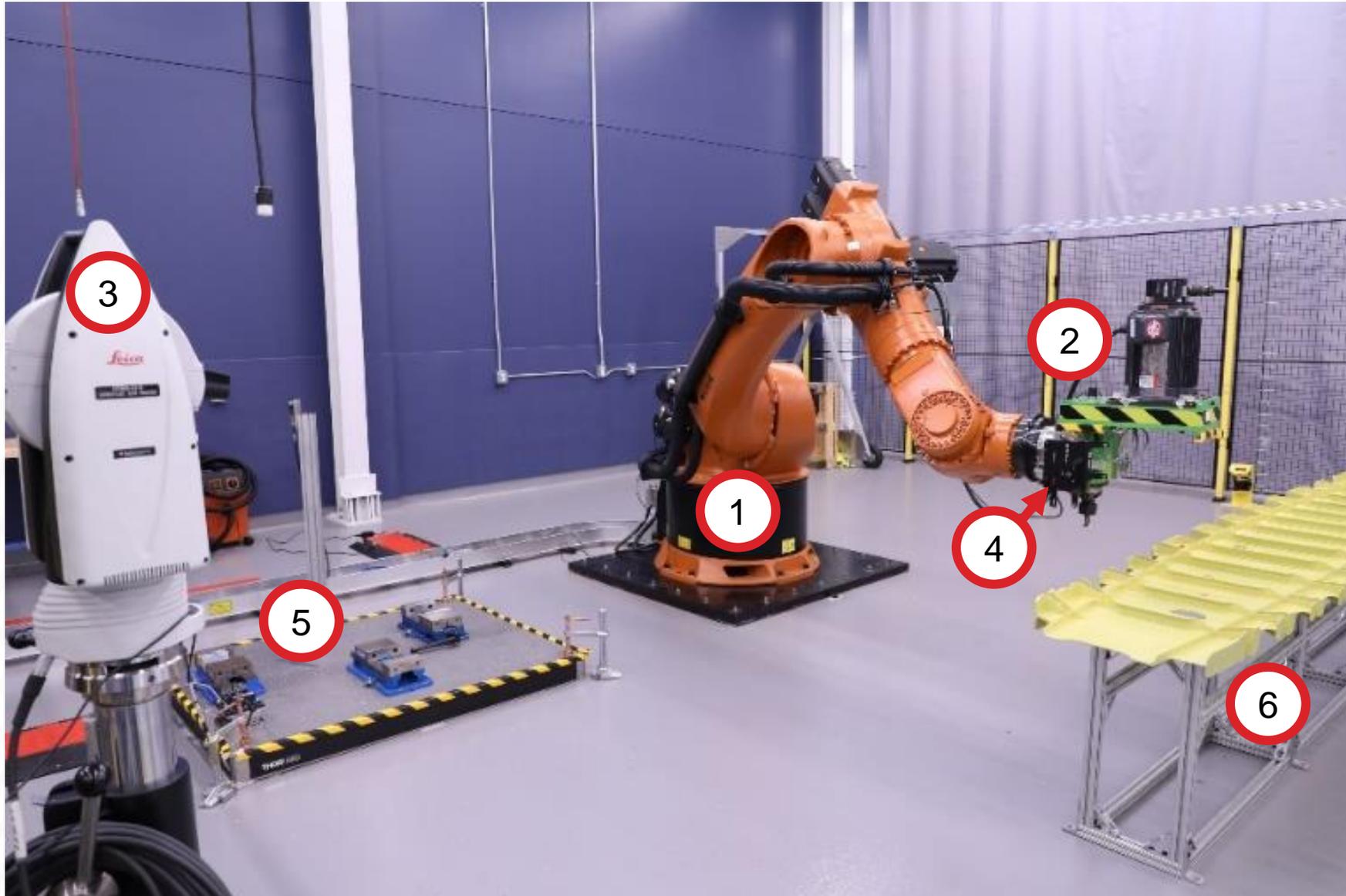
Time-varying process forces and part compliance → Part inaccuracy



# Accurate Robotic Machining Research @ GT



# Work Cell



- ① KUKA KR500-3 Robot
- ② Mill
- ③ Leica AT960 Laser Tracker
- ④ Leica T-Mac Tracker Target
- ⑤ Vices
- ⑥ Wing rib

# Control Architecture

## Sensors

- Leica AT960 laser tracker and Leica T-Mac
- Leica T-Scan
- Gladiator Technologies LandMark™60 IMU
- ATI Force Sensor

## Processing Nodes

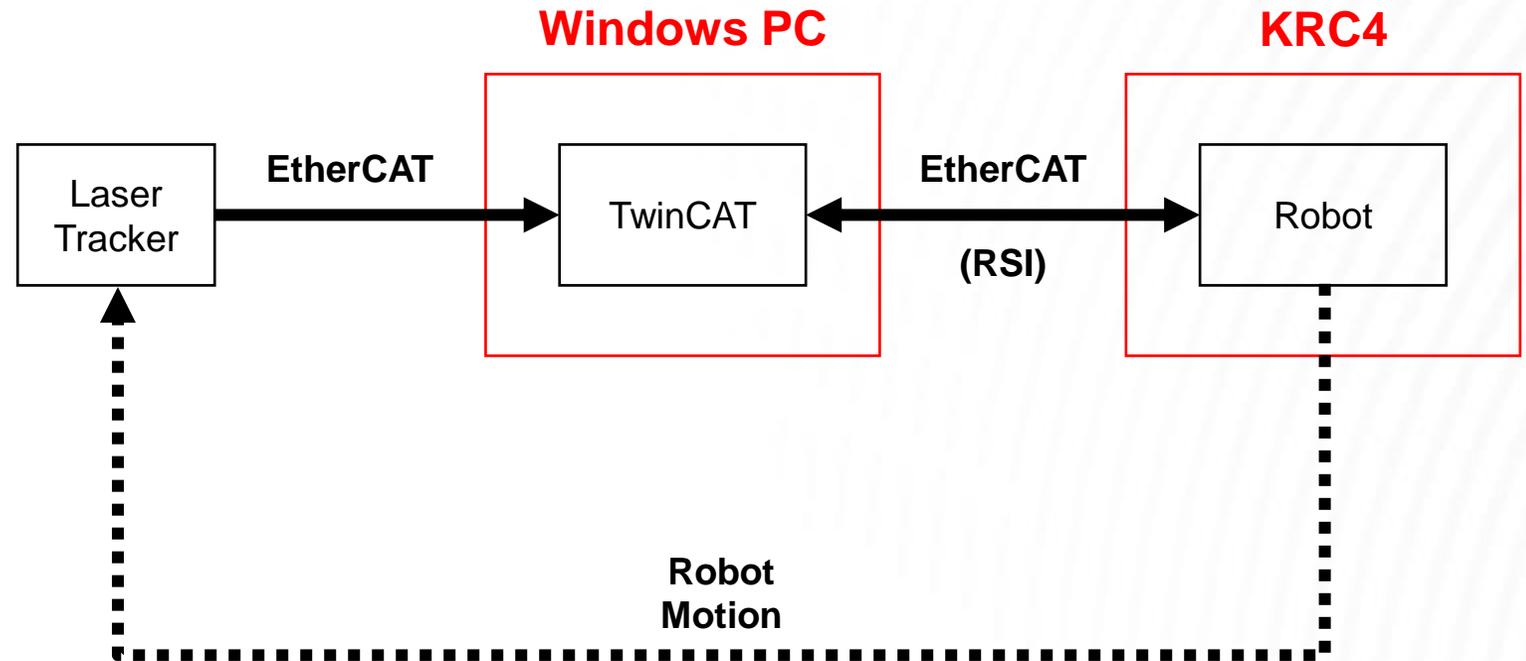
- KUKA KRC4 robot controller
- Windows workstation

## Network

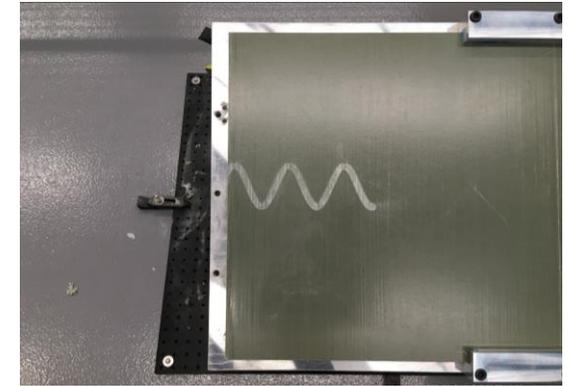
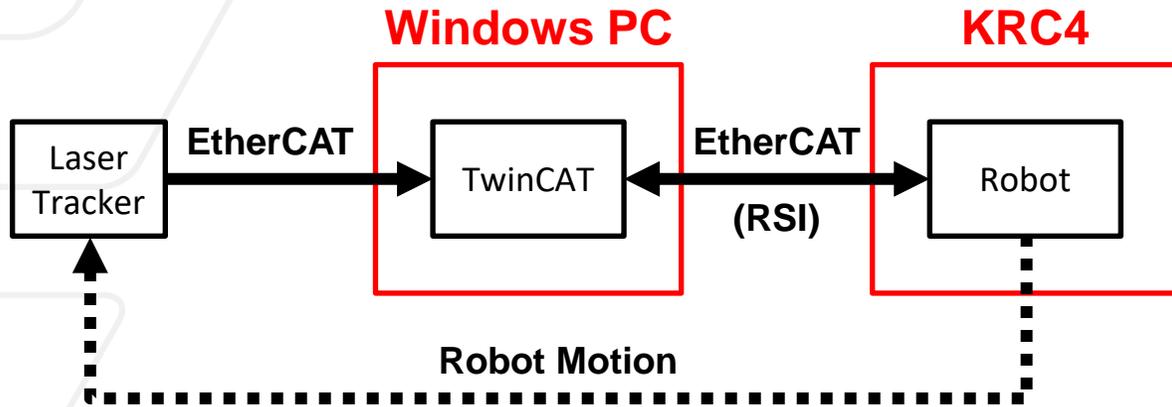
- EtherCAT

## Plant

- KUKA KR 500-3 Robot
- KUKA smartPAD Teach Pendant



# Real-Time Feedback Control

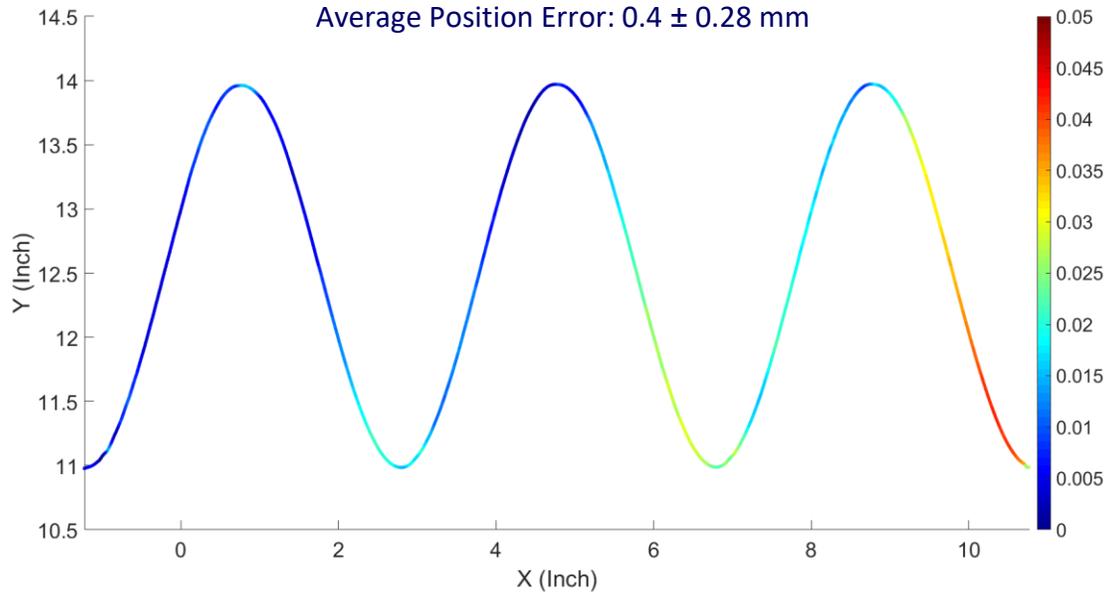


Nguyen et al., *SAE Int. J. Aero.*, 2021

Tool Tip Position Measured using Leica AT960 with T-Mac

Open Loop

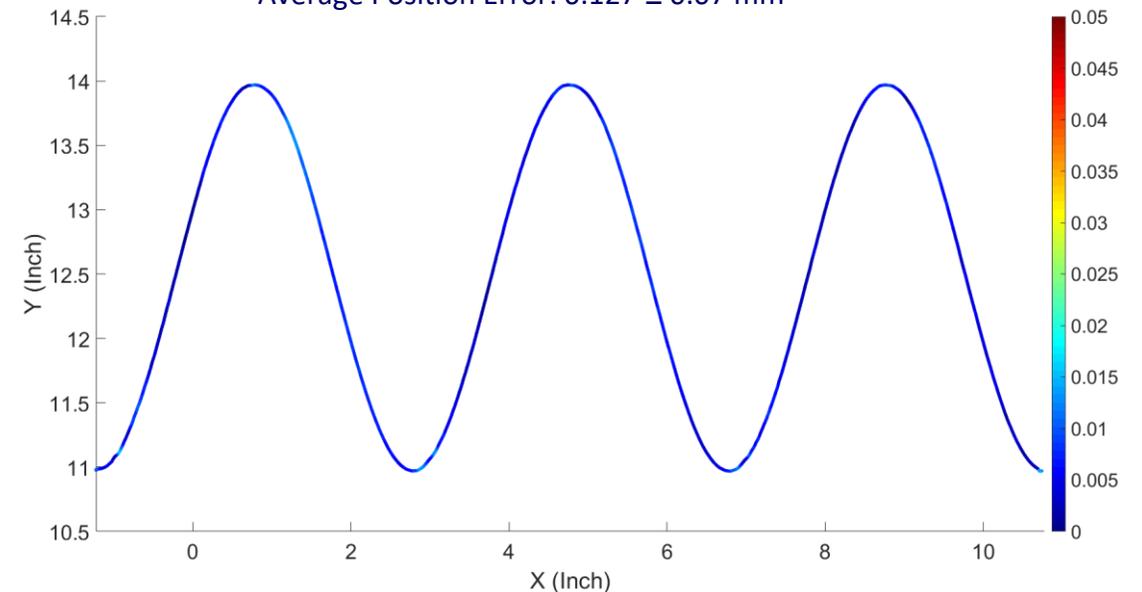
Average Position Error:  $0.4 \pm 0.28$  mm



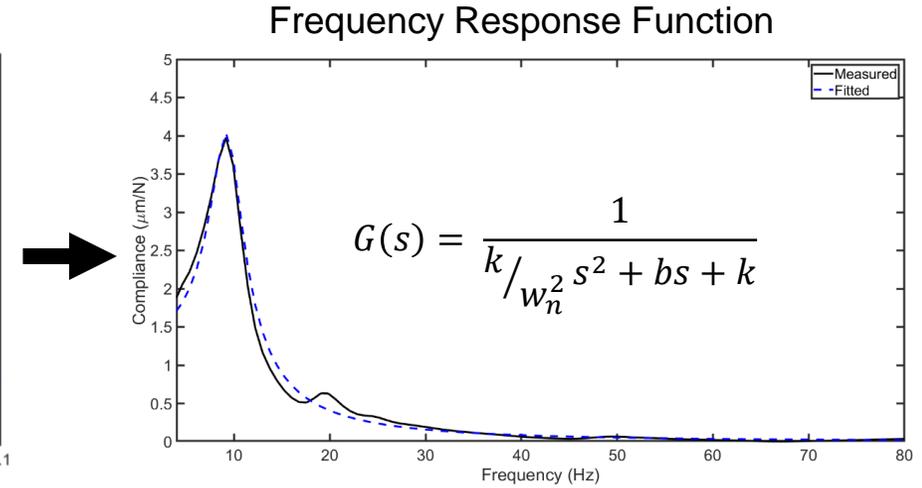
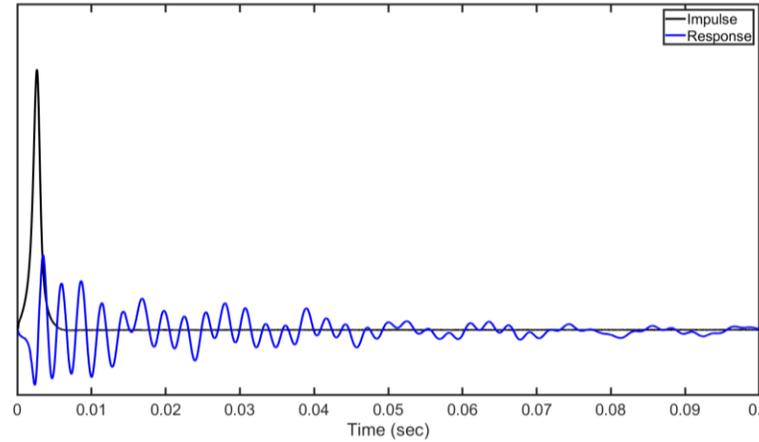
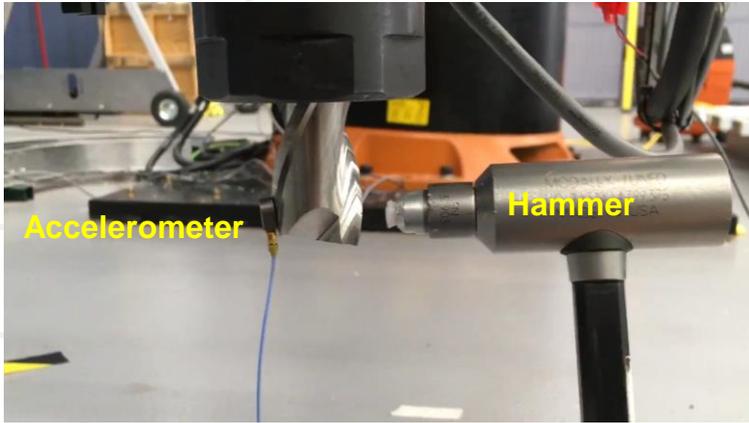
Tool Tip Position Measured using Leica AT960 with T-Mac

KRL Guided

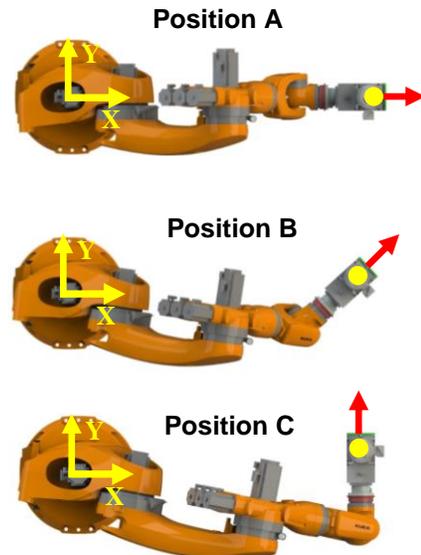
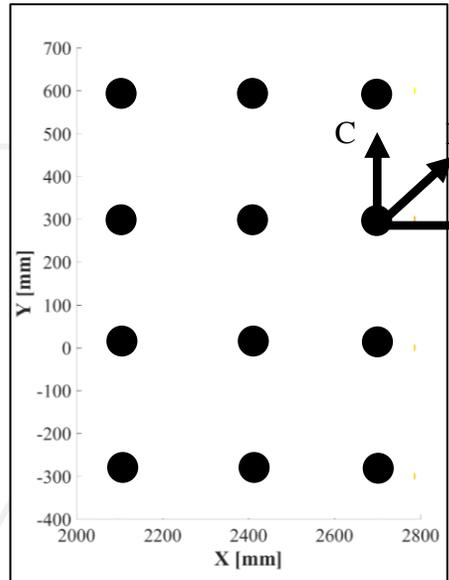
Average Position Error:  $0.127 \pm 0.07$  mm



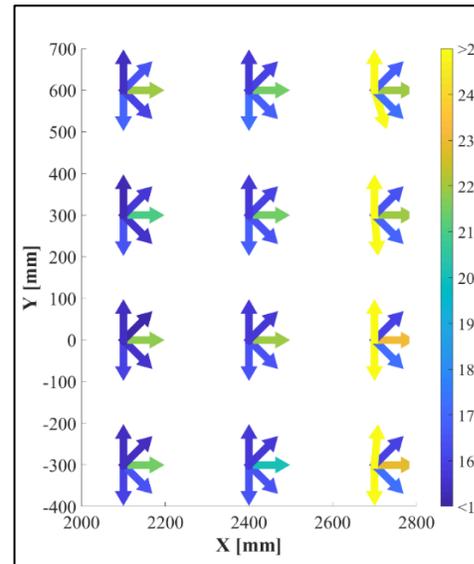
# Data-driven Modeling of Robot Dynamics



## Sampled Points in the Workspace



## Natural Frequency, $f(p_i)$ [Hz]



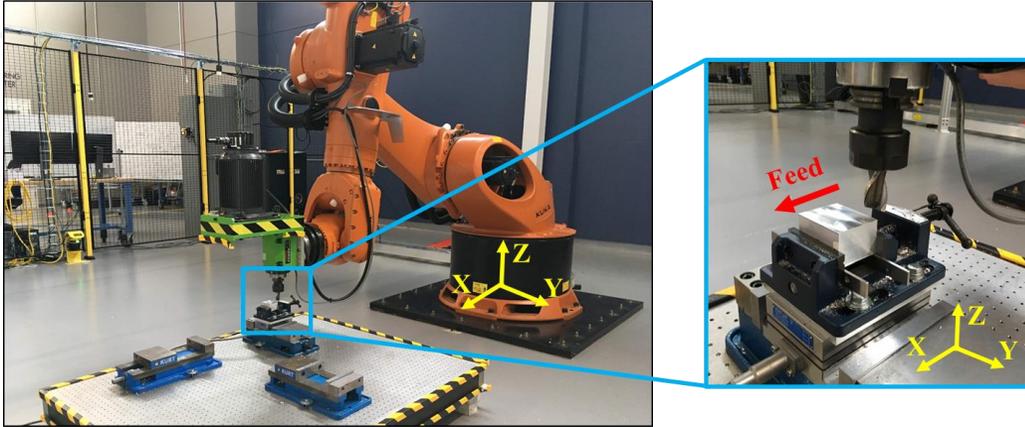
We can model  $f(p_i)$  as a Gaussian Process:

$$f(p_i) \sim GP(\mu(p_i), K(p_i, p_i'))$$

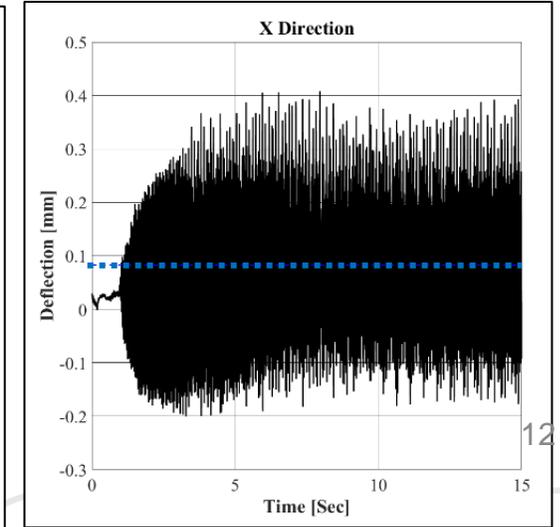
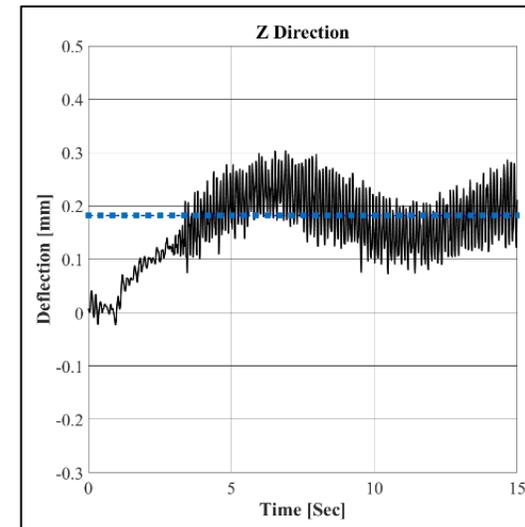
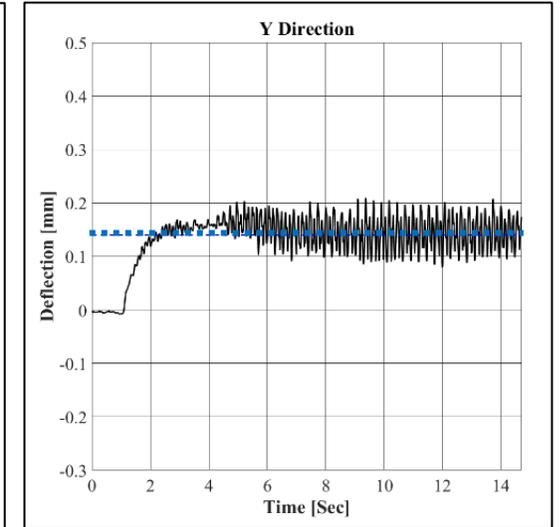
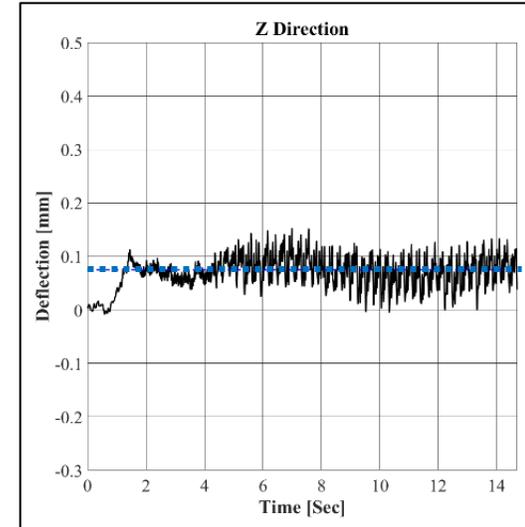
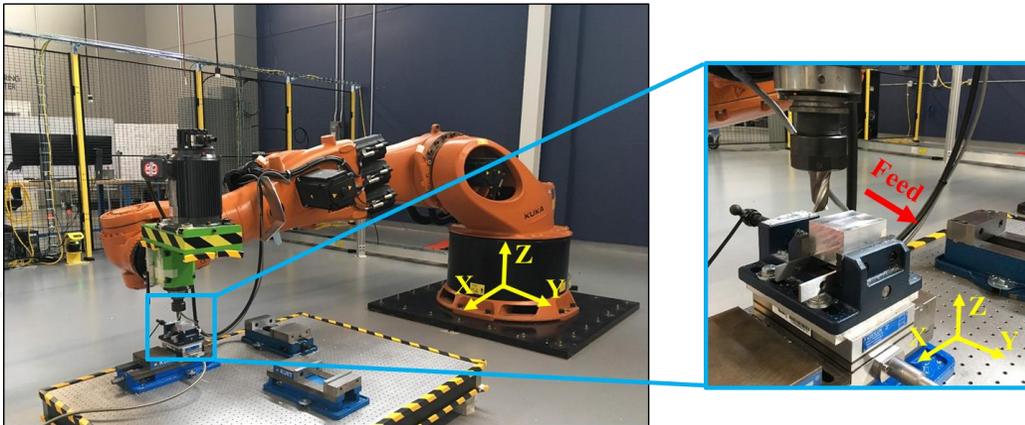
Nguyen et al., *J. Mfg. Sci. Eng.*, 2019

# Application to Milling Vibration Prediction

**Position 1**



**Position 2**



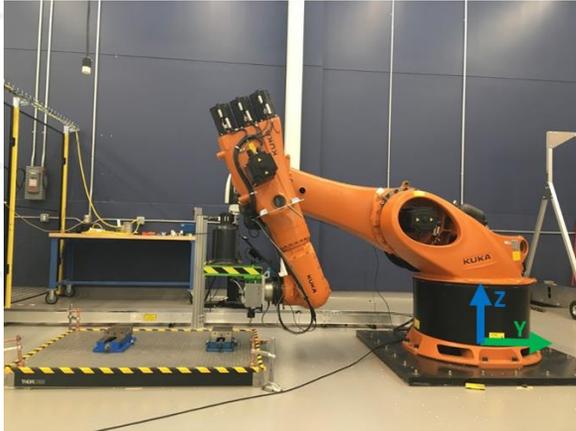
**Peak-to-Valley Vibrations (mm)**

	Position 1		Position 2	
	Y	Z	X	Z
<b>Measured</b>	$0.085 \pm 0.004$	$0.064 \pm 0.005$	$0.506 \pm 0.022$	$0.125 \pm 0.008$
<b>Predicted</b>	$0.070 \pm 0.002$	$0.058 \pm 0.001$	$0.473 \pm 0.014$	$0.100 \pm 0.006$

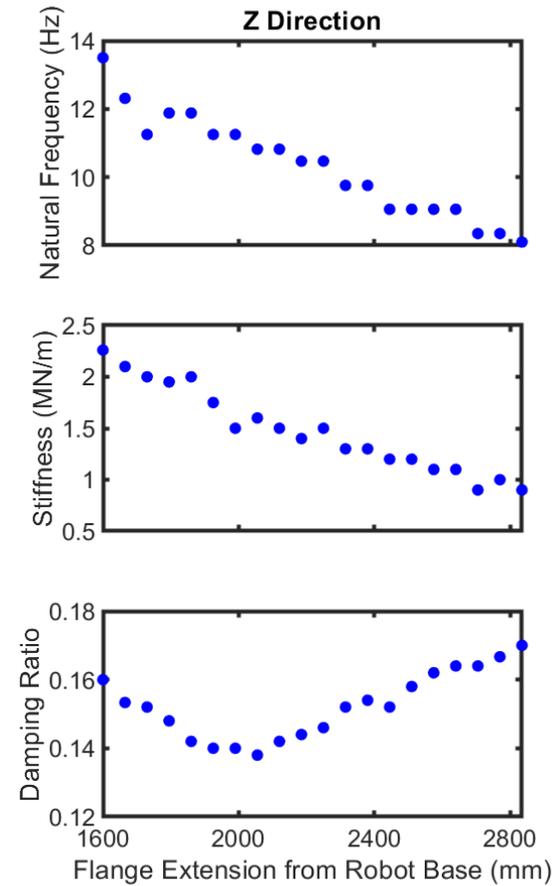
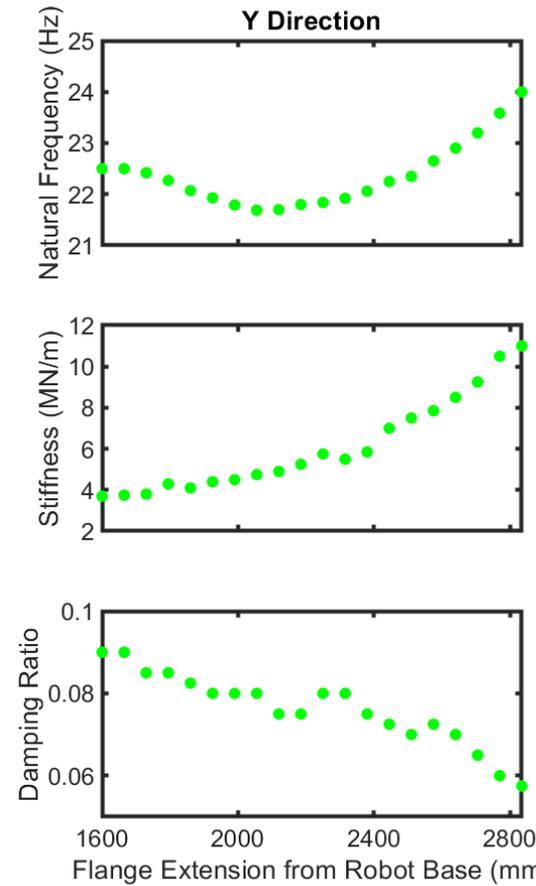
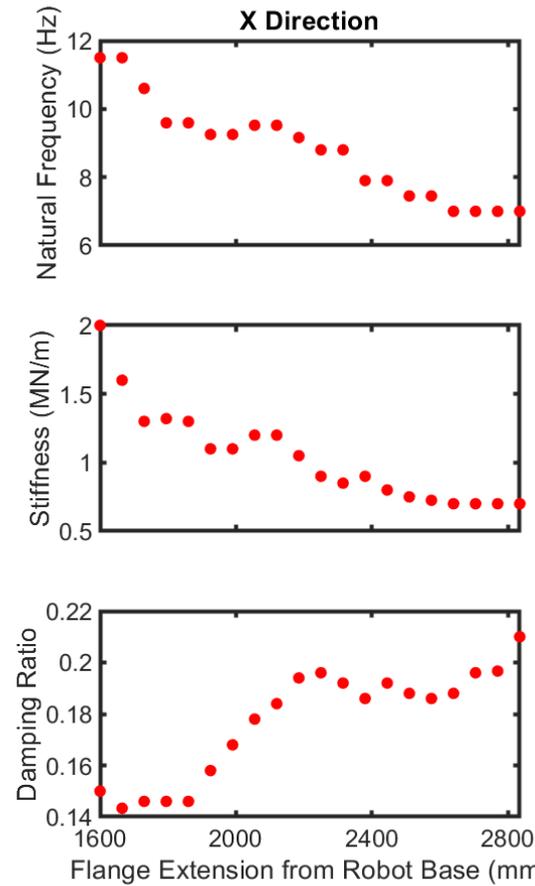
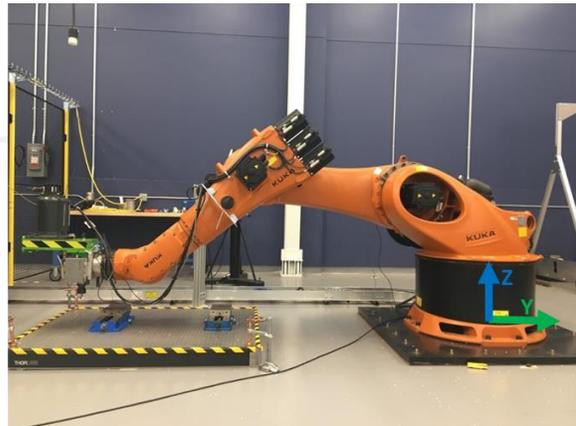
**The GPR model peak-to-valley vibration predictions are correlated to the measured behavior**

# Pose Dependent Modal Properties

Start

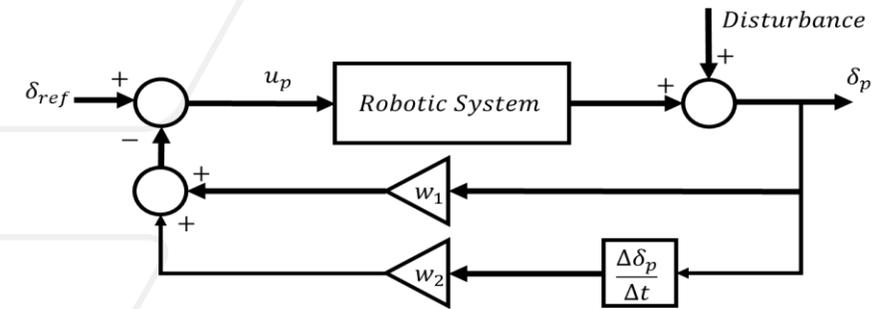


End

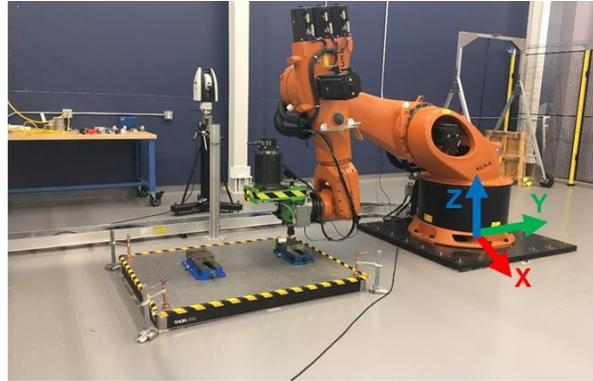


The data shows that the modal parameters change with arm configuration, and therefore arm configuration must be considered in controller tuning

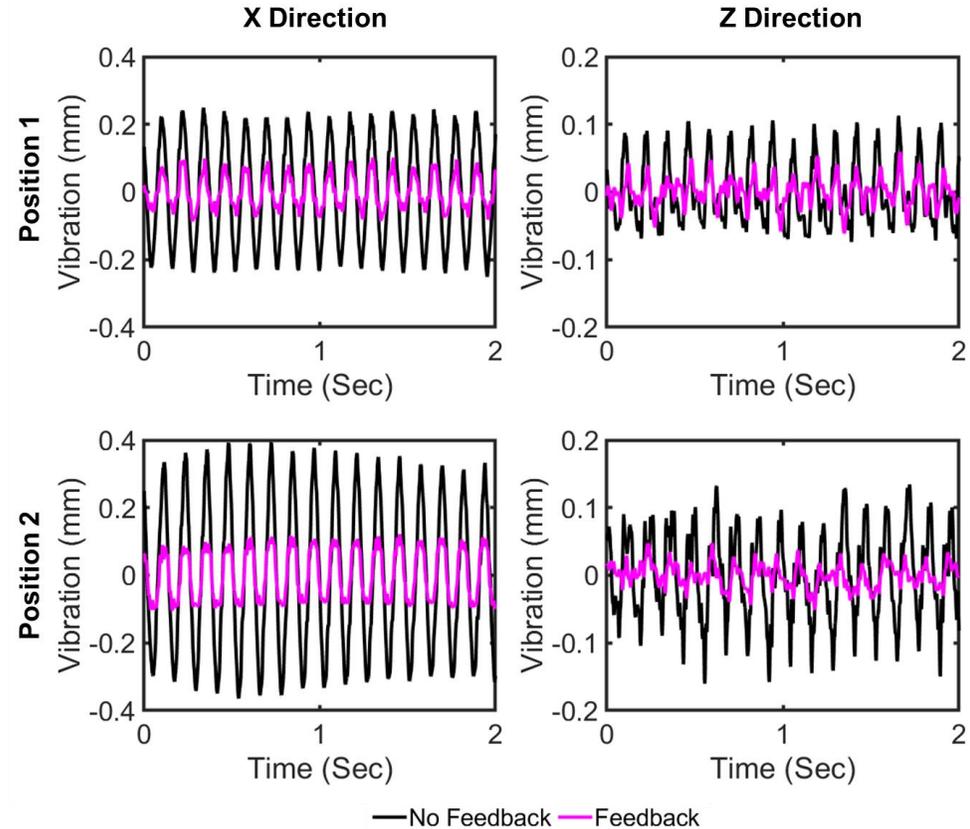
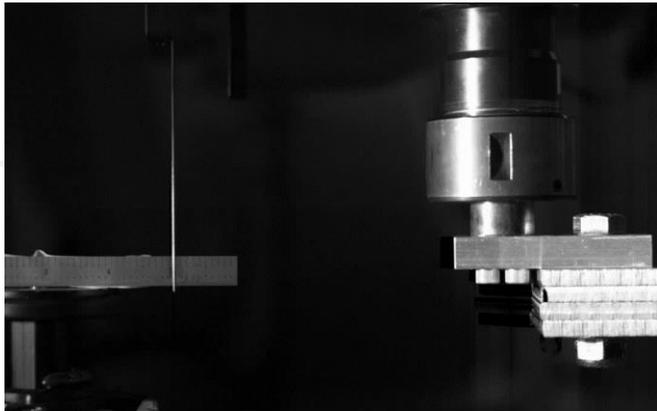
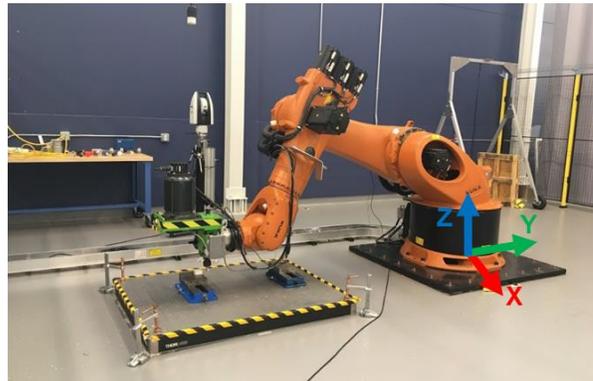
# Active Vibration Suppression



**Position 1**

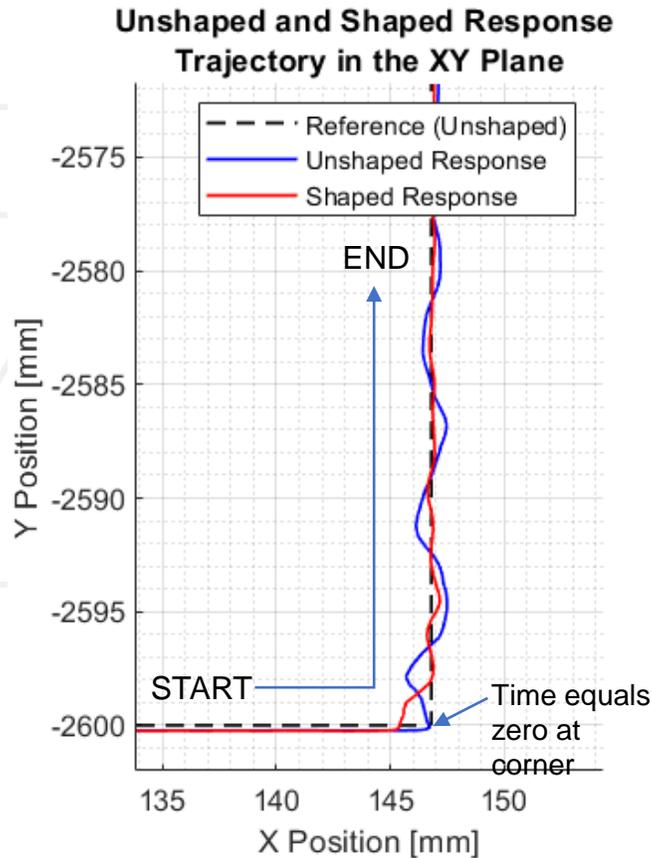


**Position 2**



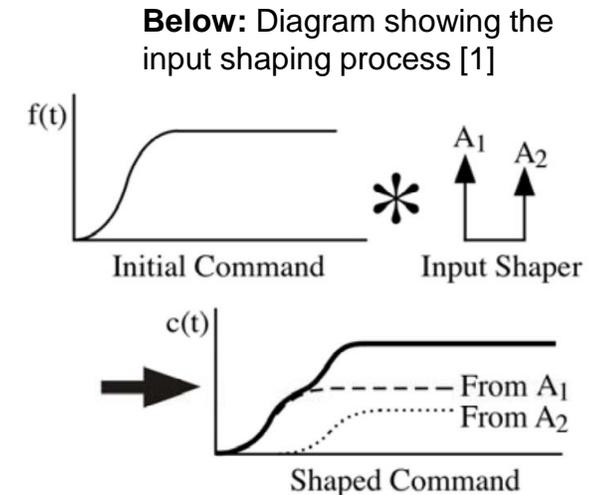
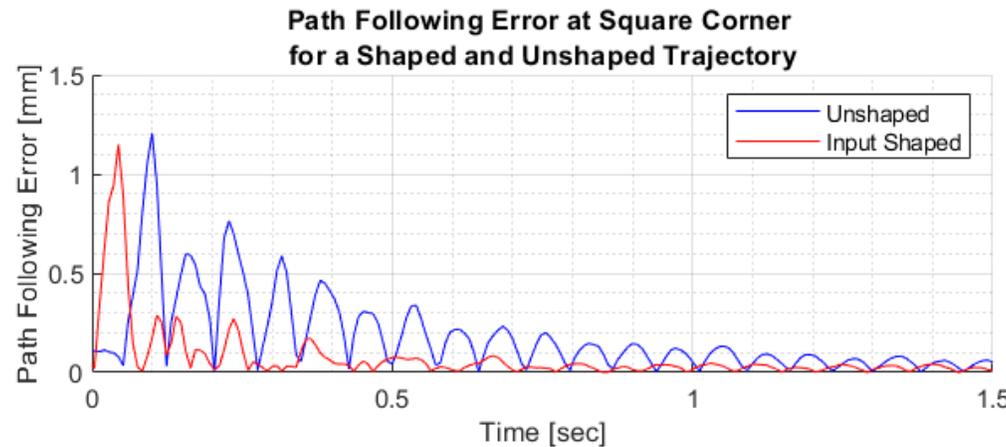
# Input Shaping to Improve Dynamic Path Accuracy

- A technique that may improve dynamic path accuracy is to apply input shaping to the desired trajectory
- Preliminary results indicate that this approach may significantly reduce vibrations from self excitation of the robot structure



**Left:** Square wave trajectory in the XY plane shows the reference trajectory, measured response to the unshaped trajectory, and measured response to the shaped trajectory, at 120 ipm velocity

**Below:** The path following error to the reference trajectory after rounding the corner in the trajectory shown at left for the shaped and unshaped trajectory



# Ongoing Work

- Integrating on-robot vision-based metrology for detection of part features and their location to guide machining trajectories (e.g., trimming to a molded scribe line)
- Input shaping of complex curvilinear trajectories to suppress structural vibration (e.g., cornering)



CFRP part with Scribe Line

# Needs

- Standard method to characterize the accuracy of external sensor-guided (e.g., laser tracker) robots
- Ability for users to modify robot control action at faster rates ( $\ll 4$  ms) based on external sensor feedback
- Standard method to determine the pose-dependent modal properties of industrial robots over a defined working region



# Acknowledgements

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