

# *Vision and Force Guided Robot Assisted Assembly*

***Smarter Robots in Manufacturing through  
Modeling, Sensing, Control, and Learning***

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American Control Conference Workshop

7/9/19



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# Outline

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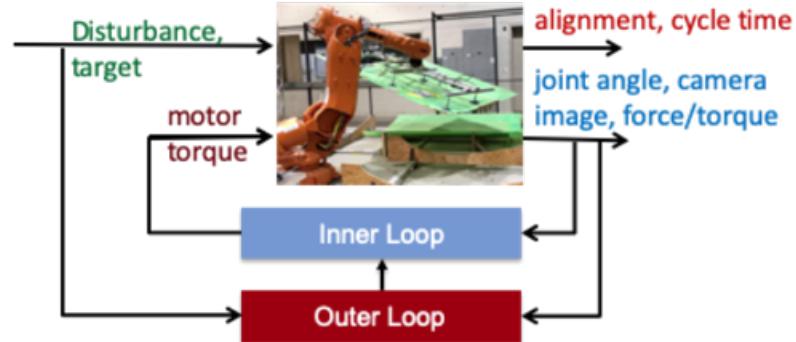
- Attributes and Challenges in Industrial Robot Control
- Case Studies
  - Composites Panel Assembly
  - Robotic Satellite Servicing
  - Assistive Robot
- Robotics Software
- Robotics education



# Industrial Robot Control

## Attributes

- Outer Loop Control
- Semi-structured
- Crowded workspace: Human workers, other robots and machineries
- Process interruptions



## Challenges/Requirements

- Speed
- Precision
- Model imprecision
- Variable operation conditions
- Extensible architecture
- Load flexibility
- Joint flexibility
- Outer Loop Dynamics (delay, nonlinearities)
- Redundancy resolution
- Interruption recovery



# Industrial Robot Control

## Challenges/Requirements      Potential Solutions

- |                           |                                  |
|---------------------------|----------------------------------|
| • Speed                   | ← Local optimization             |
| • Precision               | ← Sensor based control           |
| • Model imprecision       | ← Robust control                 |
| • Variable conditions     | ← Robust control                 |
| • Extensible architecture | ← State machine, hybrid control  |
| • Load flexibility        | ← Control design, notch filter   |
| • Joint flexibility       | ← Control design, notch filter   |
| • Outer loop dynamics     | ← Predictive compensation        |
| • Redundancy resolution   | ← Secondary control objectives   |
| • Interruption recovery   | ← Error detection and mitigation |



# Control Needs

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- Faster motion planner
- More *intuitive* robot paths
- Robustness to lighting, calibration..
- Co-simulation
- Human predictive modeling
- ....

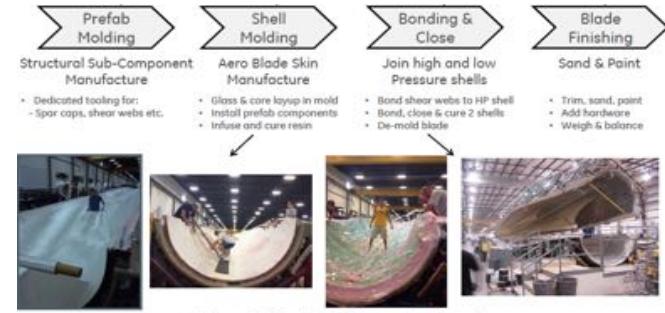


# Case Study: Composites Panel Assembly

## Motivation

- Manufacturing of large composite structures a time consuming, labor intensive process
- Challenges: Large, heavy, curved loads, structural flexibility, high precision alignment requirement

Joint project with GE GRC, funded by ARM



Labor / Capital / Space Intensive

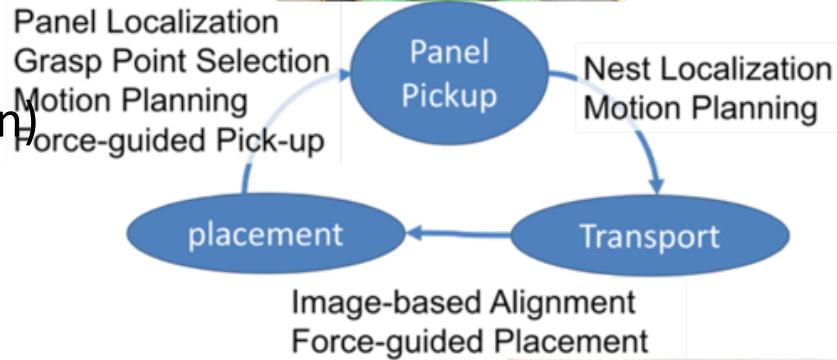
Courtesy: Shridhar Nath  
(GE GRC)



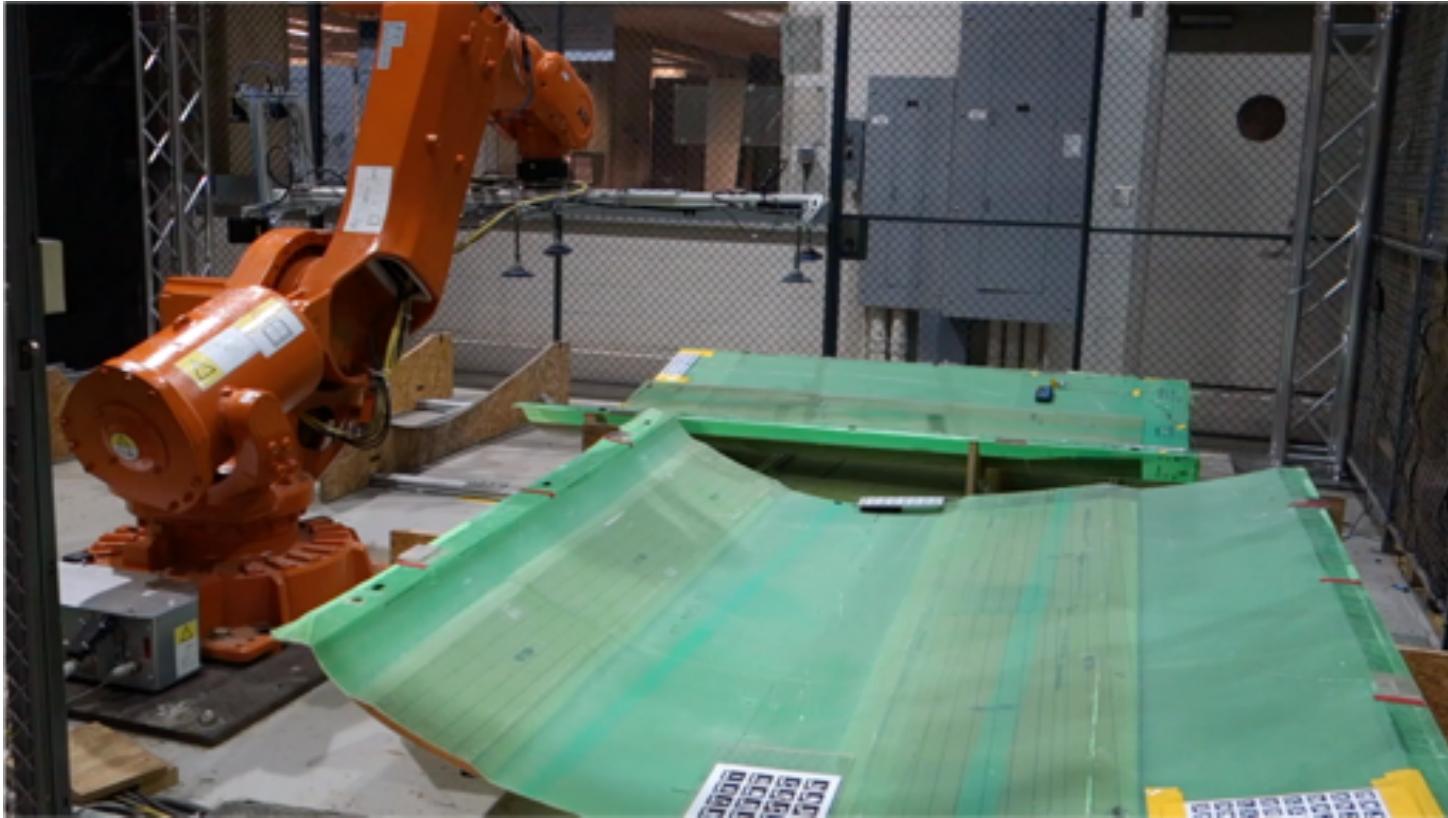
# Demonstration Testbed

Performance Metrics: Cycle Time, Alignment Tolerance, Damage avoidance (vibration, placement force)

- Speed (motion, planning)
- Precision (alignment, force)
- Model imprecision (CAD, camera calibration)
- Variable conditions (lighting)
- Extensible architecture (robot-agnostic)
- Load flexibility (panel vibration)
- Joint flexibility
- Outer loop dynamics (nonlinearity)
- Redundancy resolution
- Interruption recovery (error, user)



# Experiment

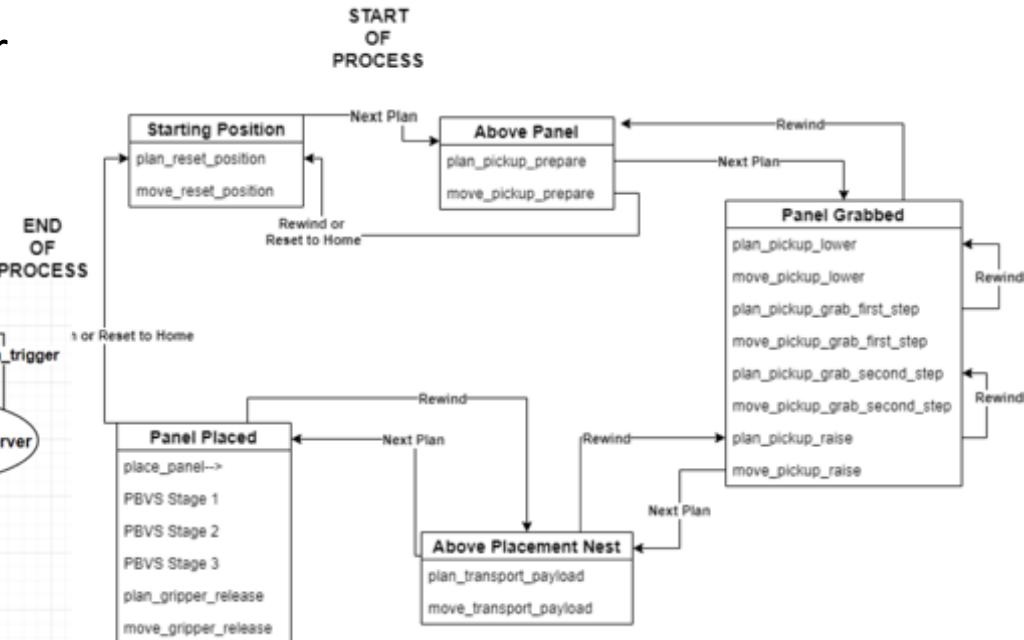
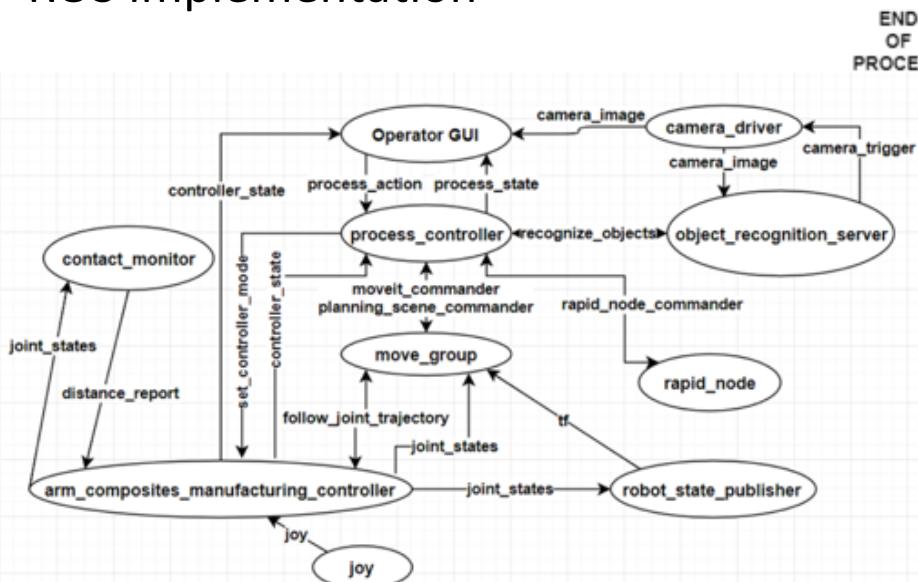


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# Process and Software Architecture

- Bidirectional process flow for operator intervention and diagnostics
- ROS implementation



Opportunity: Hybrid control analysis



# Motion Planning

Global optimization with fixed environment

- MoveIt! (TRRT)
- TrajOpt + Tesseract (SwRI)

Local Solution: Quadratic Programming with human guidance

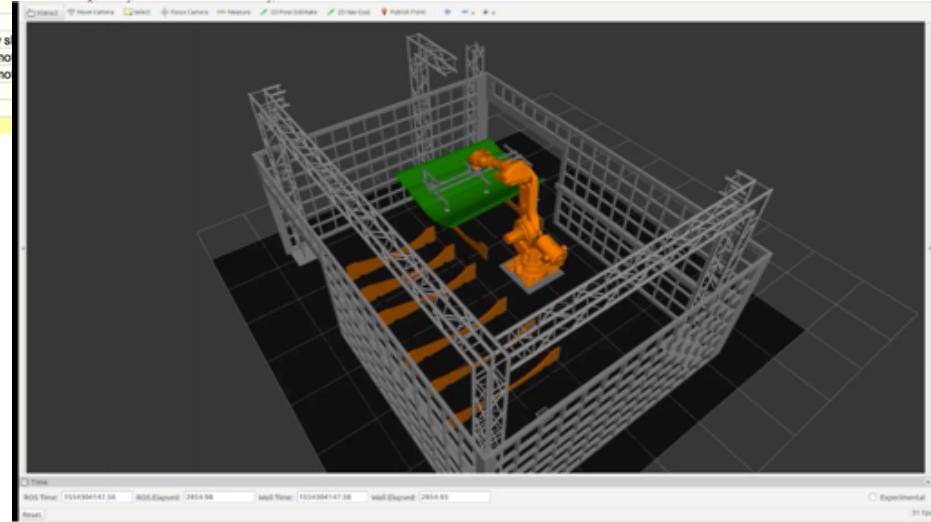
$$\min_{\dot{q}, \alpha_r, \alpha_p} \|J\dot{q} - \alpha v_d\|^2 + \epsilon_r(\alpha_r - 1)^2 + \epsilon_p(\alpha_p - 1)^2$$

$$h_I(q) \geq \eta > 0 \Rightarrow \nabla h_I(q)\dot{q} \geq \sigma(h_I(q))$$

$$\dot{q}_{\min} \leq \dot{q} \leq \dot{q}_{\max}$$

Opportunity: User directed optimal control, model predictive control

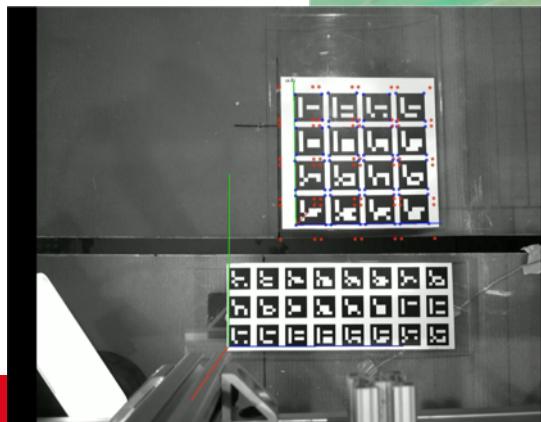
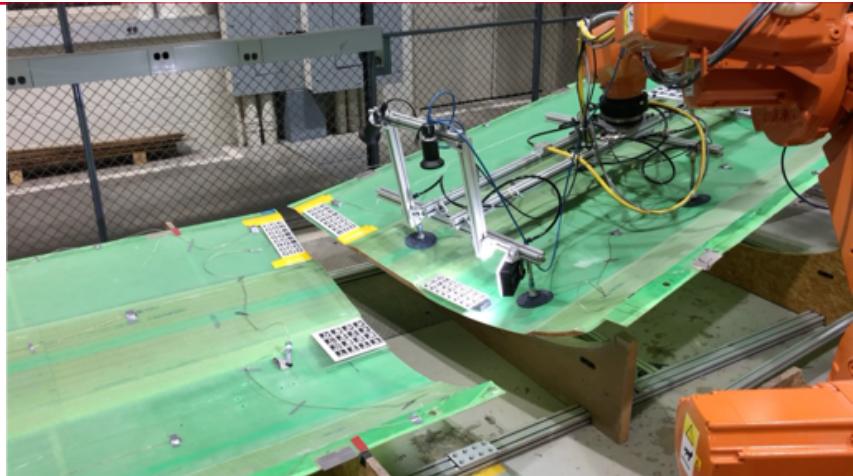
#	Planners	Trial 1		Trial 2	
		Good Path	Bad Path	Good Path	Bad Path
1	BFMTkConfigDefault	2	3		
2	BKPIECEkConfigDefault	5	0	4	1
3	BiESTkConfigDefault	4	1		
4	BiTRRTkConfigDefault	4	1 no motion plan found		
5	ESTkConfigDefault	4	1 no motion plan found		
6	FMTkConfigDefault	5	0	4	1 no motion plan found
7	KPIECEkConfigDefault	5	0	4	1 no motion plan found
8	LBKPIECEkConfigDefault	0	5		
9	LBTRRTkConfigDefault	4	1		
10	LazyPRMkConfigDefault	4	1		
11	LazyPRMstarConfigDefault	4	1		
12	PDSTkConfigDefault	4	1 no motion plan found		
13	PRMkConfigDefault	4	1		
14	PRMstarConfigDefault	4	1		
15	ProjESTkConfigDefault	5	0	5	0
16	RRTConnectkConfigDefault	5			
17	RRTkConfigDefault	3			
18	RRTstarConfigDefault	very s			
19	SBLkConfigDefault	4	1 no mo		
20	SPARSkConfigDefault	4	1 no mo		
21	SPARStwokConfigDefault	5			
22	STRIDEkConfigDefault	5			
23	TRRTkConfigDefault	5			



# Vision Guided Motion

- Placement (gripper cameras): PBVS vs. IBVS
    - PBVS pose-based vs IBVS image-based error
    - Servoing: Jacobian vs. Inverse kinematics
    - Reduce dependence on calibration accuracy
    - Single tag/camera vs. Two-tag/two-camera
      - Trade-off between orientation accuracy and camera accuracy.
- Best experimental combination:  
.75 overhead + .25 side cameras

Opportunity: Sensor-fusion, integrated lighting control, visual design



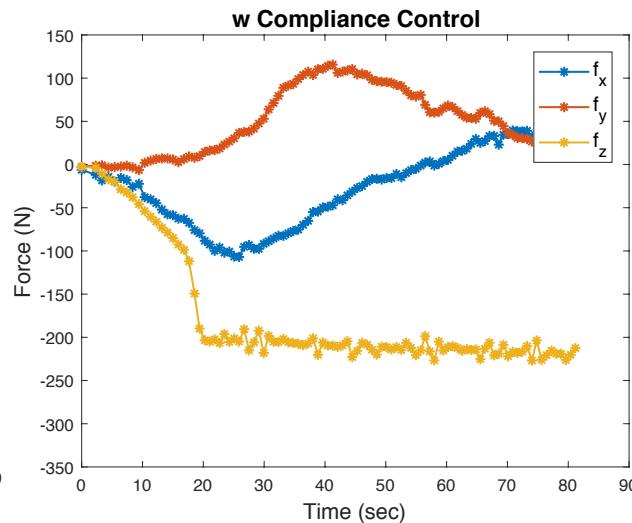
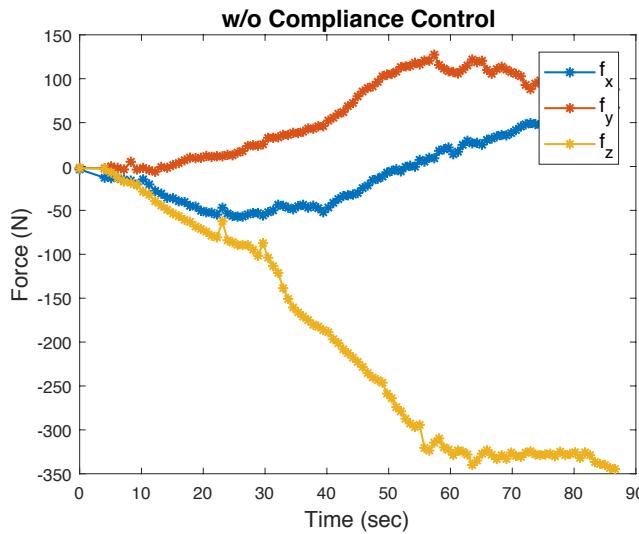
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# Force Guided Motion

- Placement: compliance force control (generalized damper) as part of QP control

$$\begin{bmatrix} \omega_T \\ v_T \end{bmatrix} = \begin{bmatrix} \omega_m \\ v_m \end{bmatrix} - K \begin{bmatrix} \tau - \tau_d \\ f - f_d \end{bmatrix}$$

$$\min_{\dot{q}, \alpha_R, \alpha_P} \left\| J(q)\dot{q} - \begin{bmatrix} \alpha_R \omega_T \\ \alpha_P v_T \end{bmatrix} \right\|^2 + \epsilon_R(\alpha_R - 1)^2 + \epsilon_P(\alpha_P - 1)^2$$

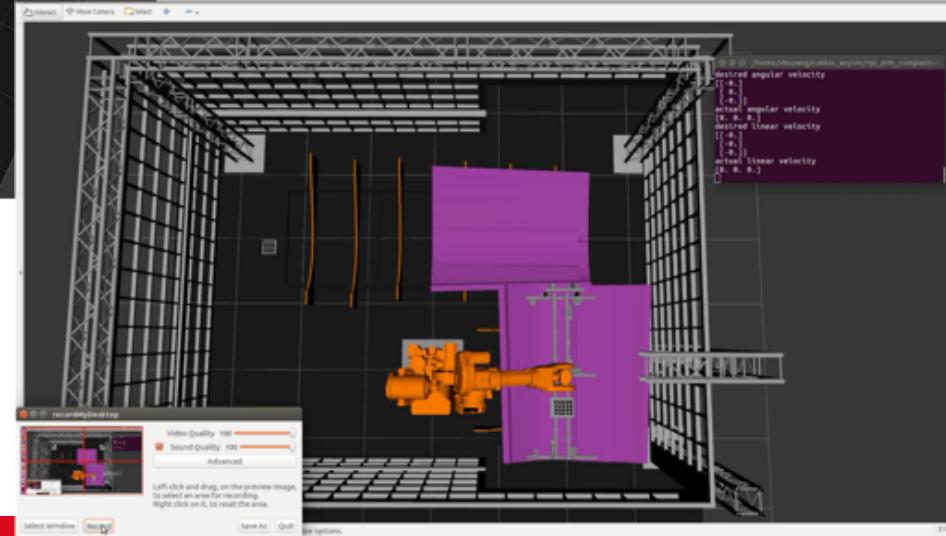
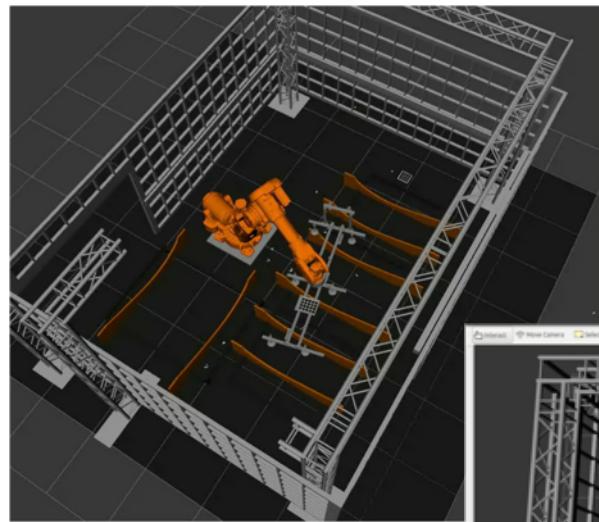


Opportunity: Multi-contact flexible load assembly/compliance control



# Human Tracking, Safe Teleoperation

- Multiple point cloud sensor for human tracking, object identification
- User teleoperation without collision
- Distance calc: 10Hz



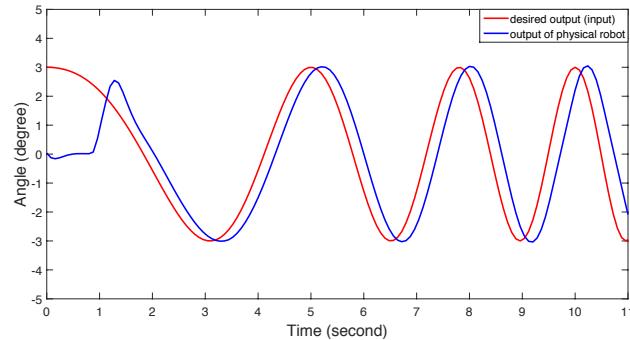
**Opportunity:**  
Guaranteed safe robot motion with predictive control



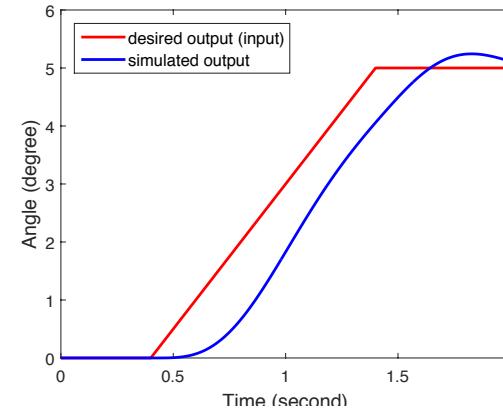
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# Iterative and Neural Learning Control

- Robot/load dynamics
- Iterative improvement over tracking error
- Simulation-based learning
- Encoding in neural network

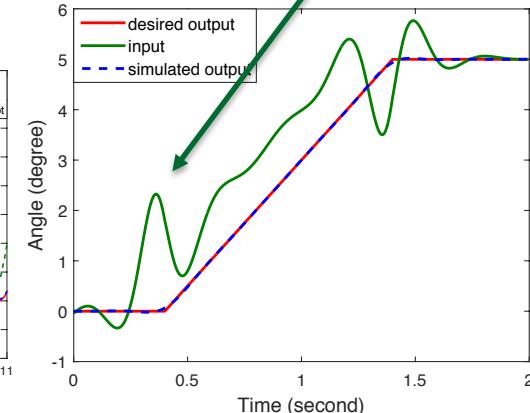
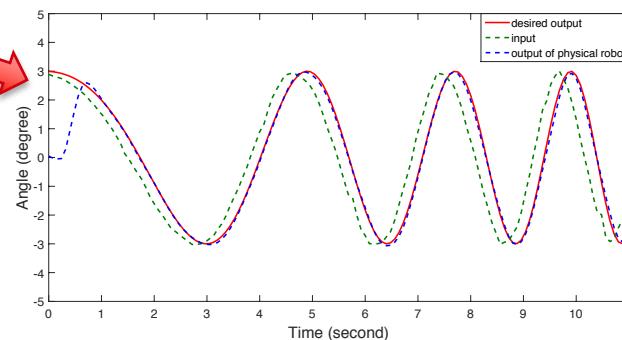


Opportunity: Neuro-adaptive control for nonlinear dynamical system



Strictly proper system →  
Non-causal input

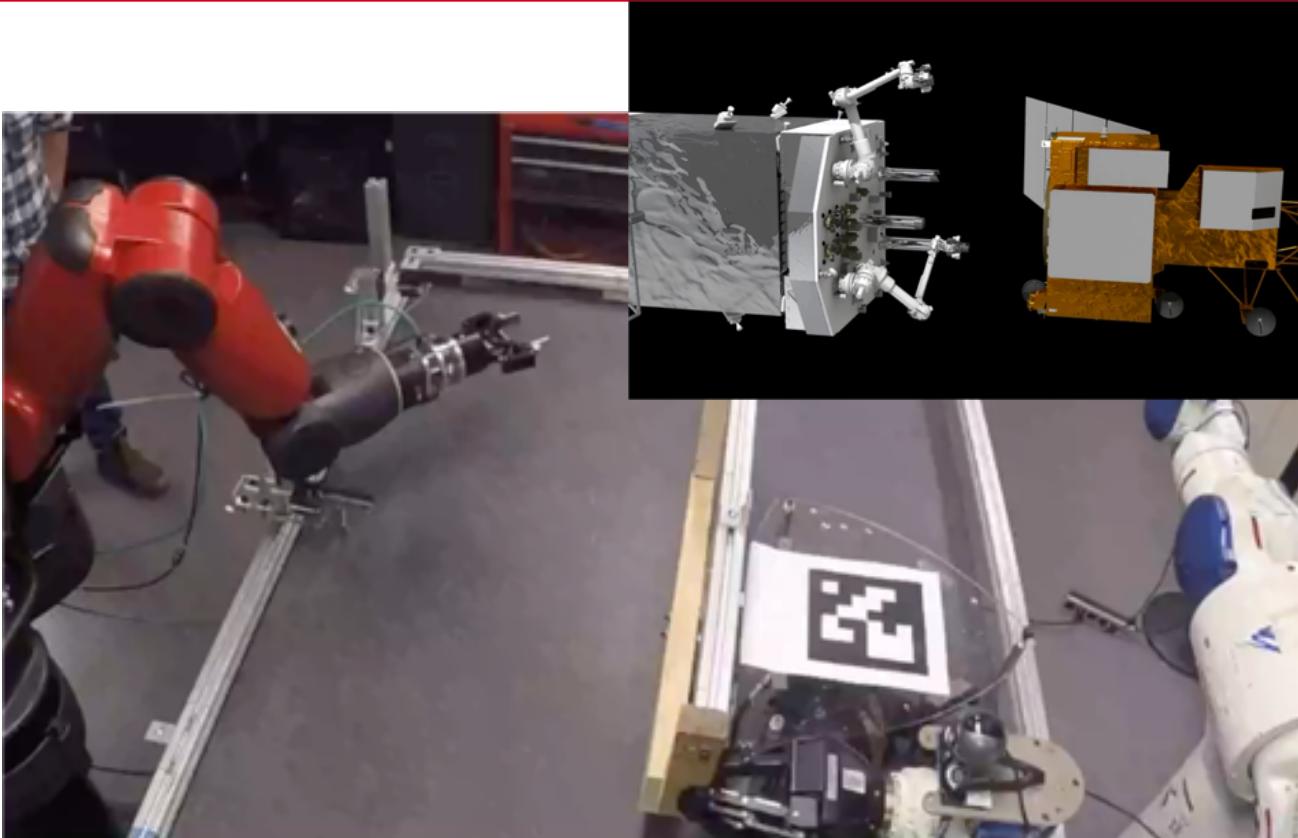
Noncausal NN FF compensation



# Robot Satellite Servicing

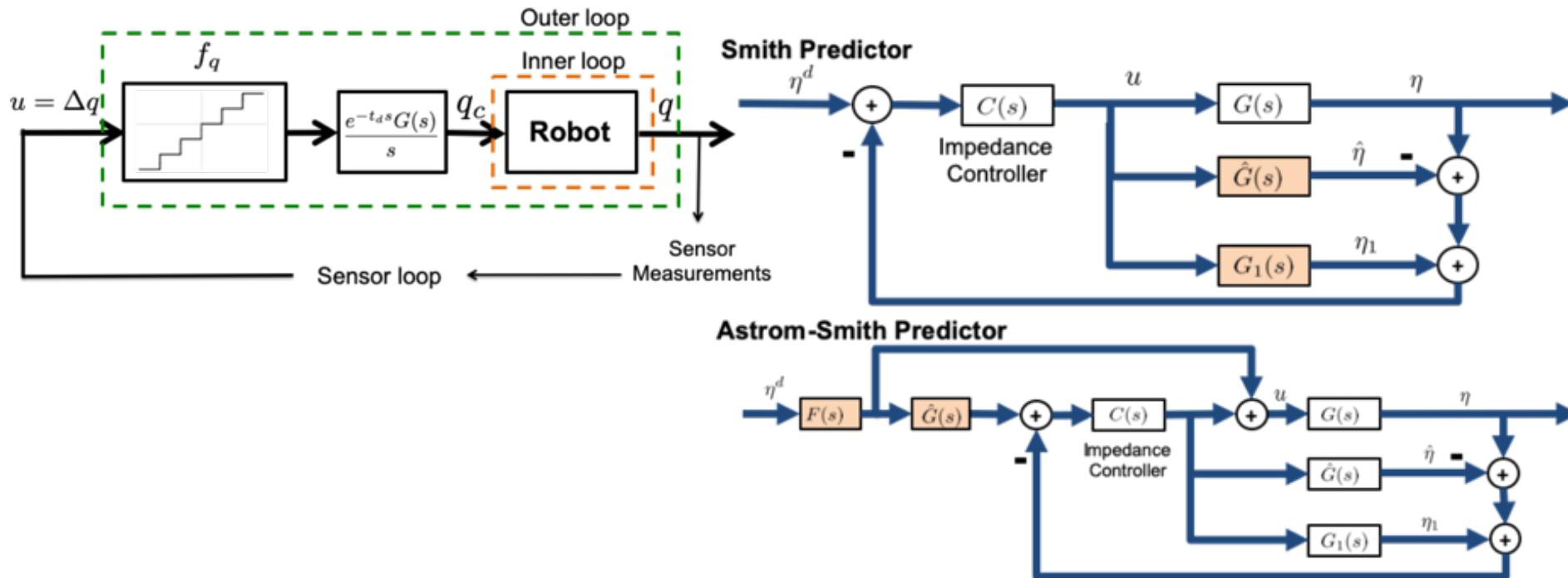
## Restore-L Mission

- Joint flexibility → bounded snap traj
- High-inertia load transport
- Compliance force control in 0-g
- Ground simulation



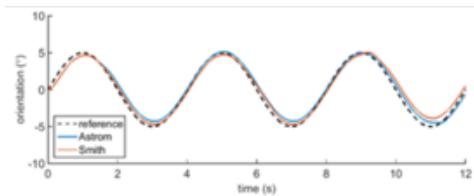
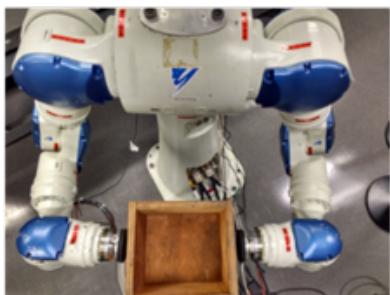
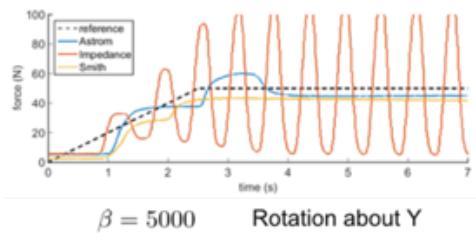
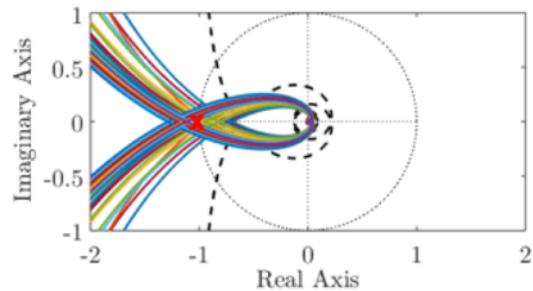
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# Impedance Control with Delay Compensation

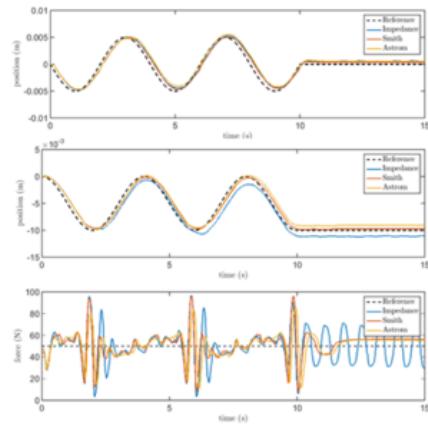
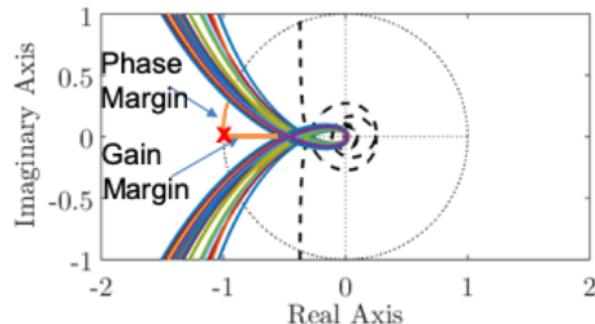


# Stability and Nonlinearity

Impedance Controller



Smith Predictor



# Assistive Robot

Human-robot collaboration  
Redundancy resolution



# Robot Software

Middleware for distributed communication and control:

- Smart Grid (DOE): Volttron message bus
- Robotics: Robot Operating System (ROS, ROS 2.0)
- Industrial Internet (GE/RTI): Data Distribution Service (DDS)
- Robot Raconteur (Wason Tech, RPI)

Distributed architecture: No master node

Object oriented: Client-service model

Ease of connection: no *a priori* info on data type and object

Plug-and-Play: Automatic node discovery

Security: user name and session authentication

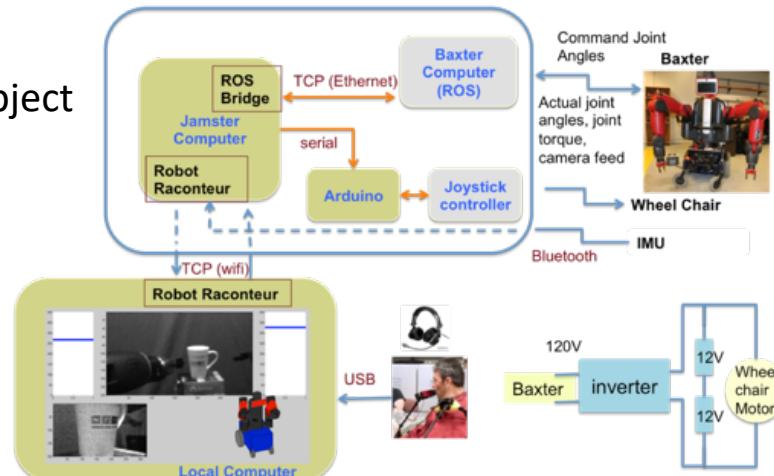
Interoperability: ROS bridge, MATLAB, LabView

Compatible OS: Windows, Linux, MacOS

Embedded platform: Raspberry Pi, Arduino

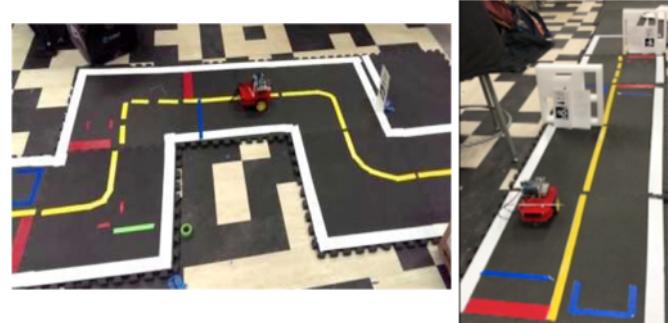
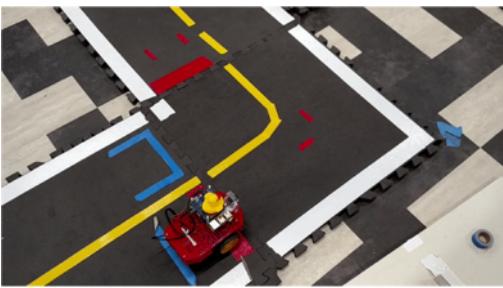
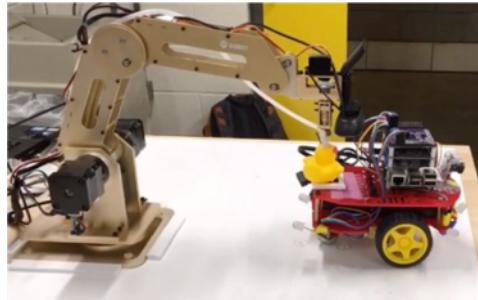
Open source, free download at [robotraconteur.com](http://robotraconteur.com)

Point-to-point  
(client-service)  
Data vs. Object



# Robot Education

- Duckietown, Dobot, ...



# Summary

- Opportunities for control community to advance robots in manufacturing
- Model-based optimization + sensor-based feedback control + data-driven learning → Improved manufacturing processes
- Numerous challenging manufacturing applications: surface modification, composites lay-up, robotic welding



# Acknowledgment

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- Robot assisted composites panel assembly: Shridhar Nath, Mark Vermilyea (GE), John Wason (Wason Technology), Matt Robinson, Levi Armstrong (SwRI), Glenn Saunders, William Lawler, Yuan-Chih Peng, Shuyang Chen, Dev Jivani, Rich Radke (RPI)



Supported by ARM Institute and NYSTAR: ARM-17-QS-F-01 Robot Assistant for Composite Manufacturing

- Satellite Servicing: David Cariabis, Kimberly Oakes (NASA Goddard)
- Assistive Robot: Dan Kruse (SRI Robotics), Lu Lu (NJIT) (Tufts University, Craig Nielsen Foundation)





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