



- Human error?, faulty sensing?
- Urgent issue or something to put on the 'to-do' list?



- Information overload or insufficient data?

**We have a long way to go in  
Monitoring, Diagnostics, Prognostics, and Decision-making...**



# Developing Measurement Science to Advance Monitoring, Diagnostics, and Prognostics in Manufacturing Operations

Brian A. Weiss, *Project Leader*  
Greg Vogl, Helen Qiao

Prognostics and Health Management for  
Reliable Operations in Smart Manufacturing

**Intelligent Systems Division**  
**Engineering Laboratory**  
**National Institute of  
Standards and Technology**



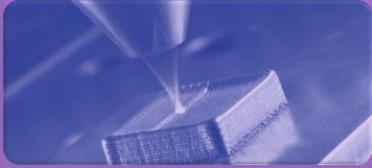
# Disclaimer



The views and opinions expressed herein do not necessarily state or reflect those of NIST. Certain commercial entities, equipment, or materials may be identified in this document to illustrate a point or concept. Such identification is not intended to imply recommendation or endorsement by NIST, nor is it intended to imply that the entities, materials, or equipment are necessarily the best available for the purpose.



# Agenda



NIST and Smart Manufacturing



Outreach



Research Focus



Use Case Development



Solution transfer to industry

# Engineering Lab's Mission...

- ...promote U.S. innovation and competitiveness by advancing measurement science, standards, and technology for engineered systems in ways that enhance economic security and improve quality of life
- Carry out mission related activities in...
  - Engineering and manufacturing products, processes, equipment, technical data, and standards
  - Manufacturing enterprise integration
  - Intelligent systems and control
  - Robotics and automation
  - Cyber-physical systems



Courtesy: Fotolia

# More on Measurement Science...

Used in the context of creating **critical-solution enabling tools** – metrics, models, and knowledge – for U.S. manufacturers. This includes:

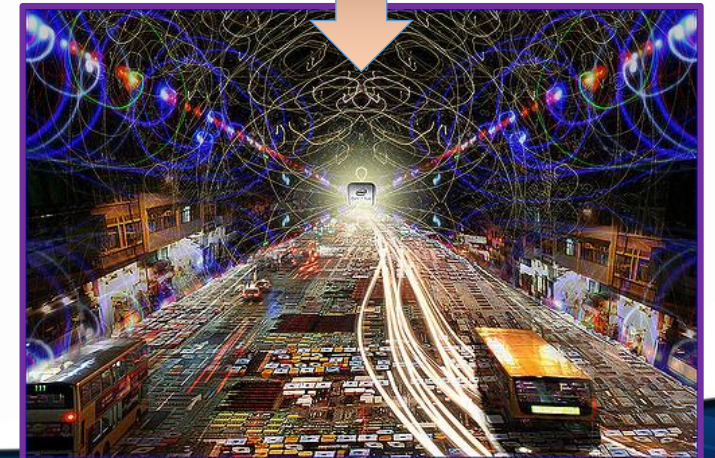
- Development of...
  - Performance metrics
  - Measurement and testing methods
  - Predictive modeling and simulation tools
  - Reference materials (e.g. data sets)
- Conduct inter-comparison studies and calibrations
- Evaluation of technologies, systems, and practices
- Development of the technical basis for standards, codes, guidelines, and/or practices



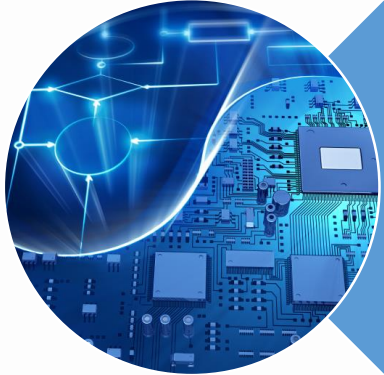


# Why Smart Manufacturing?

- Enable manufacturers to...
  - make what you want, where you want it, and when you want it.
  - **respond in real time to meet changing demands and conditions**
  - easily and rapidly reconfigure factory production and supply networks to optimize system performance
  - **deal effectively with uncertainty and abnormal events and learn from past experience to enable continuous improvement**
  - maintain seamless interoperability between factory processes and supply networks and between large manufacturers and small manufacturers



# Smart Manufacturing Programs



Model-Based  
Engineering



Trustworthy Systems,  
Components, and  
Data



Robotic Systems



Additive  
Manufacturing



# Trustworthy Systems, Components, and Data



Cybersecurity for Smart Manufacturing



High Performance Wireless Systems for Smart Manufacturing



Prognostics and Health Management for Reliable Operations in Smart Manufacturing



Supply Chain Traceability for Agri-Food Manufacturing

# Research Objective and Deliverables

To develop and deploy measurement science

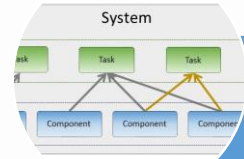
to promote the implementation, verification, and validation of advanced monitoring, diagnostic, and prognostic technologies to increase reliability and decrease downtime in smart manufacturing systems



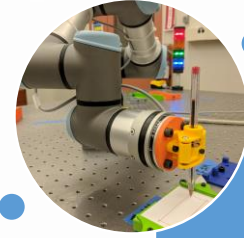
Standards and Guidelines

A screenshot of a data table with multiple columns and rows, likely representing a reference dataset or software tool configuration. The table has a header row and several data rows with various numerical and text values.

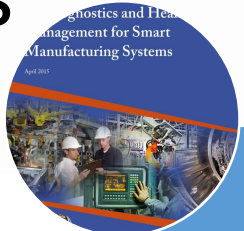
Reference Datasets and Software Tools



Test Methods and Performance Metrics



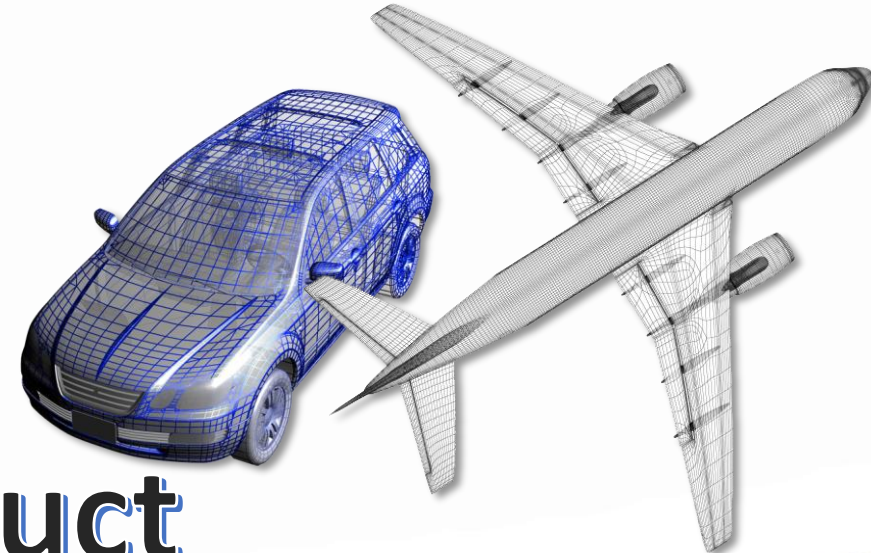
Use Cases and Test Scenarios



Roadmaps and Case Studies

- Range from very simple to very complex
  - Few to many moving parts
  - Few to many relationships among components, sub-processes, etc.
- Both consist of physical components that work together to produce one or more capabilities
- Physical components (and therefore, functional capabilities) will degrade over time
- Maintenance may or may not be required throughout its life

## Process

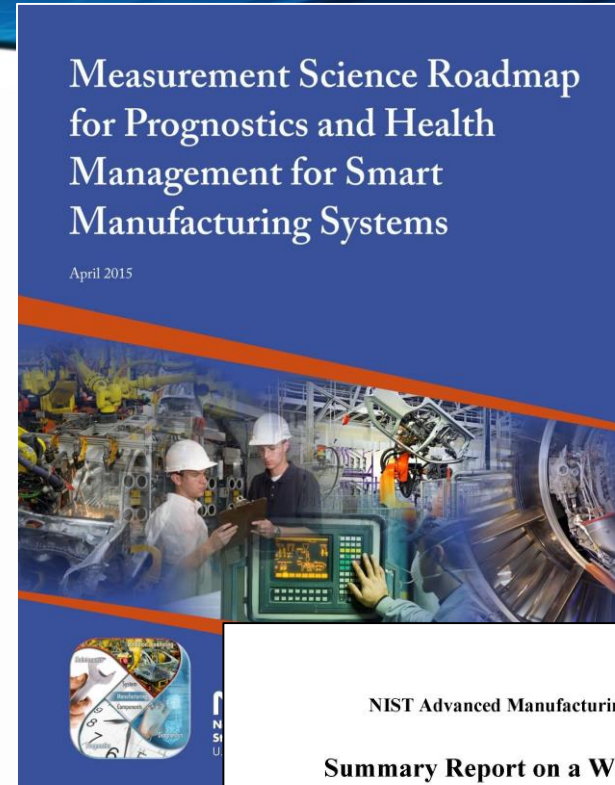


## Product



# How do we know this is Important?

- Measurement Science Roadmapping Workshop
- Manufacturing Standards Requirements Gathering Workshop
- Collaborative studies with university and industry partners
- Interactions with various technical organizations



# Outreach: Roadmapping Workshop and Report

**Goal: Identify goals, desired capabilities, and challenges to develop necessary measurement science**

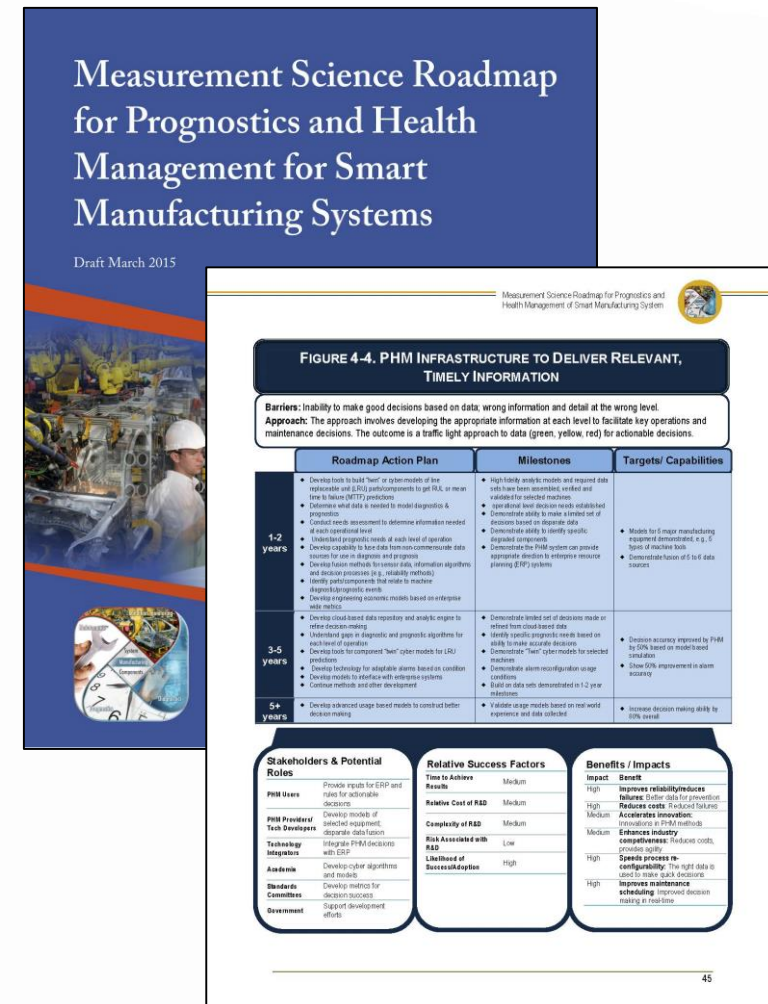
## The Workshop

- Featured over 70 people from 35 organizations including manufacturers, technology vendors, academia, standards development organizations, and other government agencies

## The Report

- Extensively documents breakout groups' efforts to identify goals, desired capabilities, challenges and barriers, and priority roadmap topics relating to three critical PHM topic areas

- <http://www.nist.gov/el/isd/phm4sms-workshop.cfm>



# Workshop Roadmap Action Plans

## *PHM Manufacturing Process Techniques and Metrics*

- **Advanced Sensors for PHM in Smart Manufacturing**
- **PHM Data Format, Taxonomy, and Architecture**
- Enterprise-wide PHM for Maintenance Planning

## *PHM Performance Assessment*

- **Overarching Architecture Framework for PHM with Standards and KPIs**
- **Identification of PHM Performance Metrics**
- **Failure Data for Prognostics and Diagnostics**
- **Determination of PHM Data and Information Needs**
- Cost model for PHM Performance
- Taxonomy for Applications

## *PHM Infrastructure – Hardware, Software, and System Integration*

- **Open-Source Community for PHM**
- **PHM Infrastructure to Deliver Relevant, Timely Information**
- Embedded Sensors for PHM of Emerging Manufacturing Technologies
- PHM as an Equipment Design Feature



# Outreach: Sensing, Diagnosis, and Prognosis in Manufacturing

- Goal: Identify industry needs and priorities
- Approach: Industrial case studies
  - SMEs versus Large Manufacturers
  - Collaboration with UC/UM IMS
- Key findings:
  - Real data illuminates actual performance
  - Initial area of interest is equipment utilization, but interest quickly shifts to diagnosis, prognosis, and dynamic scheduling
  - Large barriers include systems integration and lack of sufficient data to support analysis



- X. Jin, B.A. Weiss, D. Siegel, J. Lee, and J. Ni, "Present Status and Future Growth of Advanced Maintenance Technology and Strategy in US Manufacturing," *Special Issue: PHM for Smart Manufacturing Systems*, vol. 7 Special Issue, pp. 17-34, September, 2016.
- X. Jin, D. Siegel, B.A. Weiss, E. Gamel, W. Wang, and J. Ni, "The Present Status and Future Growth of Maintenance in U.S. Manufacturing: Results from a Pilot Survey," *Manufacturing Review*, vol. 3, June, 2016.
- M. Helu and B.A. Weiss, "The Current State of Sensing, Health Management, and Control for Small-to-Medium-Sized Manufacturers," *ASME 2016 Manufacturing Science and Engineering Conference (MSEC2016)*, 2016.
- G. Vogl, B.A. Weiss, and M. Helu, "A Review of Diagnostics and Prognostic Capabilities and Best Practices for Manufacturing," *Journal of Intelligent Manufacturing*, pp. 1-17, June, 2016.

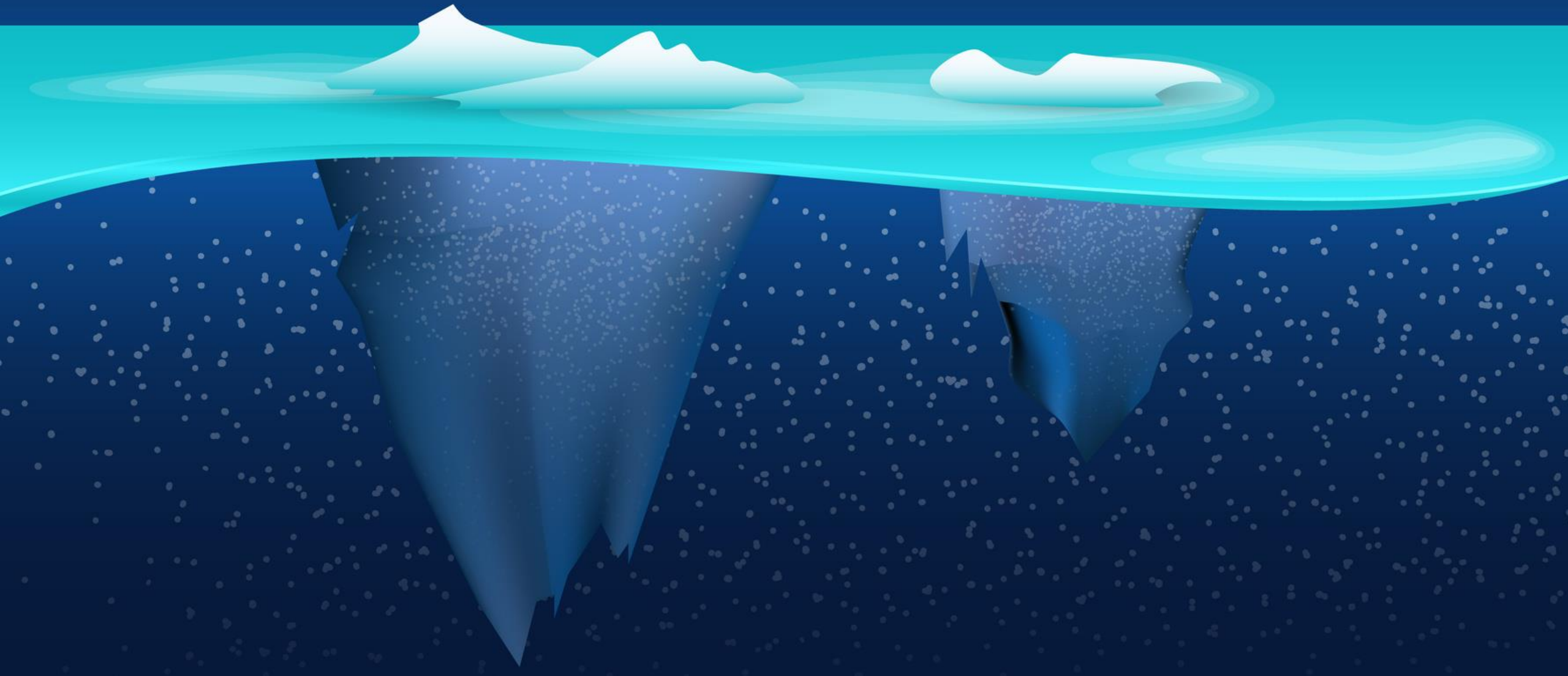


# Reference Data Challenge – *How much is enough?*



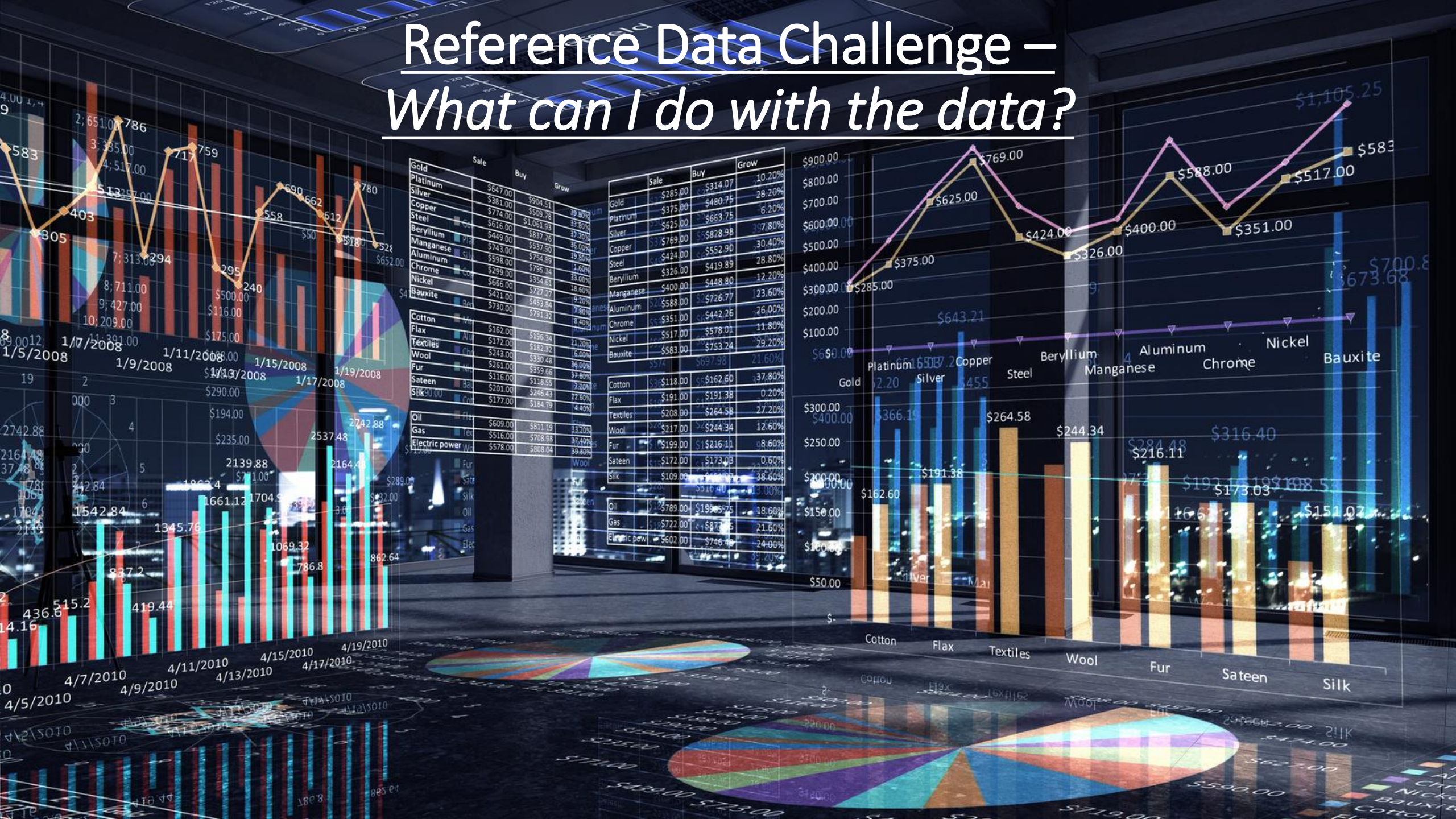


Reference Data Challenge –  
*Do I truly understand my data?*





# Reference Data Challenge – *What can I do with the data?*

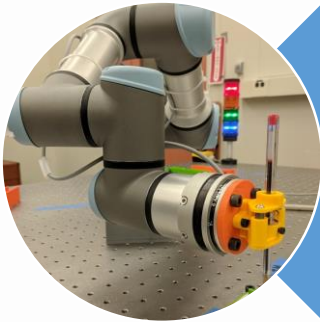




# Reference Data Challenge – *How do I quantify the human impact?*

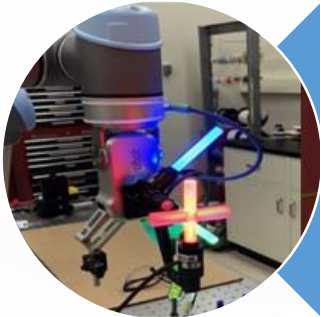


# Research Levels and Testbeds



## Identification of Robot Workcell Degradation

- Work Cell-Level Research
- PHM for Robot Systems Lab/Testbed



## Assessment of Robot Accuracy Degradation

- Robot-Level Research
- PHM for Robot Systems Lab/Testbed



## Machine Tool Linear Axes Diagnostics and Prognostics

- Component-Level Research
- Linear Axis Test bed & 'Shops' Machine Tools



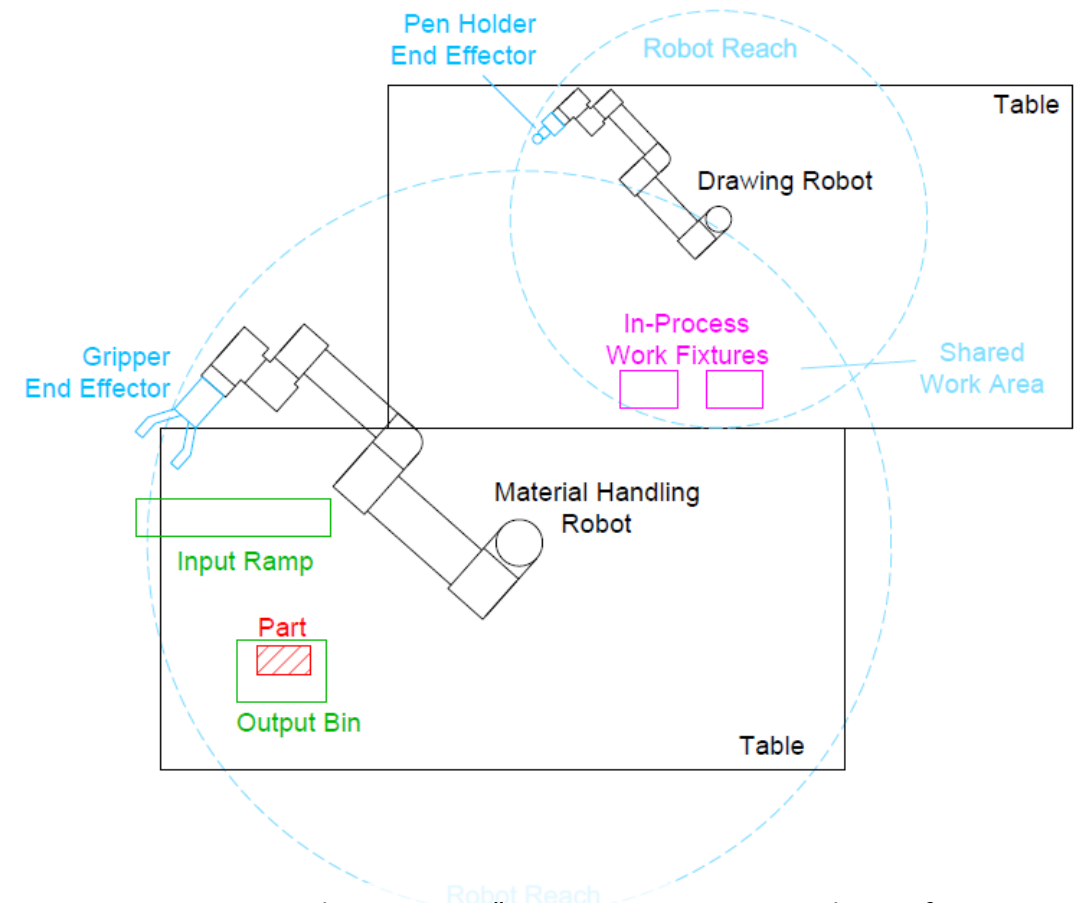
# Research Focus and Plan

## Verification & Validation of Equipment and Process Health Tech...

- Identify sources throughout the workcell that influence performance on positioning tasks using an artifact-based approach
- Enhance capability to measure end-of-robot-arm positional accuracy through development of active smart targets and corresponding test procedures
- Extend single-axis diagnostics to multi-axis prognostics of machine tool linear stages
- Generate datasets from NIST testbeds and external pilot sites
- Contribute to development of guidelines within an ASME Standards subcommittee on Advanced Monitoring, Diagnostics, and Prognostics

# NIST PHMC Robotics Testbed

- 2 robots
  - Material Handling
  - Path Following
- End Effectors
  - Electric Parallel Finger Gripper
  - Pen holder
- Fixtures
- Parts



- A. Klinger and B.A. Weiss, "Robotic Work Cell Test Bed to Support Measurement Science for Monitoring, Diagnostics, and Prognostics," *ASME 2018 International Manufacturing Science and Engineering Conference (MSEC2018)*, June 2018
- B.A. Weiss and A. Klinger, "Identification of Industrial Robot Arm Work Cell Use Cases and a Test Bed to Promote Monitoring, Diagnostic, and Prognostic Technologies," *Annual Conference of the Prognostics and Health Management Society 2017*, October 2017.
- B.A. Weiss, M. Helu, G. Vogl, and G. Qiao, "Use Case Development to Advance Monitoring, Diagnostics, and Prognostics in Manufacturing Operations," *12th International Federation of Automatic Control (IFAC) Workshop on Intelligent Manufacturing Systems IMS 2016*, December 2016.

# Kinematic Breakdown to Assess Workcell Health

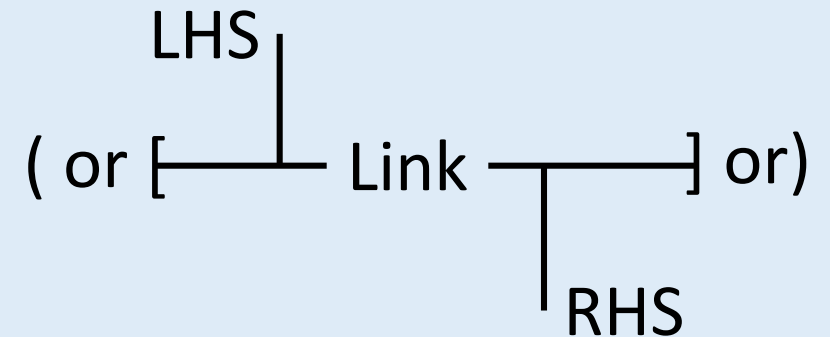
- Flexible method to identify sources of positioning repeatability degradation within a robotic workcell beyond the robot
  - Robots
  - End effectors
  - Fixtures
  - Etc.
- Low Cost, Minimally Invasive, In-Process Test
- Identify degradation before fault or failure
- Measure rate of degradation before fault or failure



# Kinematic Links

- Links can be defined at any level of detail
  - Component
  - Sub-system
  - Physical mechanical link
- Each link has a left-hand-side (LHS) and a right-hand-side interface (RHS)
- Intermittent and non-intermittent interfaces can be represented with “()” and “[ ]” respectively
- Textual representation

## Graphic Representation

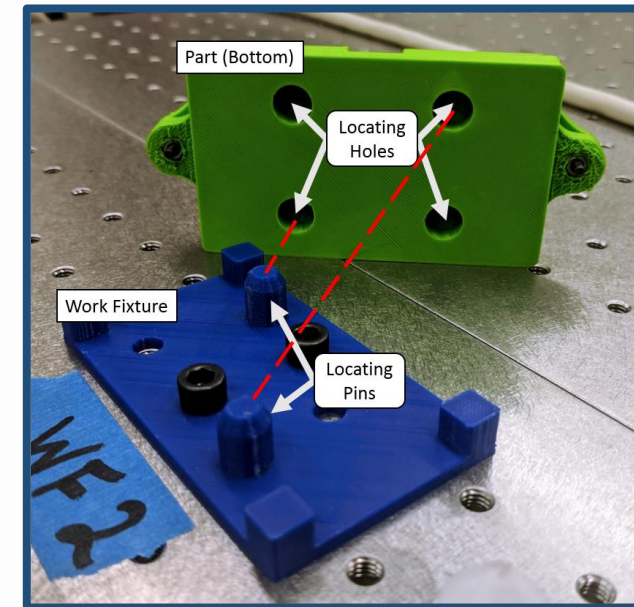
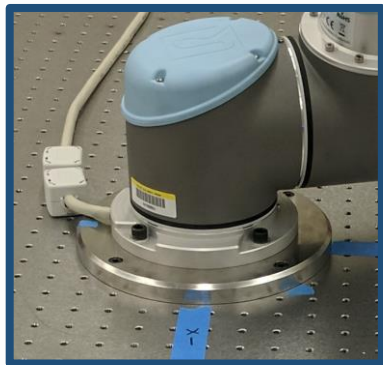
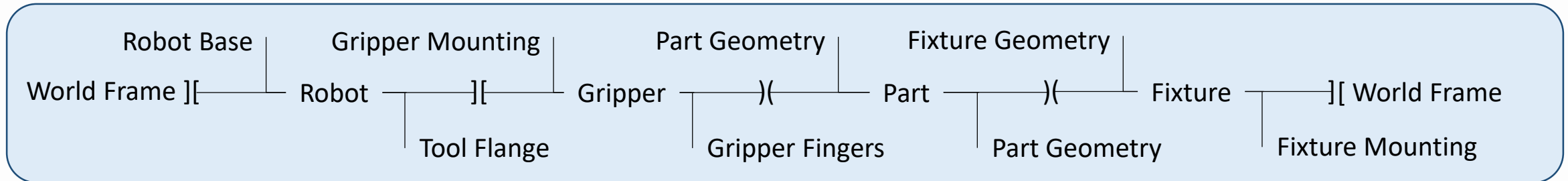


## Textual Representation

( or [ LHS, Link, RHS ] or )

# Material Handling Robot Kinematic Chain

## Graphic Representation



# Use Case Degradation Modes

- Assume each link can cause positioning degradation due to deformation
- List relationships between links
- Identify positioning degradation modes which may be present during operation

Kinematic Chain Section	Degradation Mode(s)
World Frame ][ Robot Base	Loose Connection
[ Robot Base, Robot, Robot Tool Flange ]	Robot Wear
Robot Tool Flange ][ Gripper Mounting	Loose Connection
[ Gripper Mounting, Gripper, Gripper Fingers ]	Gripper Wear
Gripper Fingers )( Part Geometry	Finger Positioning, Bad Part*
[ Part ]	Bad Part*
Part Geometry )( Fixture Geometry	Bad Part*, Bad Fixture**
[ Fixture Geometry, Fixture, Fixture Mounting ]	Bad Fixture**
Fixture Mounting ][ World Frame	Loose Connection

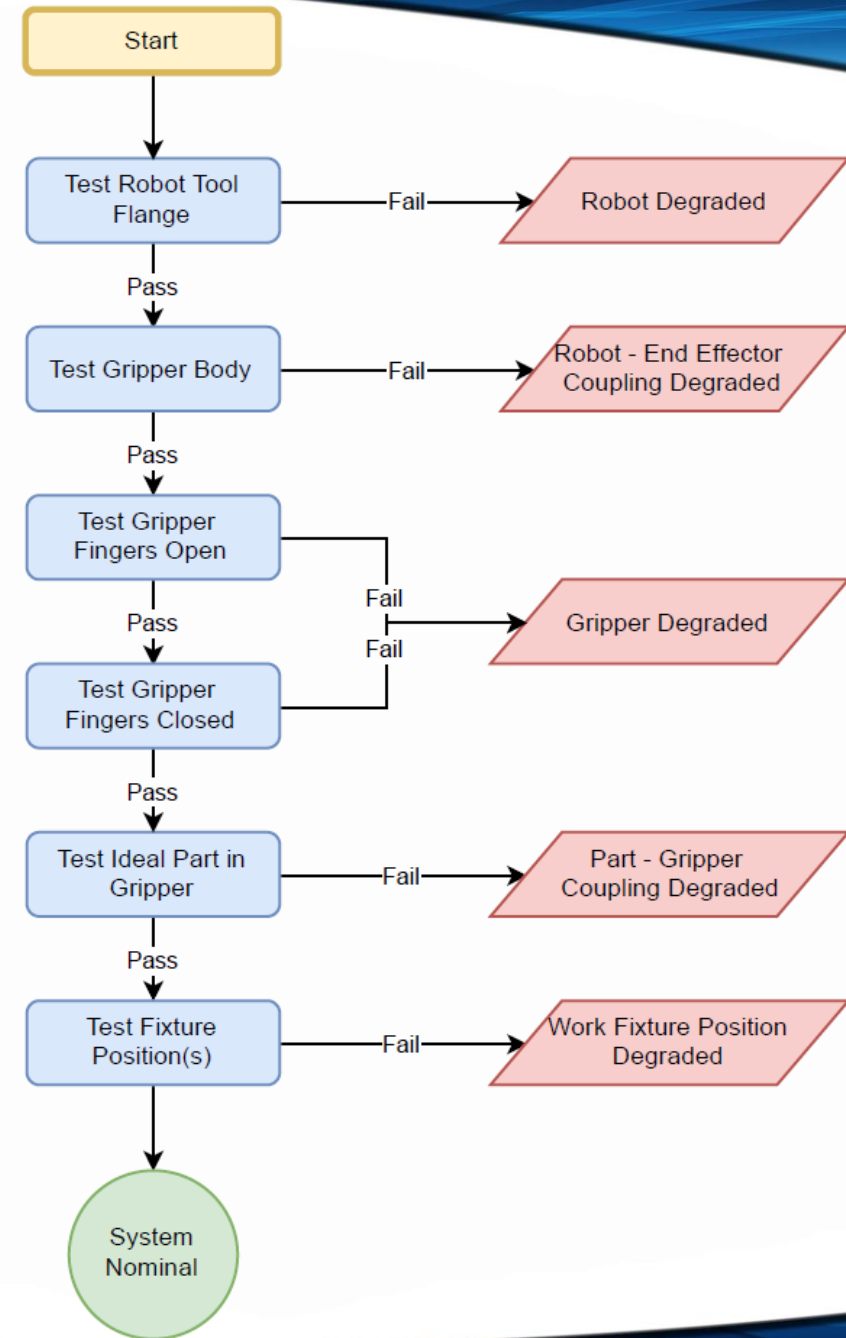
\*From either part inaccuracy or damage

\*\*From either fixture positioning inaccuracy or fixture wear/damage



# Test Decision Tree

Test in order working away from the reference frame / measurement device



# Testbed Application Example

- Implemented on testbed with inhouse designed sensor
- Measurement points:
  - Robot tool flange
  - Gripper Body
  - Fingers, Open
  - Fingers, Closed
  - Grasped Ideal Part
  - TODO - Fixtures

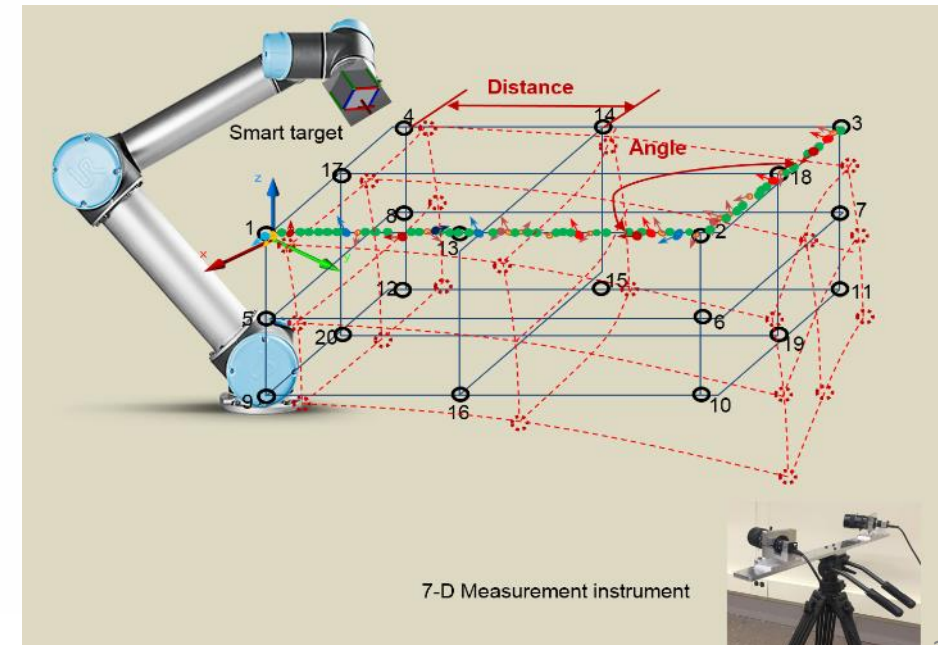
The screenshot displays a control interface for a testbed application. At the top, a 'Process Status' box shows 'Assigning Tasks' with a green indicator. To its right, a table shows 'Completed Part Count' as 1 and 'Average Part Time' as 56. Below this is a navigation bar with tabs for 'System Process', 'Position Check', 'Force and Tor...', and 'System Settings'. The main area is titled 'Position Check' and contains several sections:

- Enabling / Disabling Position Checking:** A table with two rows: 'Enable / Disable' (green) with the note 'MUST BE ENABLED FOR THRESHOLD AND FORCED CHECKS', and 'Force Check Next' (red) with the note 'WILL PERFORM CHECK OPERATION ONCE AS NEXT TASK'.
- Settings:** A table for 'Part Between Checks Threshold' set to 100, with buttons for -10, -1, Set at 50, +1, and +10.
- Status:** A table showing 'Parts Since Last Check' as 0 and 'Position Check Assigned Next'. Below it, a 'Sensor State' indicator shows a red bar.
- Diagnostics:** Two columns of status indicators. The first column, 'UR3', shows 'End Effector' and 'Robot' as red. The second column, 'UR5', shows 'Robot', 'Fingers Open', 'Fingers Closed', 'Part Grasped', and 'Gripper Body' as green. Below this, a 'Work Fix' section shows 'WF1' and 'WF2' as red.

At the bottom, a diagram titled 'UR5 Kinematic Chain Testing' shows a flow from 'Robot' to 'Gripper Body', which then branches into 'Fingers Open' and 'Fingers Closed'. 'Fingers Closed' is further connected to 'Grasped Part'.

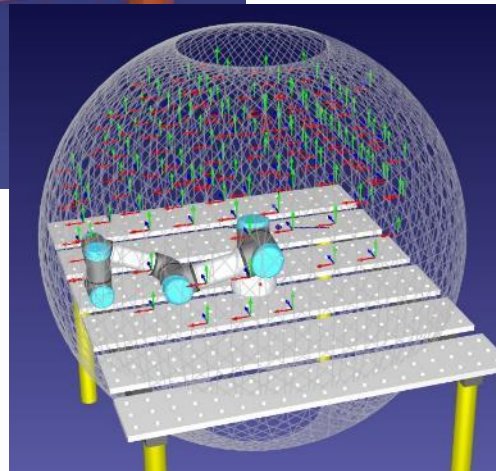
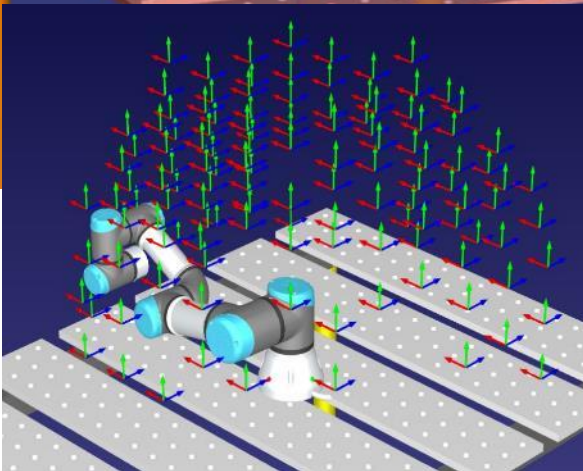
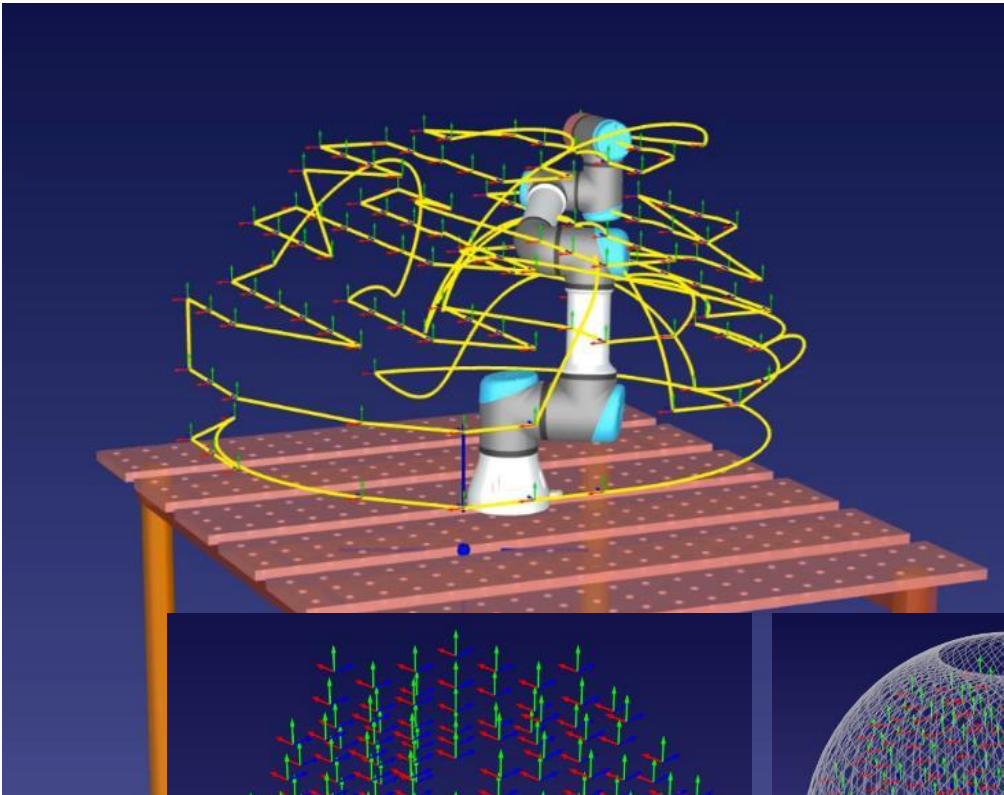
# Robot Level Research – Assessment of Robot Accuracy Degradation

- **Research Objective:** Develop a quick health assessment methodology to enable manufacturers to assess the health of their robot arms by monitoring accuracy degradation
- **Key Output to Date:**
  - Algorithms and test method for quick robot position and orientation accuracy assessment
  - Advance sensing - 7-D measurement system
  - Innovative target – smart target
- **Impact:** Reference test methods will educate and guide manufacturers in deploying PHM to quickly assess robot health promoting greater employment of predictive maintenance strategies (e.g. robot system calibration, joint and gear box replacement etc.) that will increase efficiency and productivity while decreasing downtime.

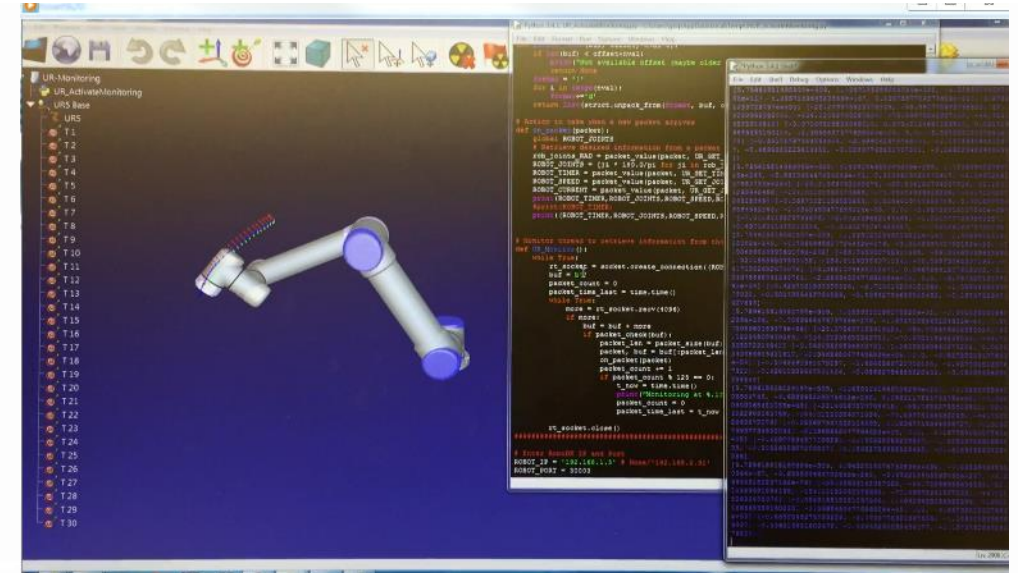




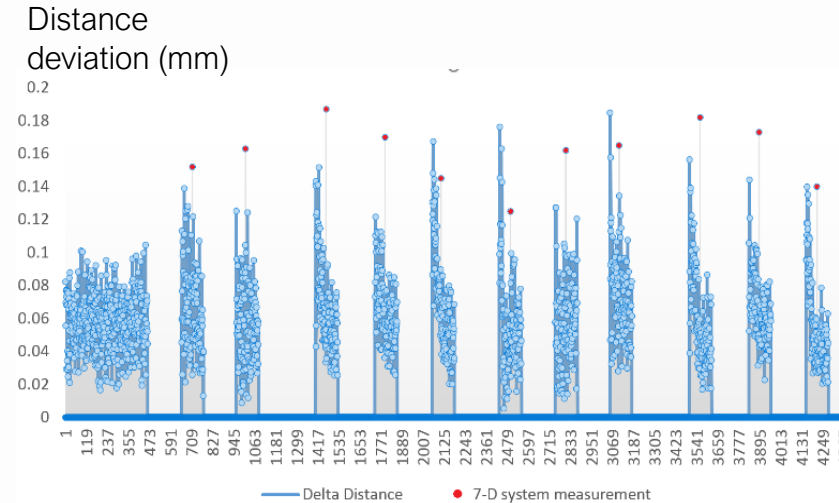
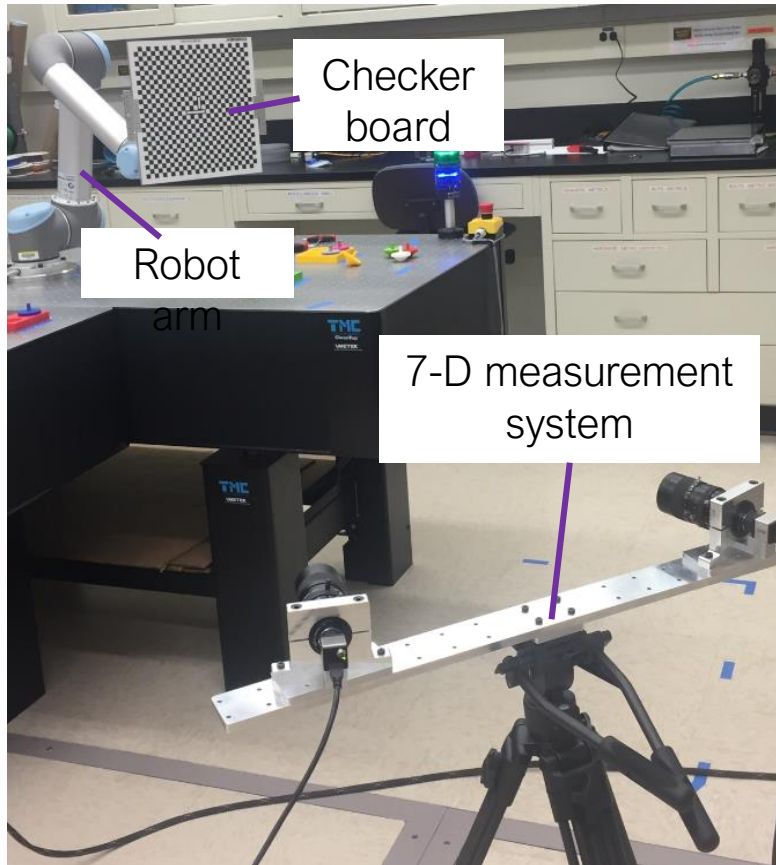
# Test Method Development and Reference Data Collection



Real-time controller data collection



# Robot – Reference Data Sets



TCP deviations: 7-D system measured vs. calculated deviations from controller actual joint positions minus target joint positions

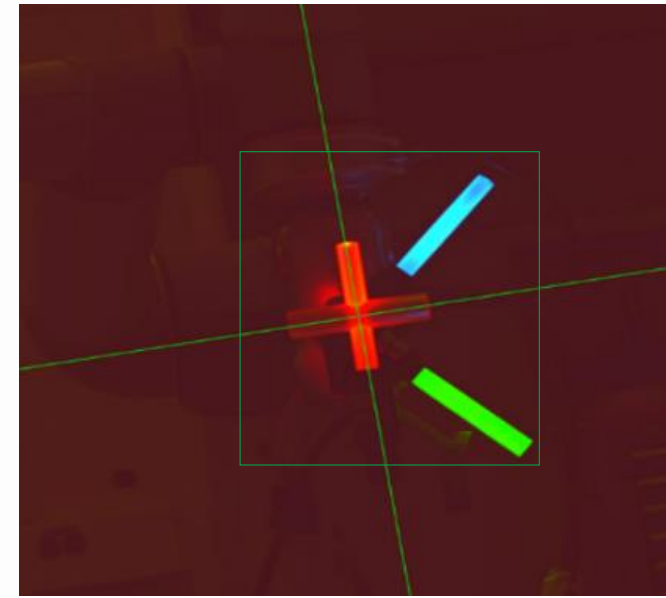
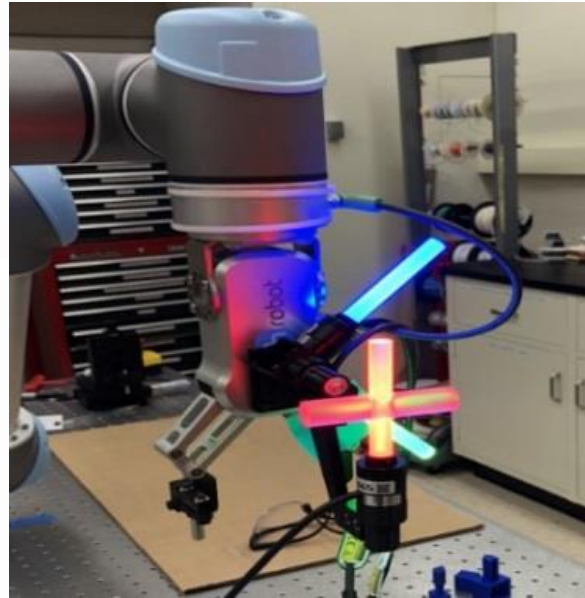
Reference data set URL:

<https://www.nist.gov/el/intelligent-systems-division-73500/cognition-and-collaboration-systems/degradation-measurement>



# 6-D Smart Target for Vision-based Measurement

- **What is it?** – A smart target (patent pending) that can be integrated with vision-based measurement instruments to acquire six-dimensional (6-D) information (x, y, z, pitch, yaw, and roll) of a moving object with high accuracy (accuracy down to  $50\mu\text{m}$ ).





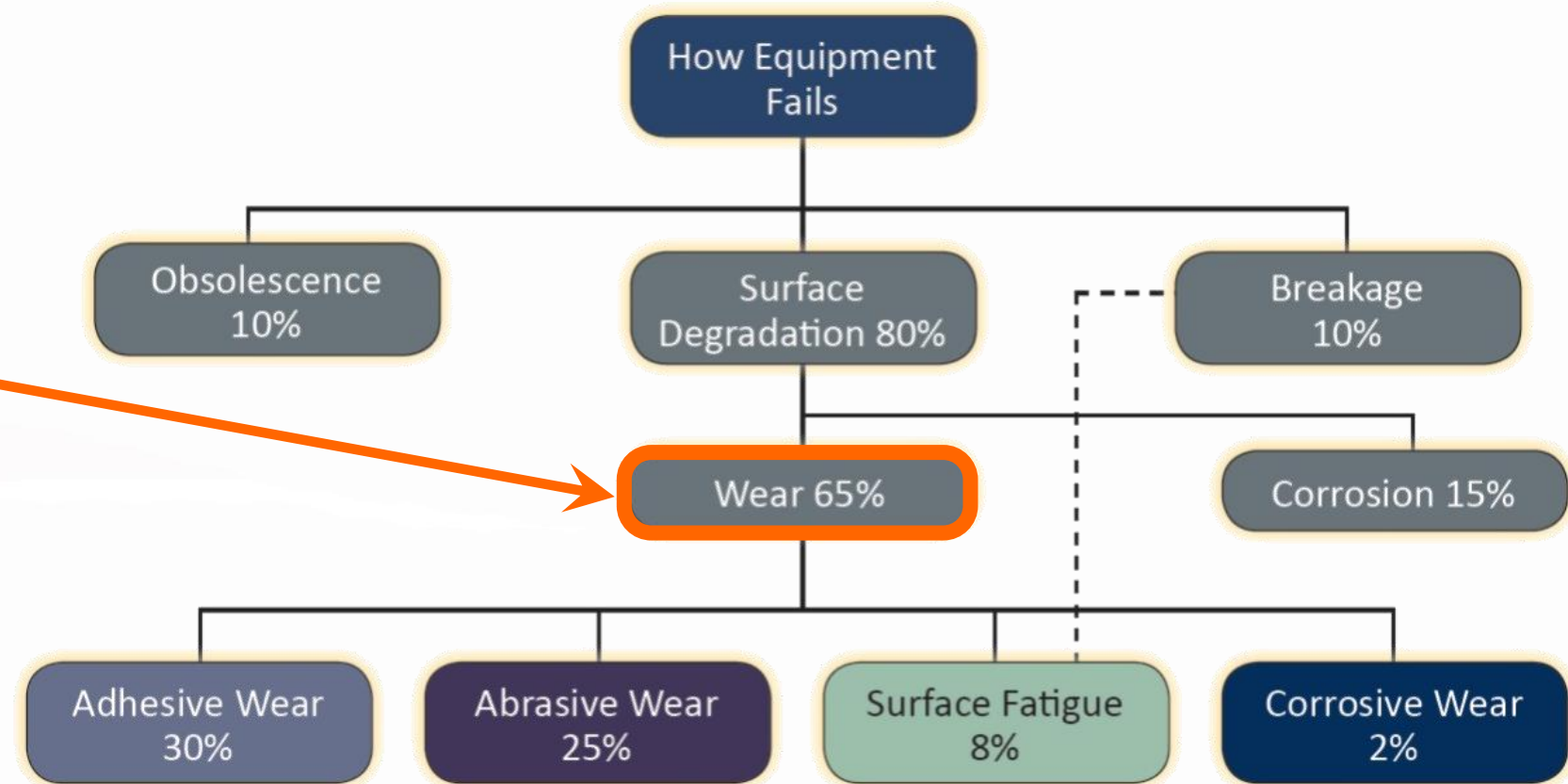
# Problem = Unplanned Downtime

- Faults/failures → 10s of \$Billions per year (> new machines!)
- Machine tool degradation causes performance changes and unplanned downtime

## Wear



[Machinery Lubrication \(2004\), \*Wear in Rolling Element Bearings and Gears\*](#)



[Reliabilityweb.com \(2018\), \*Lubrication FMEA: The Big Picture\*](#)

# Why Not Measure Health?

- Major manufacturers say routine tracking of performance is **too expensive**
- Accuracy a pro, but setup and operation time/cost a con
  - Offline
  - Lack of periodic data
  - Expensive

Laser → 1-2 days



[API](#)



Cap probes → hours



[IBS Precision Engineering](#)

Ballbar → 1 hour

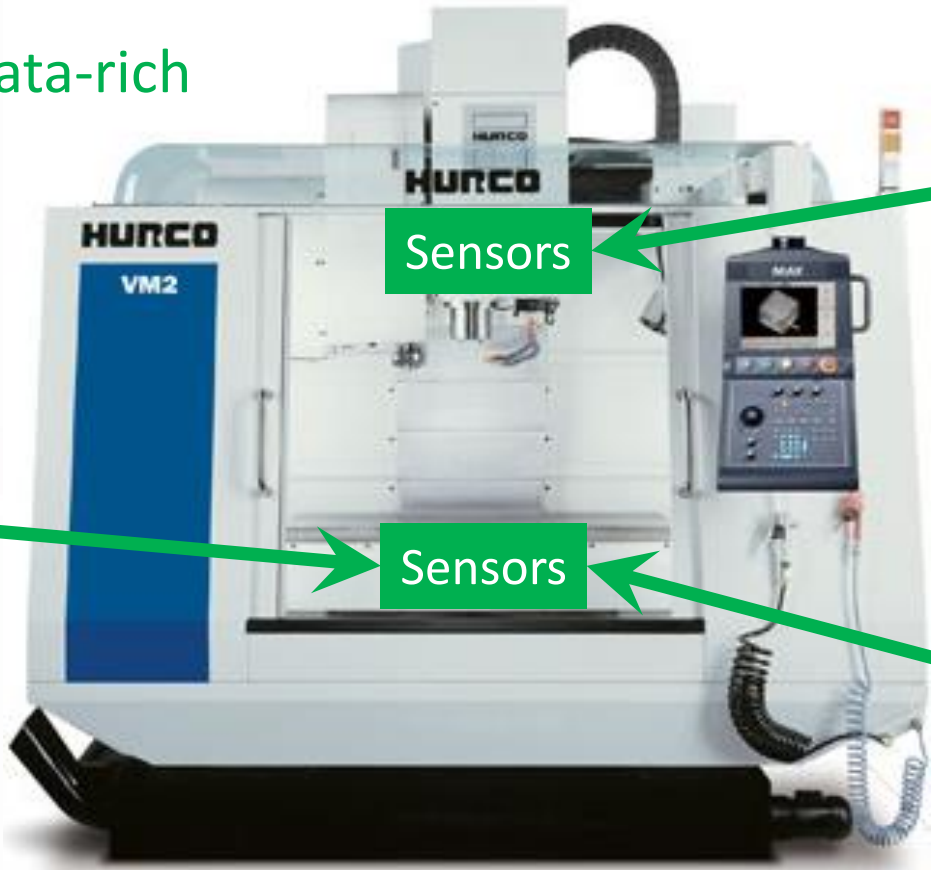


[Renishaw](#)

# GOAL: Smart Machine Tools

- Industry challenge: “Machine health in 5 min?”
- On-machine measurement science to diagnose performance and root-causes
  - ~~Offline~~ Online
  - ~~Lack of periodic data~~ Data-rich
  - ~~Expensive~~ Inexpensive

Linear Axis Health Tracking  
[How?]

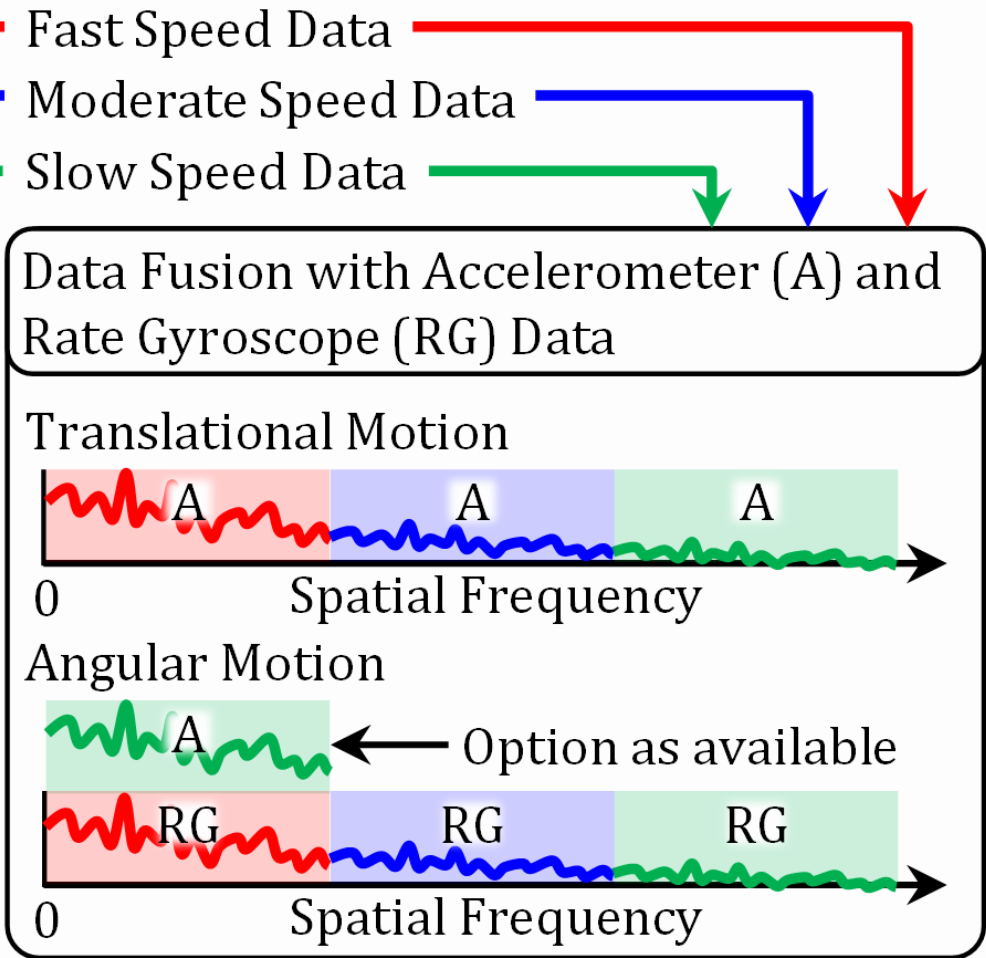
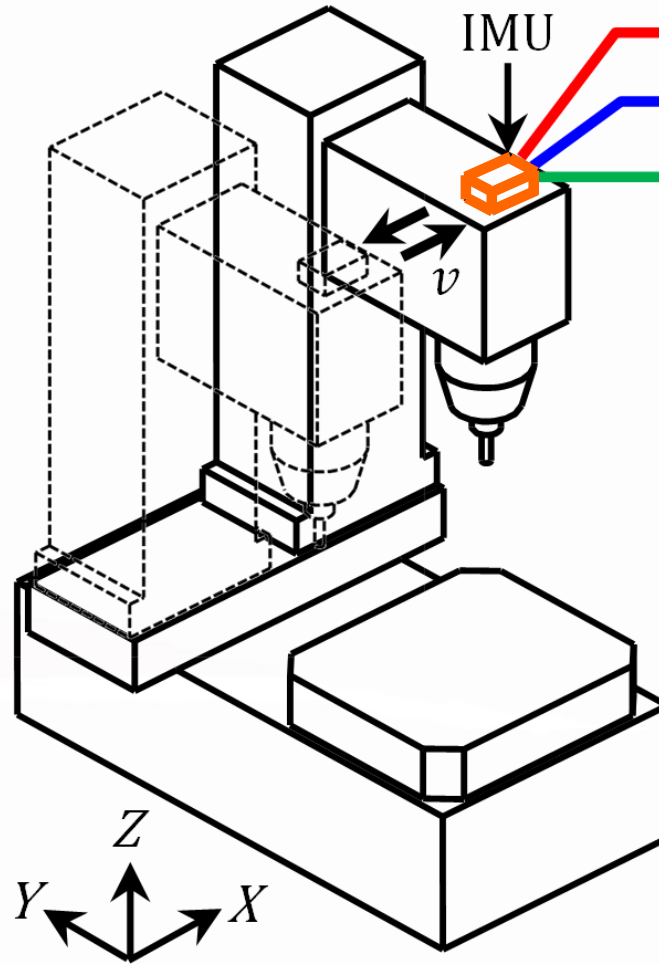
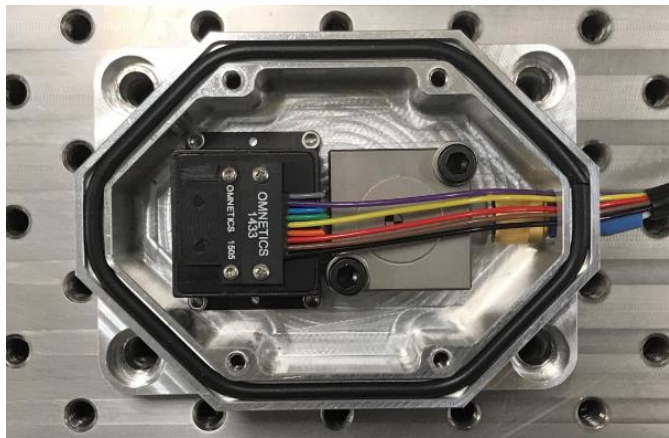
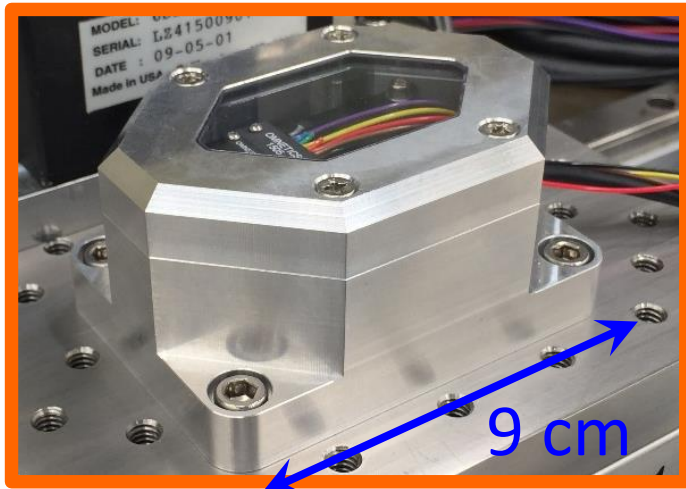


Spindle Health Tracking  
[How?]

Squareness Health Tracking  
[How?]



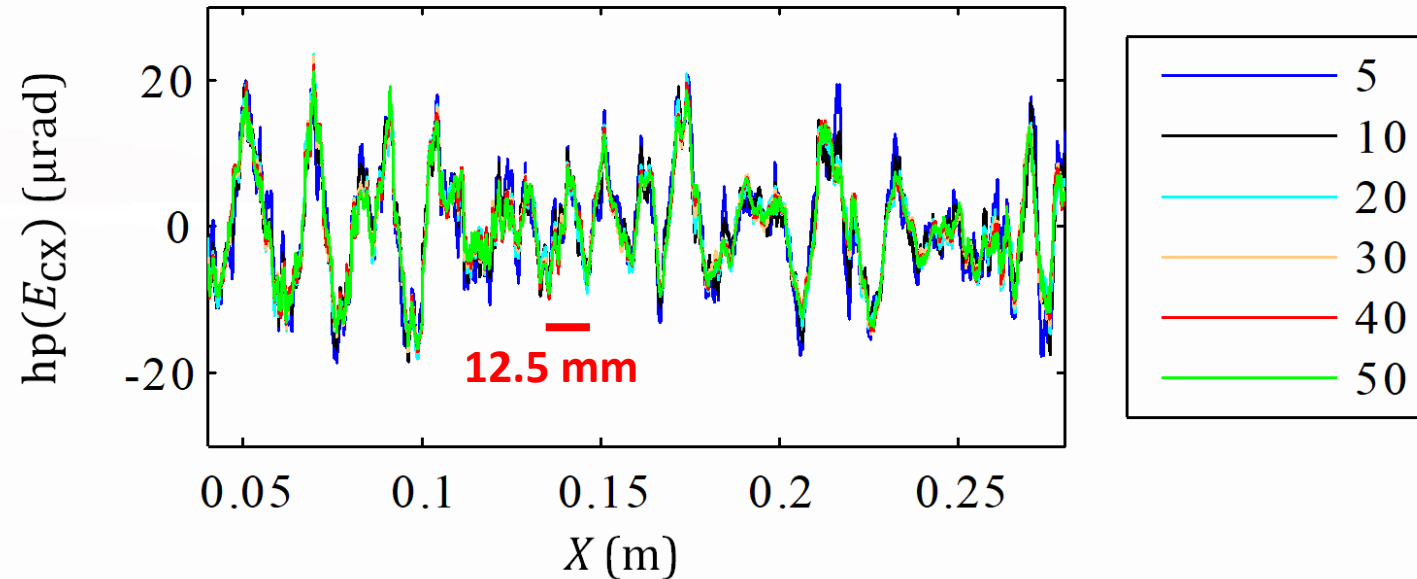
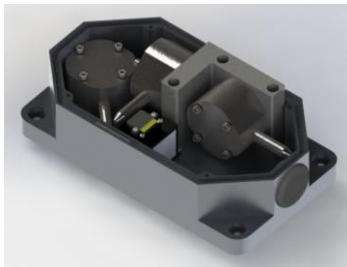
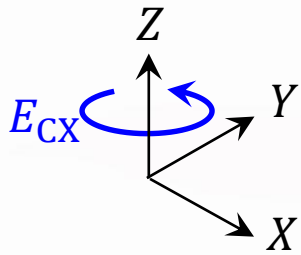
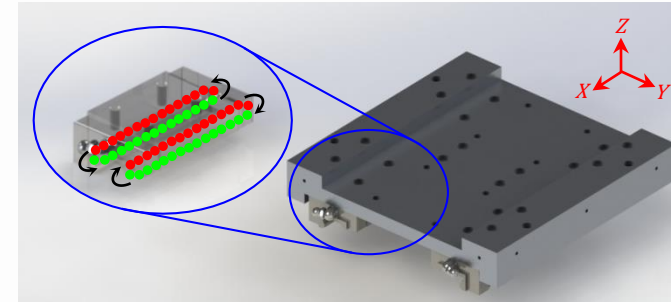
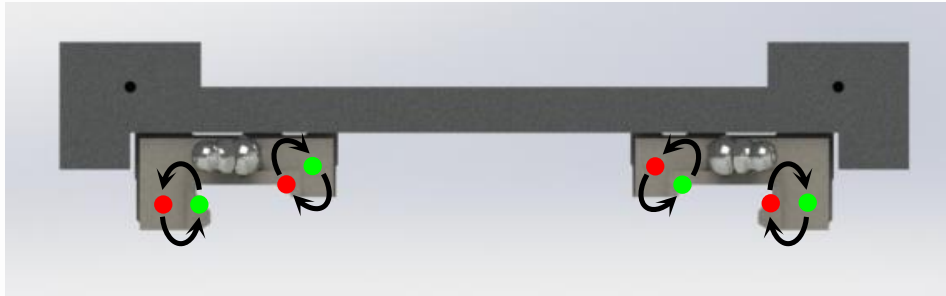
# IMU for Linear Axis Monitoring



- G. Vogl, R. Pavel, A. Archenti, T. Winnard, M. Mennu, B.A. Weiss, and M.A. Donmez, "Identification of machine tool geometric performance using on-machine inertial measurements," *6th International Conference on Virtual Machining Process Technology (VMPT 2017)*, May 2017.
- G. Vogl, M.A. Donmez, A. Archenti, and B.A. Weiss, "Inertial Measurement Unit for On-Machine Diagnostics of Machine Tool Linear Axes," *Annual Conference of the Prognostics and Health Management Society*, October 2016.

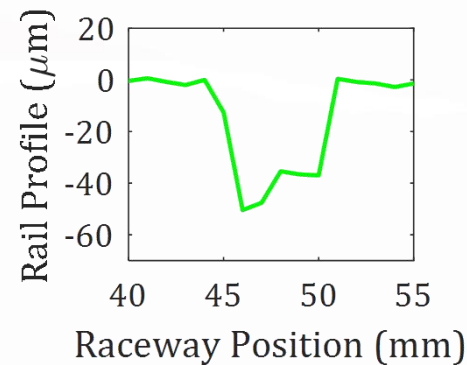
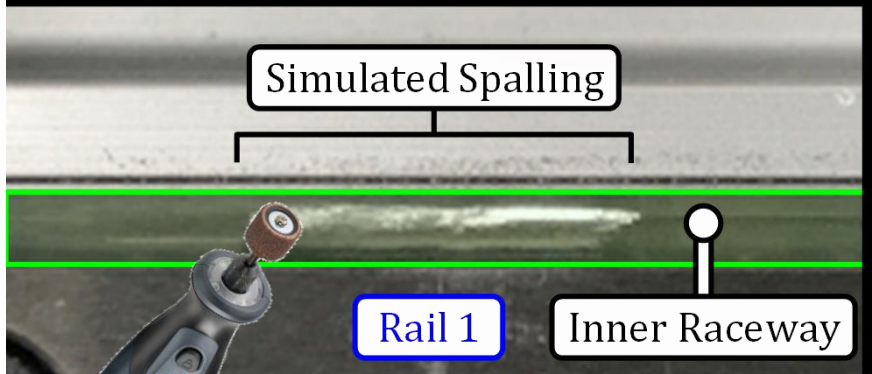
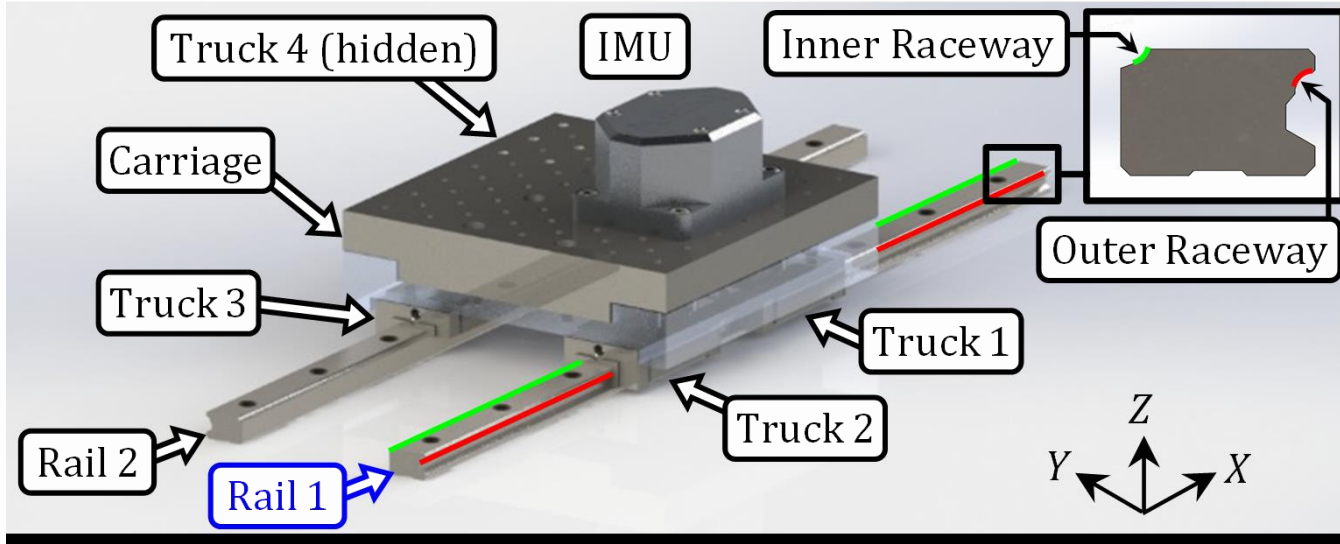
# Performance Tracking to Diagnostics

- 3.97 mm DIA  $\times \frac{\pi}{2}$  12.5 mm ball-passing distance

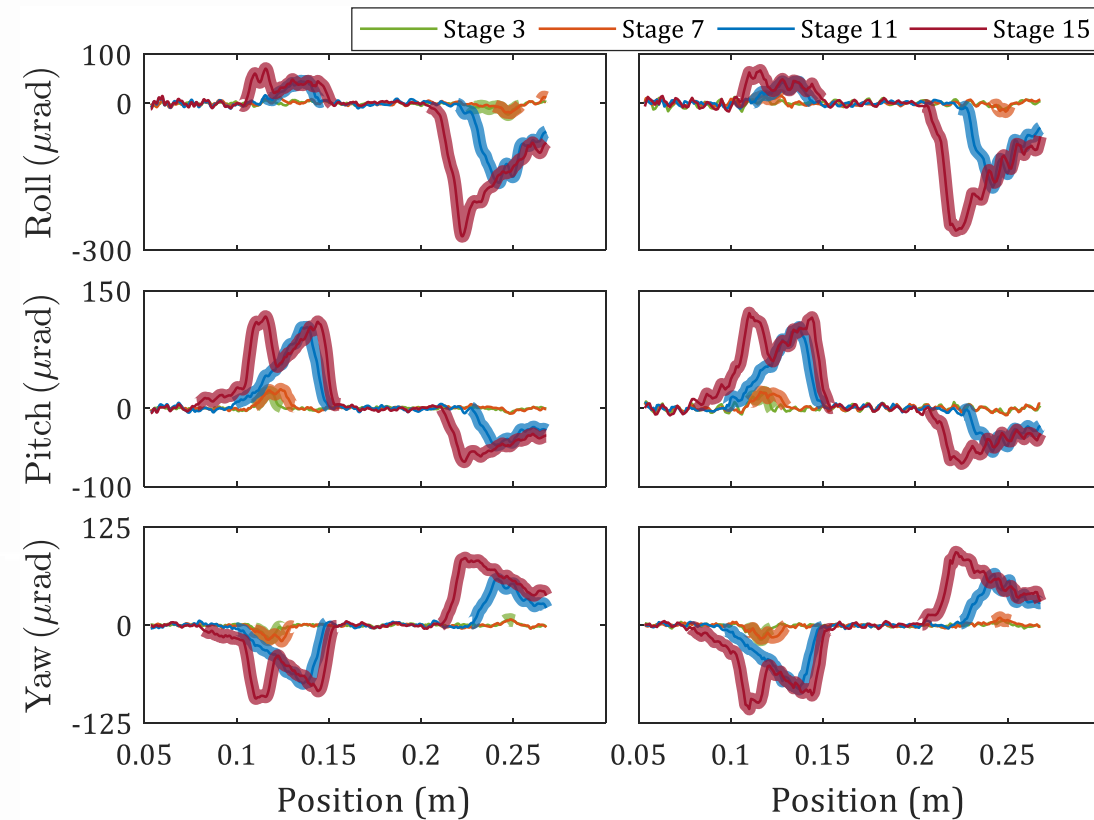


# NIST Linear Axis Testbed

- Testbed to study IMU-based method & diagnostics / root-cause analysis



## $\Delta$ Error vs. Position



IMU

REF



# Further Enabling PHM...

- Aid manufacturers in designing, deploying, verifying, and validating PHMC strategies within their manufacturing operations



# Questions to Answer During PHM Design & Deployment

- What physical or task degradation has the potential to impact the metrics I care about most in my process?  
*What health degradations can impact my quality, productivity, scrap, etc.?*
- What data, leading to intelligence, do I need about my process to determine where and when health degradation will occur?  
*What can be monitored and how?*
- How do I prioritize the risk of faults and failures in my system and process?  
*Where should I deploy PHM since I can't put it everywhere?*
- How does the health of my physical system, and its constituent elements, influence the health of my process?  
*How can I map the relationships between the physical and functional to better understand my process?*

**FAILURE**



**SUCCESS**



# Next Steps

- Continue research in the machine tool and robotics domains
- Output additional reference datasets
- Pilot test methods and protocols in manufacturing environments
- Contribute to standards development
- Be responsive to the manufacturing community's needs





Brian A. Weiss  
Intelligent Systems Division  
[brian.weiss@nist.gov](mailto:brian.weiss@nist.gov)  
(301) 975-4373

NIST  
100 Bureau Drive, MS 8230  
Gaithersburg, MD 20899



## *Prognostics and Health Management for Reliable Operations in Smart Manufacturing*

[www.nist.gov/el/isd/ks/phmc.cfm](http://www.nist.gov/el/isd/ks/phmc.cfm)