



# Building BRAINs for Autonomous Systems

Microsoft Autonomous System Toolchain Overview

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# Evolution of industrial systems

## Manual systems



Labor intensive  
Error prone  
Difficult to scale

## Automated systems



Task specific installations  
Effective and scalable  
Programmed behavior  
Limited in scope and flexibility

## Autonomous systems

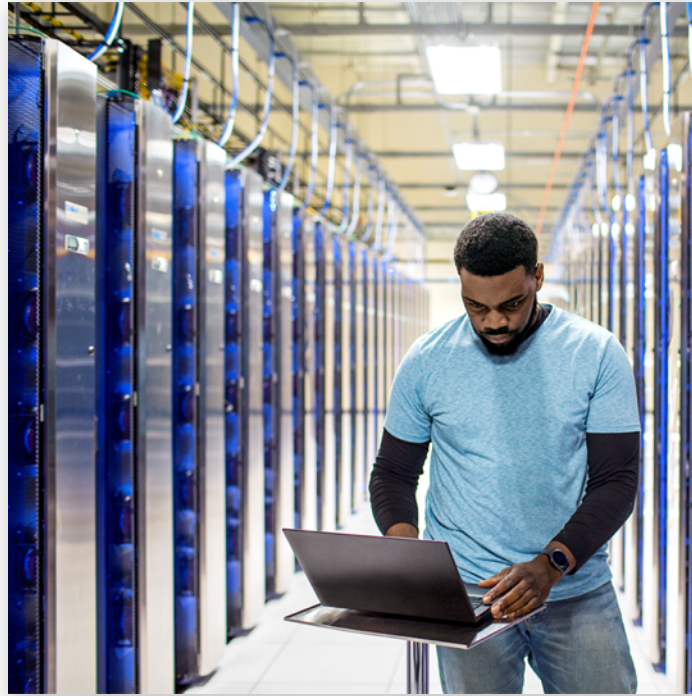


Solve previously unsolvable problems  
Robust and flexible decision making  
Learned policy  
Human in the loop

# Microsoft platform for autonomous systems



Scale Human Expertise



Trustworthy Autonomy



Real World Scenarios

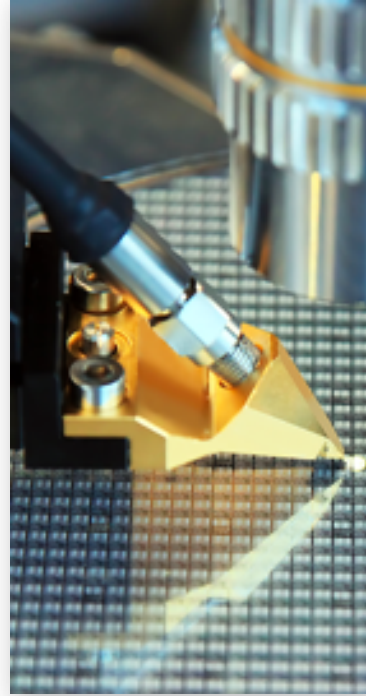
# Autonomous systems scenarios



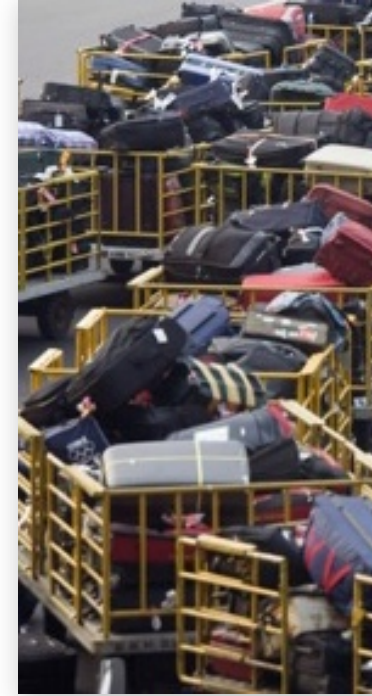
Motion control



Smart buildings



Machine calibration



Process control



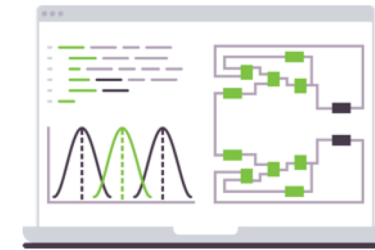
Industrial robotics

# BUILDING INTELLIGENCE FOR AUTONOMOUS SYSTEMS DEMANDS A FUNDAMENTALLY DIFFERENT APPROACH

1. A technique that can combine human and machine intelligence



2. Simulation integration and scalability for training



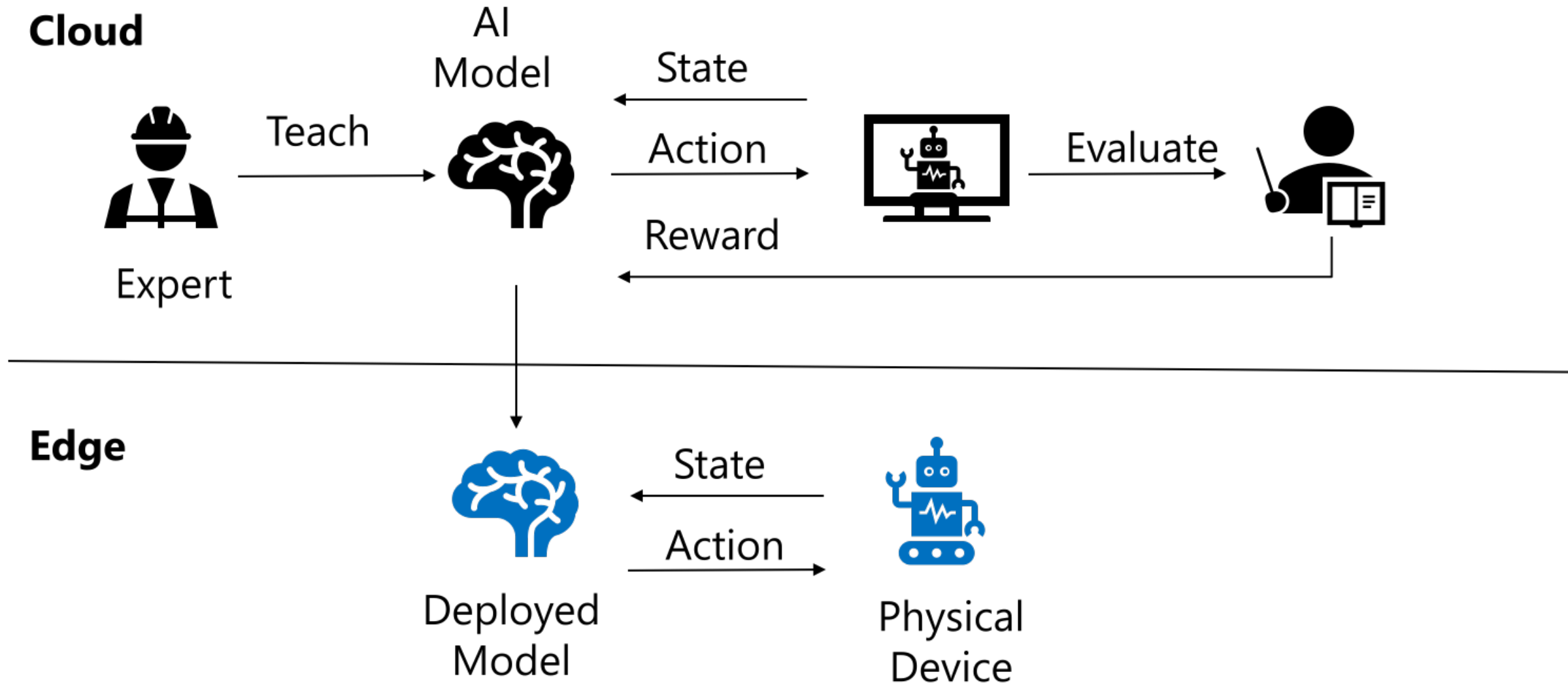
3. Automated generation and management of neural networks and DRL algorithms



4. A runtime to deploy and scale your models in the real world



# Autonomous Systems toolchain



# Reward functions

## Cobra effect

You get what you incentivize,  
not what you intend.



# Case study

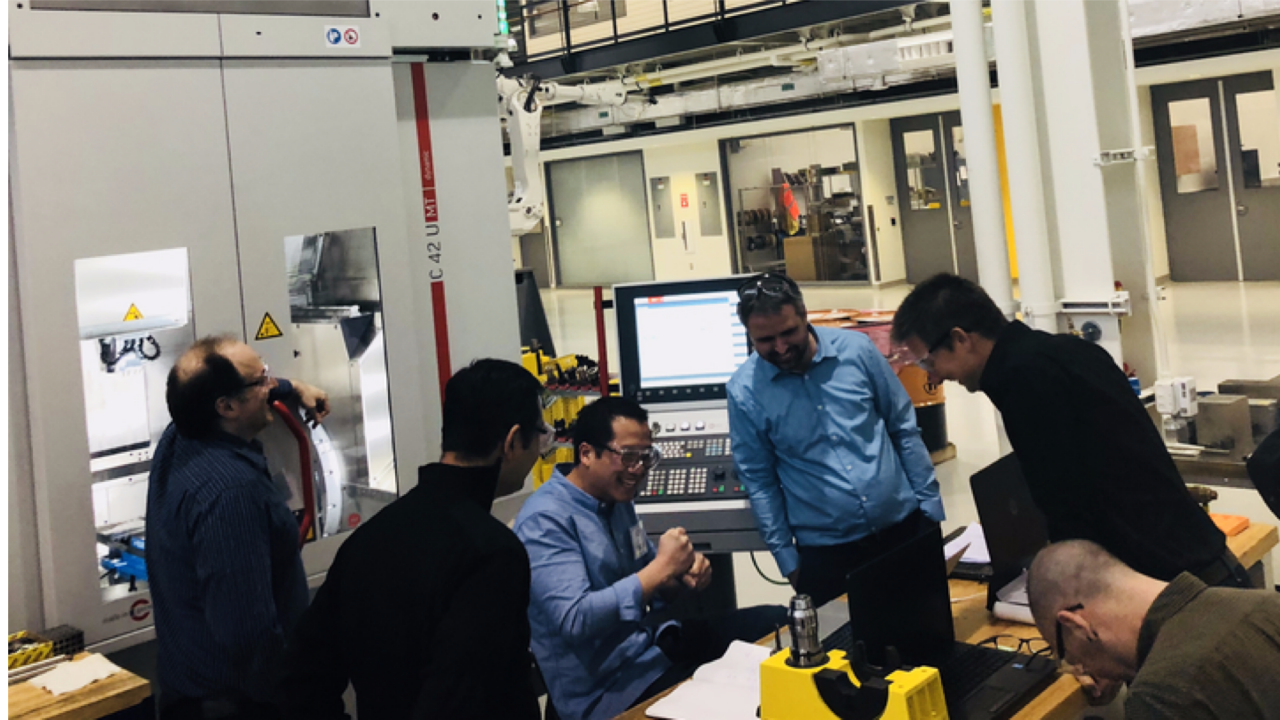
## CNC machine calibration

### Business problem

CNC machines cut metal with spinning tools. Friction reduces precision and periodically demands recalibration. An expert operator must travel to calibrate the machine, repeatedly turn the knobs and take measurements until the machine regains precision.

### Objective

Build autonomous system to calibrate machine to offset friction error to within 2 microns.



### Results

A manual process requiring ~~Sensors~~ human operators—averaging 20–25 iterative steps over 2 hours—was fully automated to an average of 4–5 iterative steps over 13 seconds



Engine friction



Screw friction



Slide friction

Above was achieved at 2 micron precision—the system could achieve superhuman precision (1 micron) in <10 iterative steps

All built by a non-RL expert (mechanical engineer SME)



# Case study

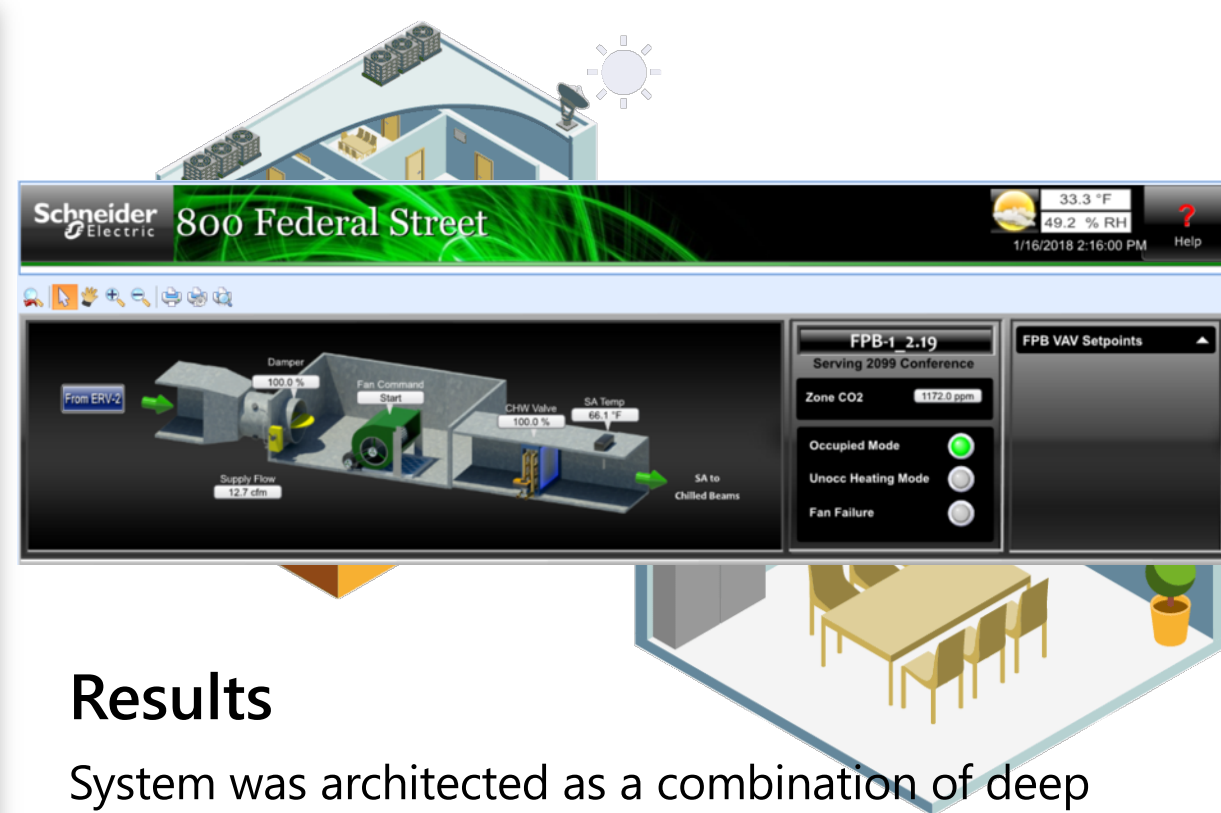
## Smart buildings/homes

### Business problem

HVAC systems comprise most of commercial energy consumption. Traditional controls struggle to save energy keep CO2 levels safe while keeping occupants comfortable.

### Objective

Train autonomous system to reduce energy consumption while maintaining occupant comfort and CO2 safety in a conference room.

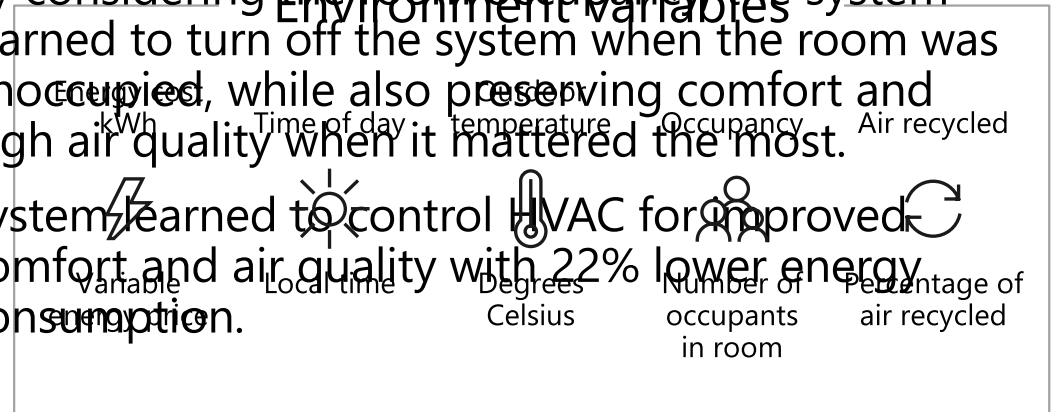


### Results

System was architected as a combination of deep neural networks and classical control systems.

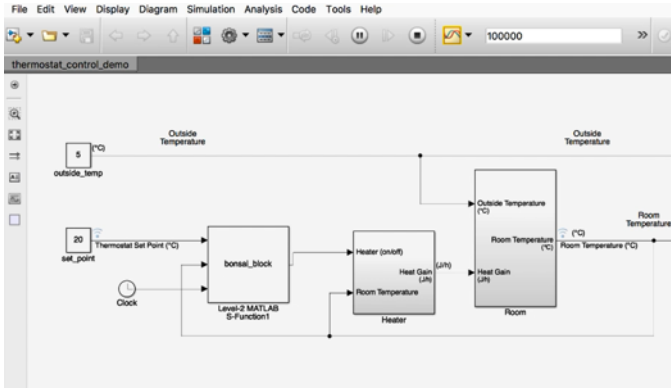
By considering the room occupancy, the system learned to turn off the system when the room was unoccupied, while also preserving comfort and high air quality when it mattered the most.

System learned to control HVAC for improved comfort and air quality with 22% lower energy consumption.



# Industrial Simulations

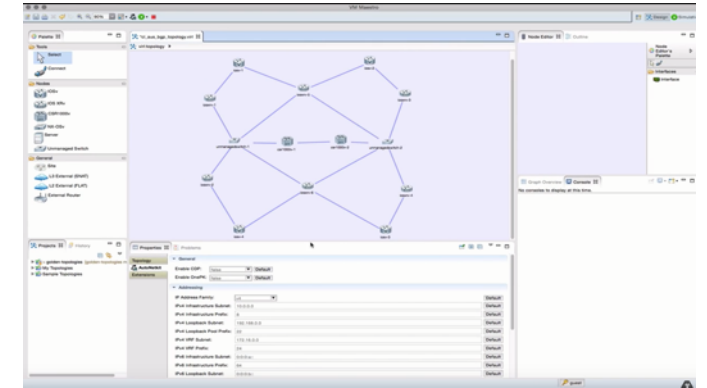
Available across a broad range of verticals and systems



Mechanical & electrical engineering



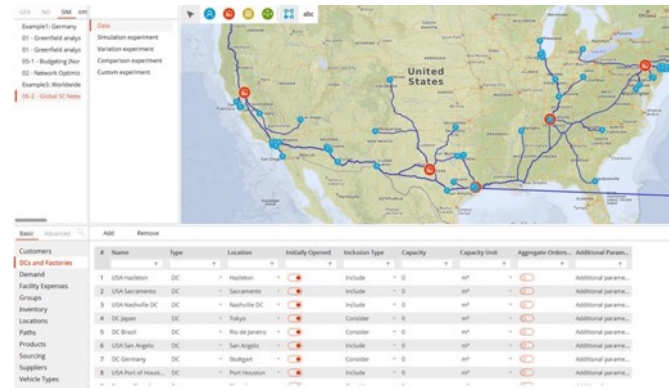
Autonomous vehicles



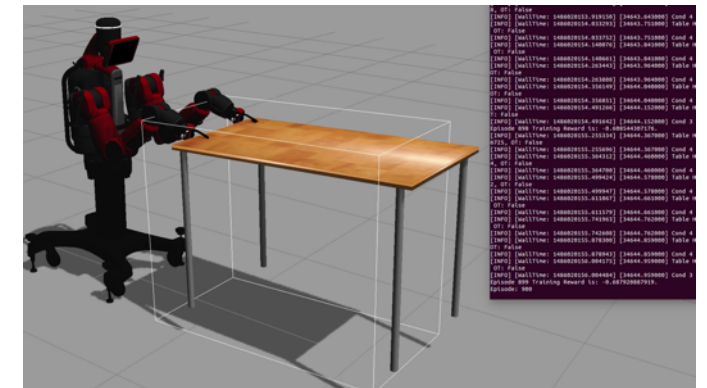
Security & networking



Discrete event simulations



Transportation & logistics



Robotics

**DEMO**



Microsoft and The MathWorks are collaborating to deliver a best in class User Experience for building Autonomous Systems

- Easy to use Simulink Toolbox provided by Microsoft
- Microsoft service deliver highly scalable training environment for Simulink and MATLAB based models

Preview available at: <https://aka.ms/as/preview>