Context Aware Safety Monitoring in Medical Cyber-Physical Systems

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Electrical and Computer Engineering

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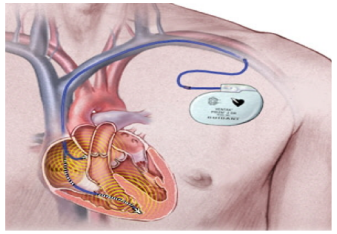
UNIVERSITY of VIRGINIA

Medical Cyber-Physical Systems

Pacemakers



Insulin Pumps

Wearable Monitors

Patient Monitors

Infusion Pumps

Defibrillators

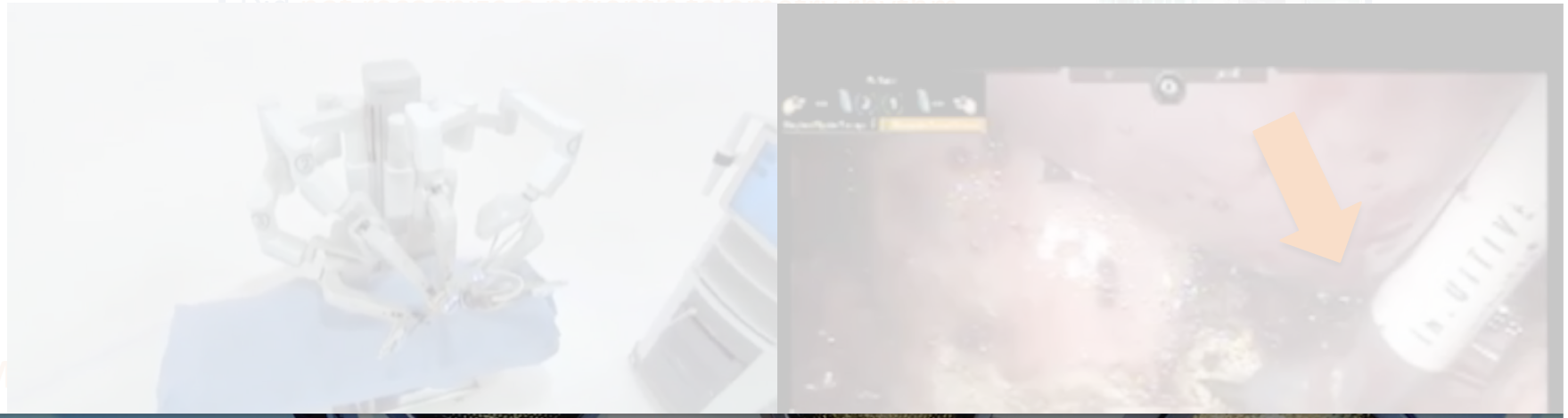
Surgical Robots

Imaging Systems

Linear Accelerators

Catastrophic Events

GE Healthcare - Telemetry Monitoring Systems

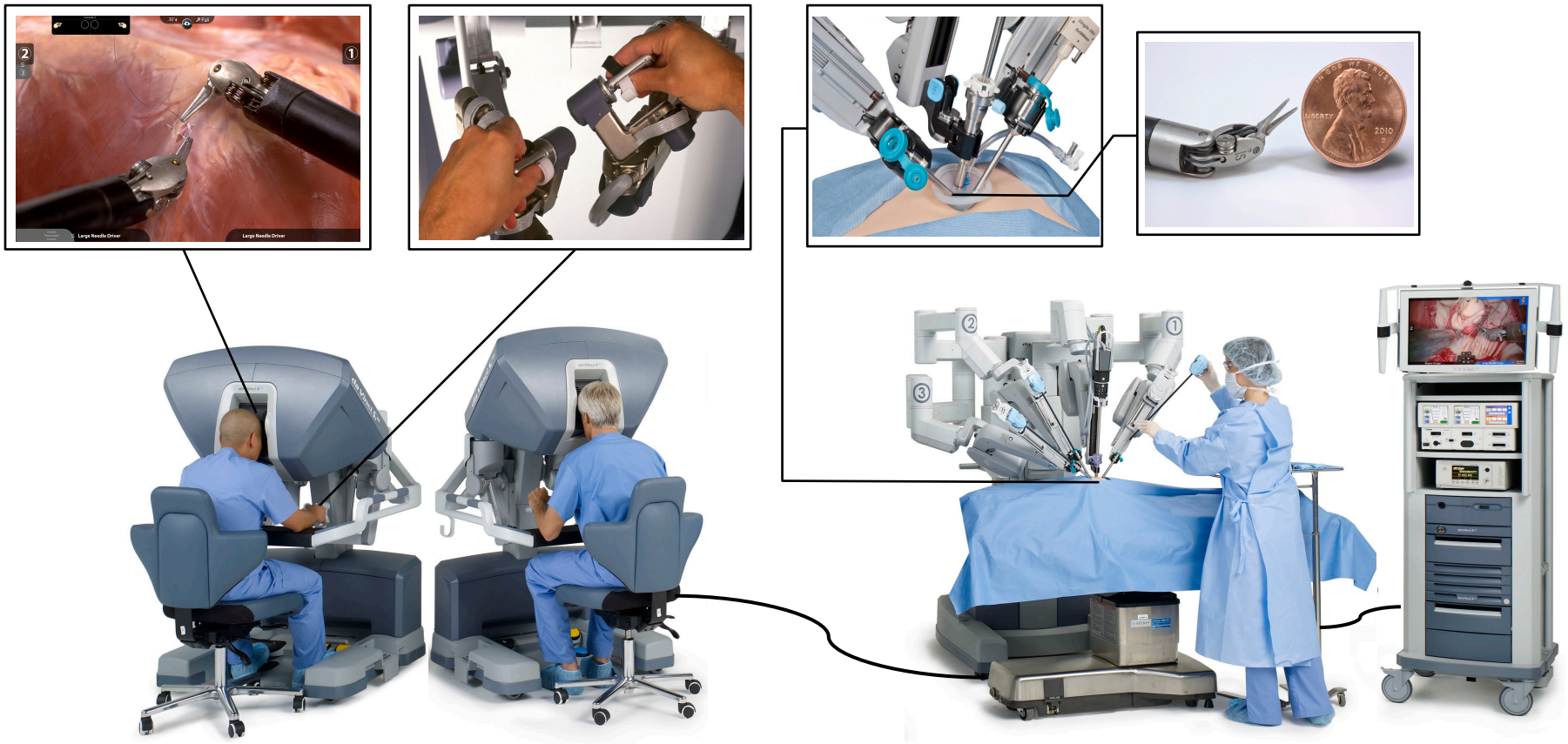


Hidden FDA Reports Detail Harm Caused By Scores Of Medical Devices

The Food and Drug Administration has let medical device companies file reports of injuries and malfunctions outside a widely scrutinized public database, which leave doctors and medical sleuths in the dark.

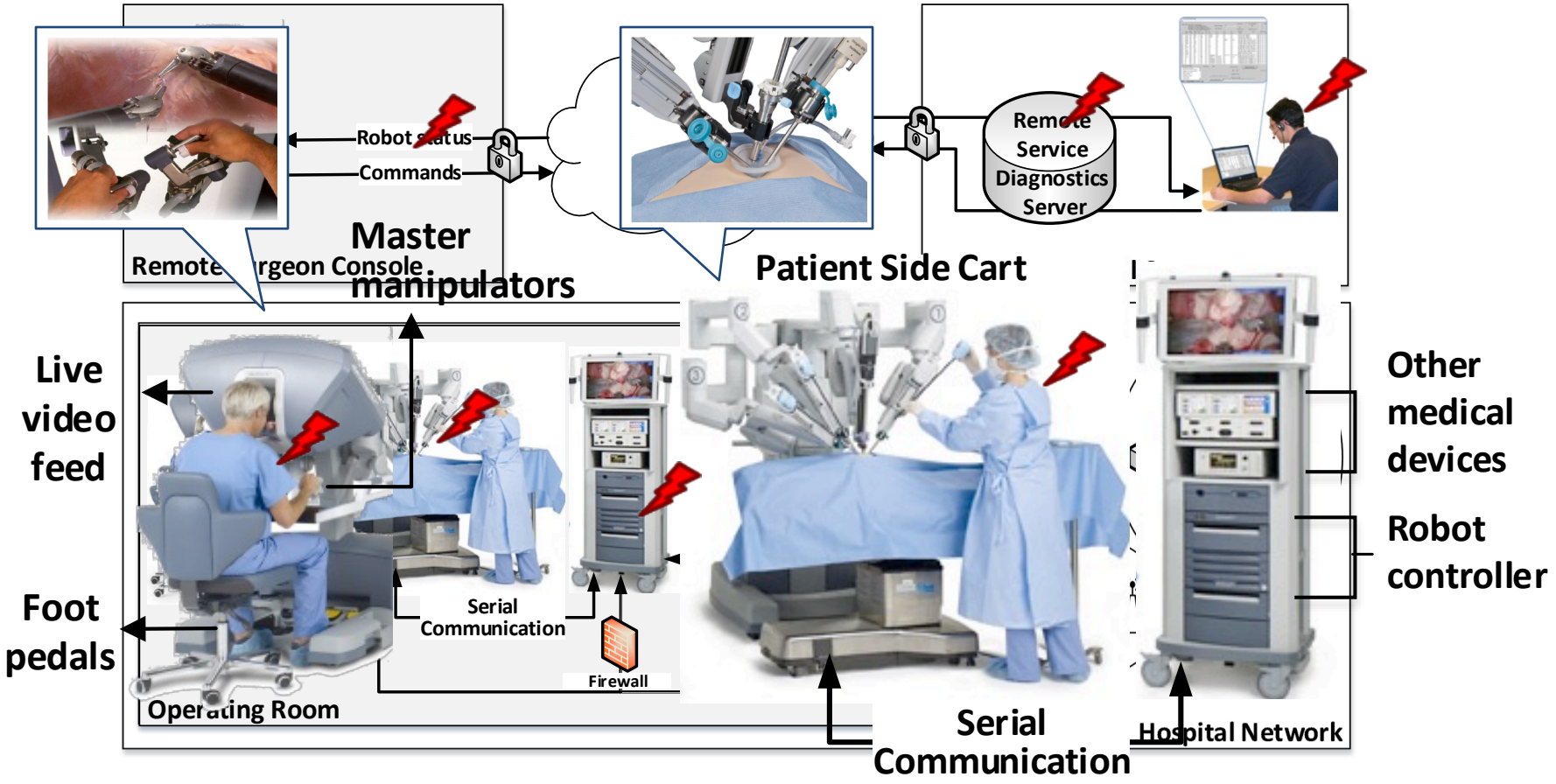
By **Christina Jewett** • Photos by **Heidi de Marco** • MARCH 7, 2019

Human-Cyber-Physical Systems



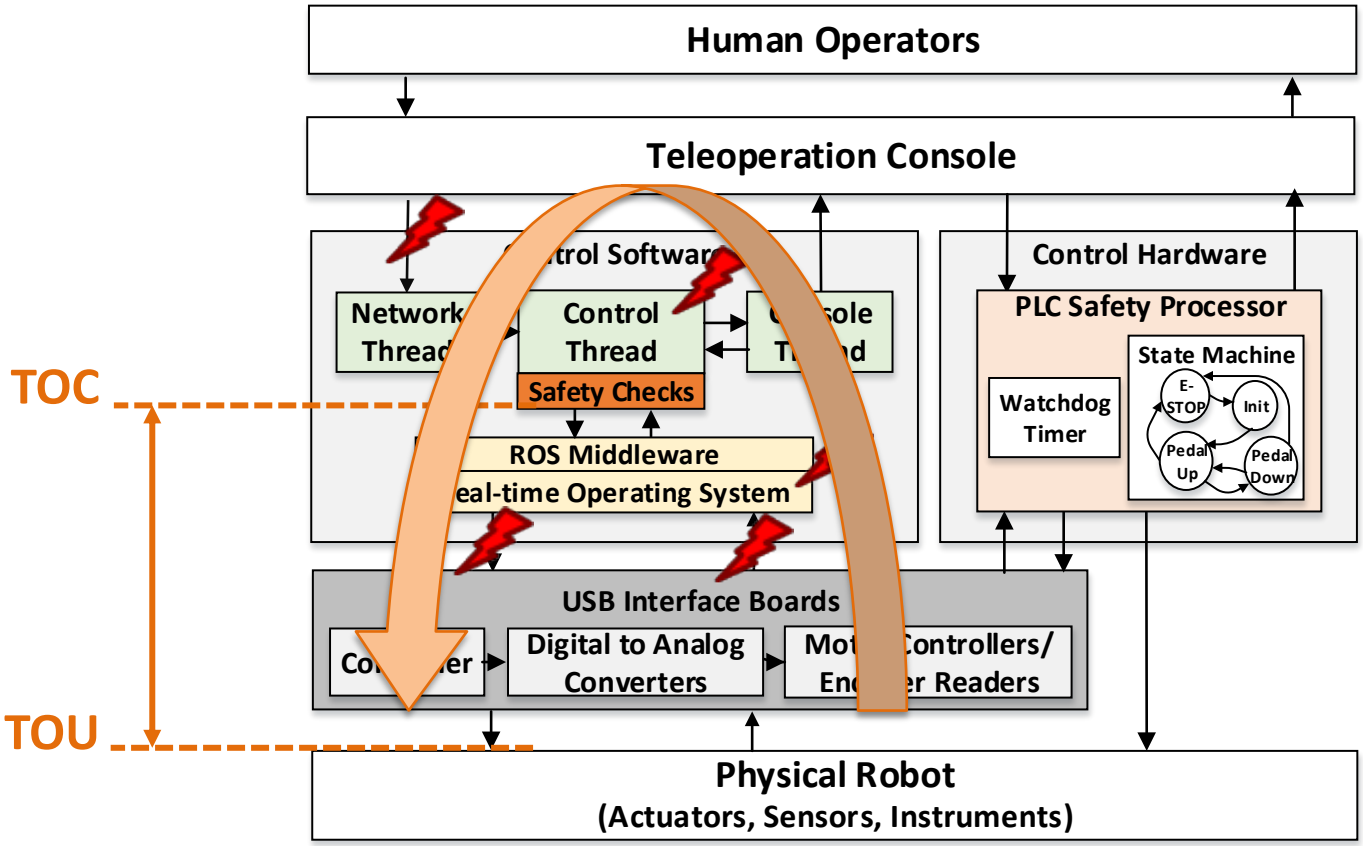
da Vinci Surgical System, Intuitive Surgical, Inc.

Human-Cyber-Physical Systems



da Vinci Surgical System, Intuitive Surgical, Inc.

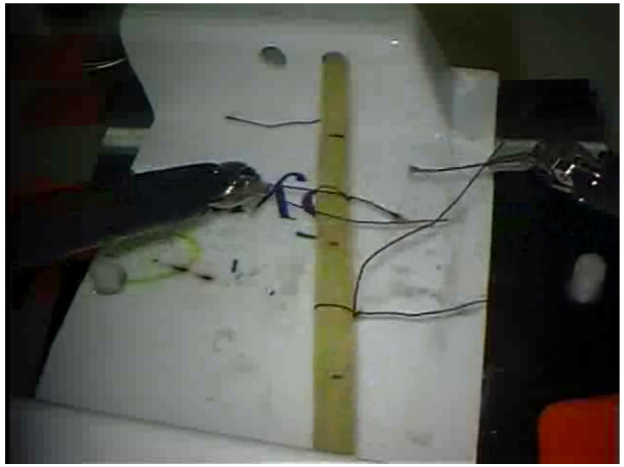
Vulnerabilities of Control System



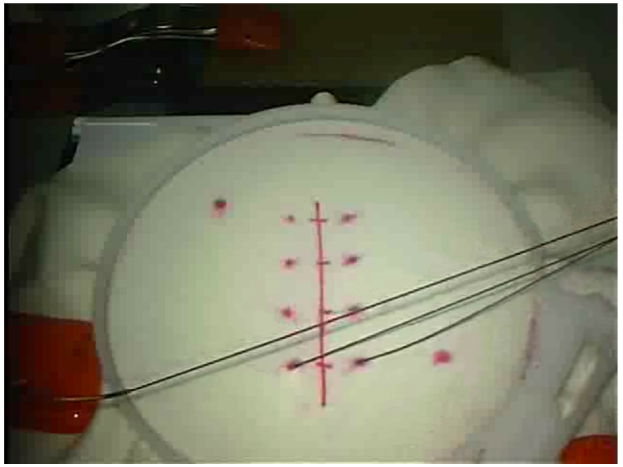
Each control cycle
1 milliseconds

RAVEN II, Applied Dexterity, University of Washington

Loosely Closed-loop System: No haptics, limited vision feedback



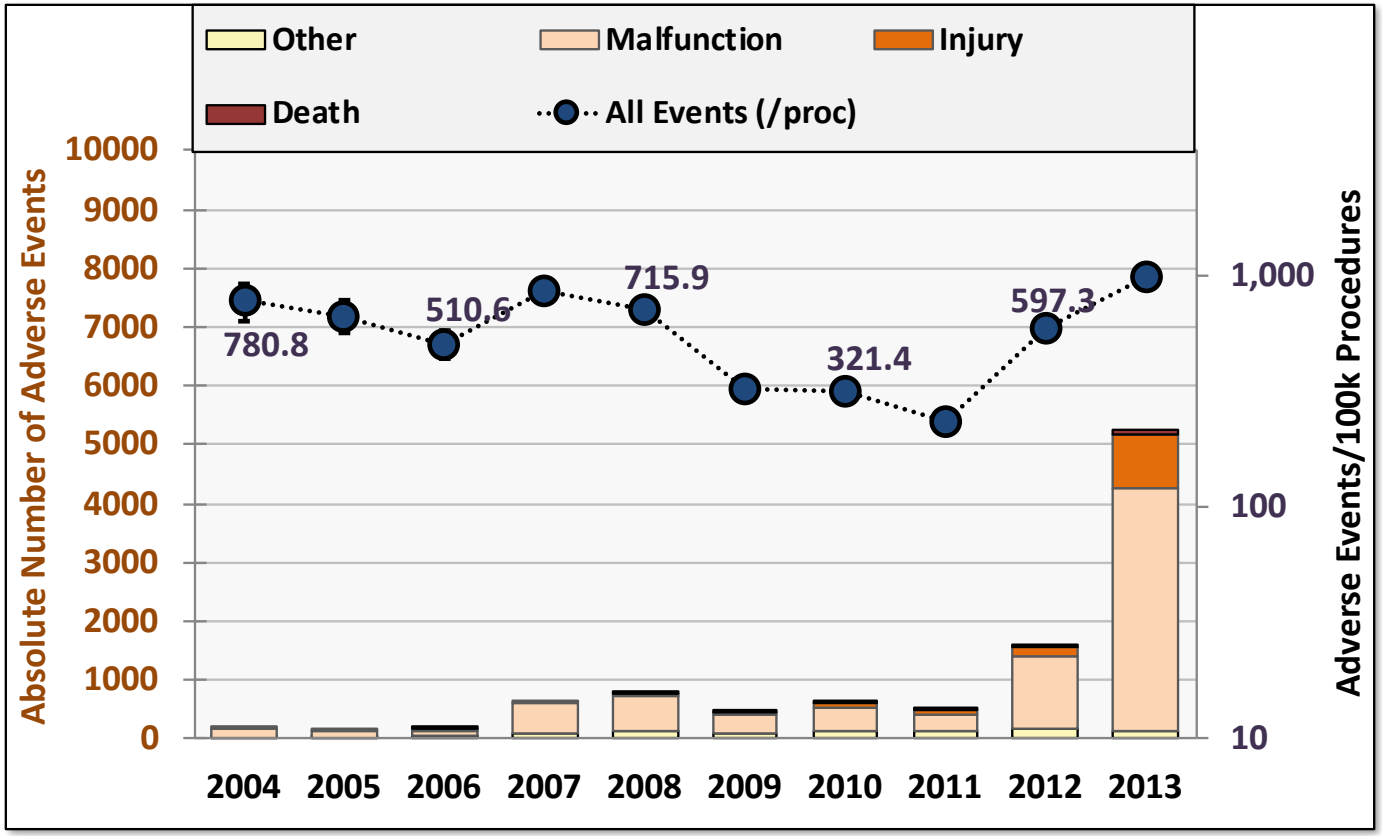
High Force



Graspers out of sight

Image Source: Gao, Yixin, et al., MICCAI Workshop: M2CAI. Vol. 3. 2014

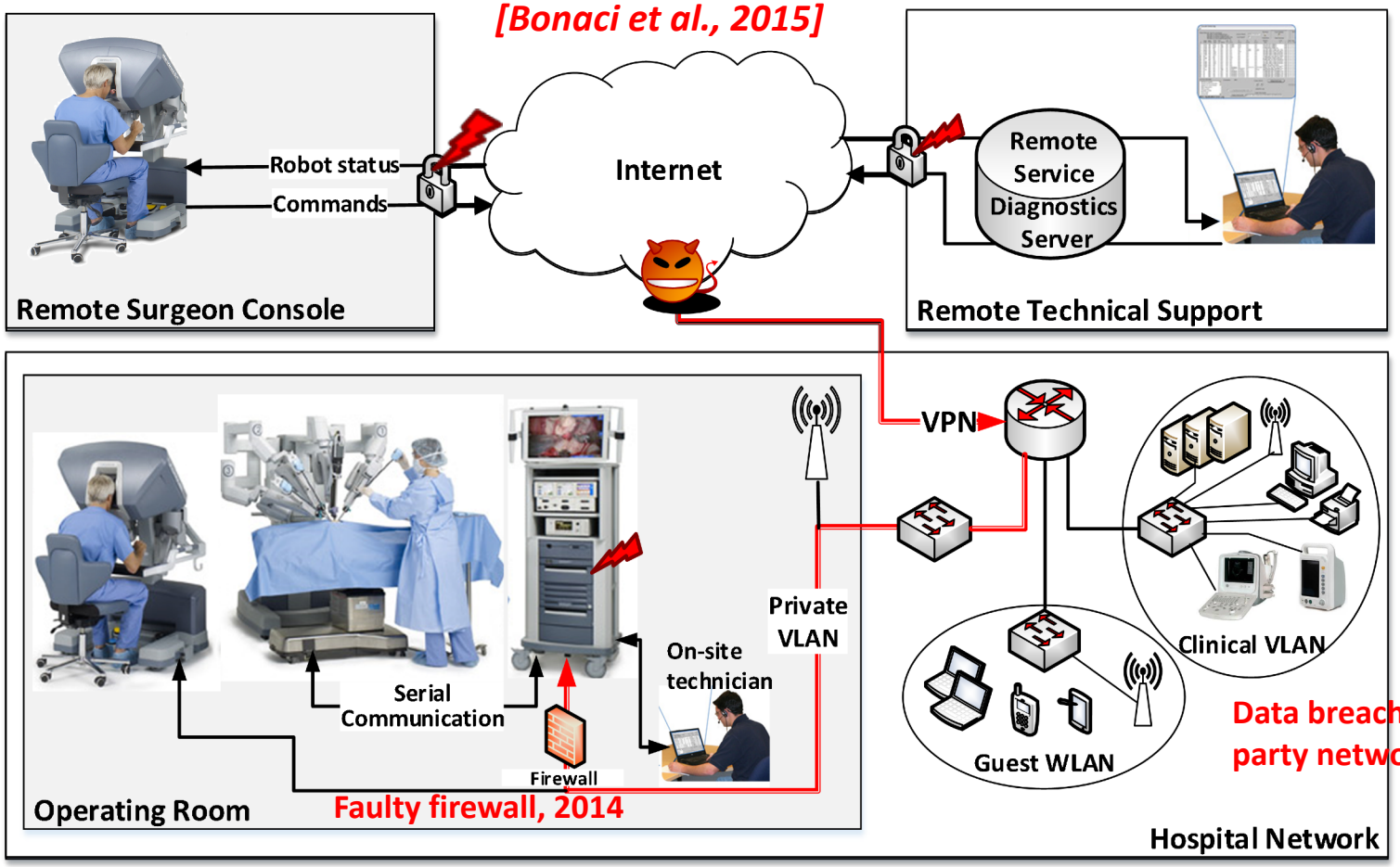
Once in every 100 procedures, an unexpected adverse event is likely to happen.



Alemzadeh et al., PLOS One 2016.
Featured in Wall Street Journal, MIT Technology Review, BBC, NBC News, Gizmodo, among others.

Malicious Attacks

DOS and MITM Attacks
[Bonaci et al., 2015]



Email phishing, 2014

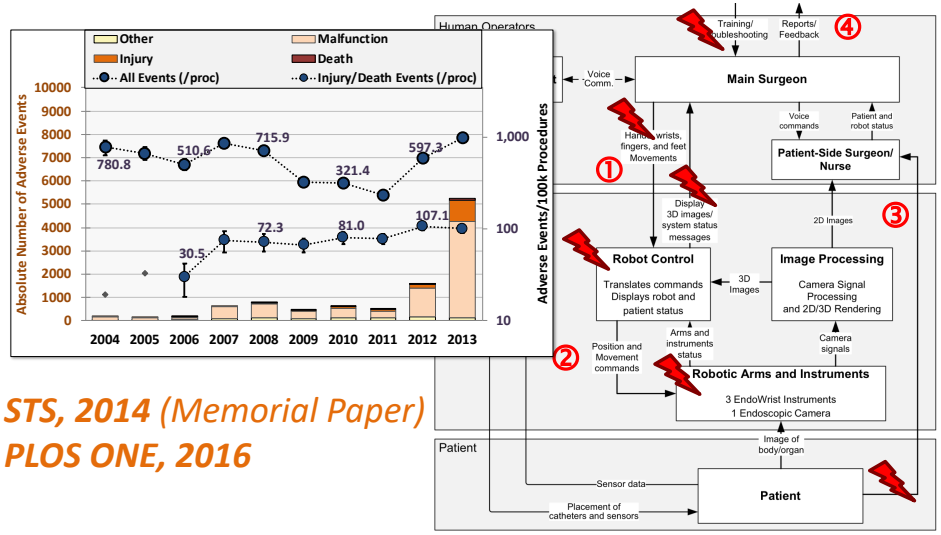
Default passwords, 2014

Medical devices, 2015

Data breach through 3rd party networks, 2015

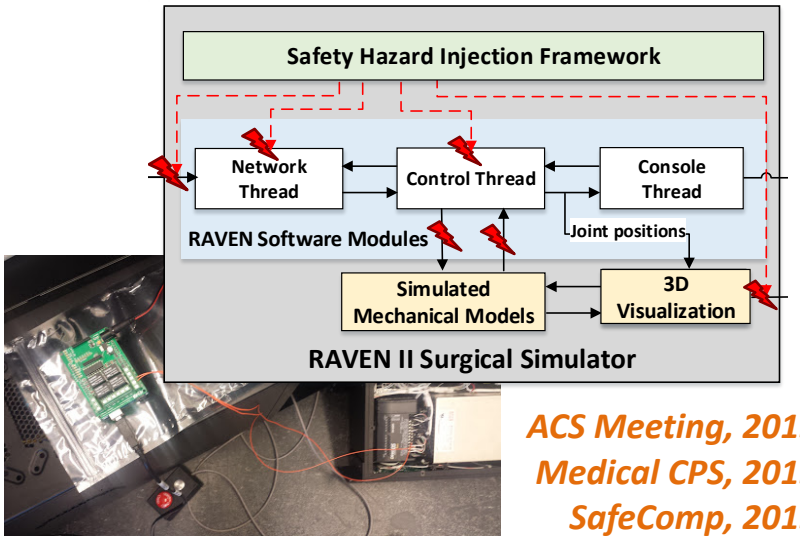
Attacks on robot control system

Analysis of Adverse Events



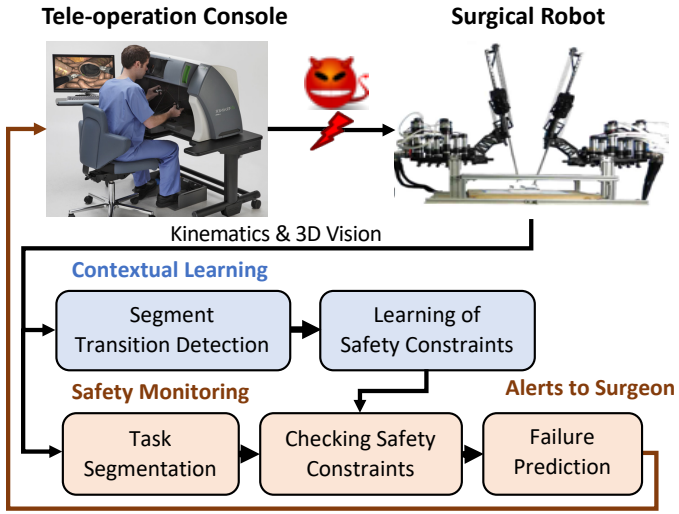
STS, 2014 (Memorial Paper)
PLOS ONE, 2016

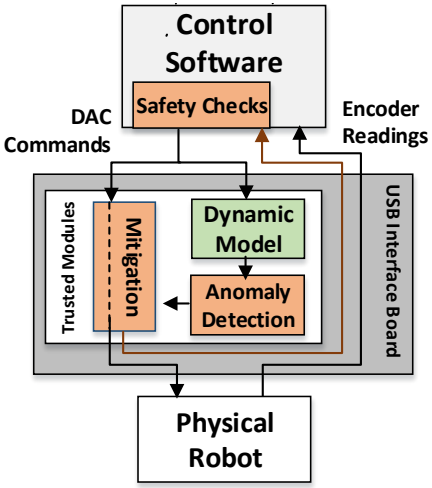
System Resilience Assessment



ACS Meeting, 2015
Medical CPS, 2015
SafeComp, 2015
IROS, 2016
PRDC, 2018

Real-time Safety Monitoring

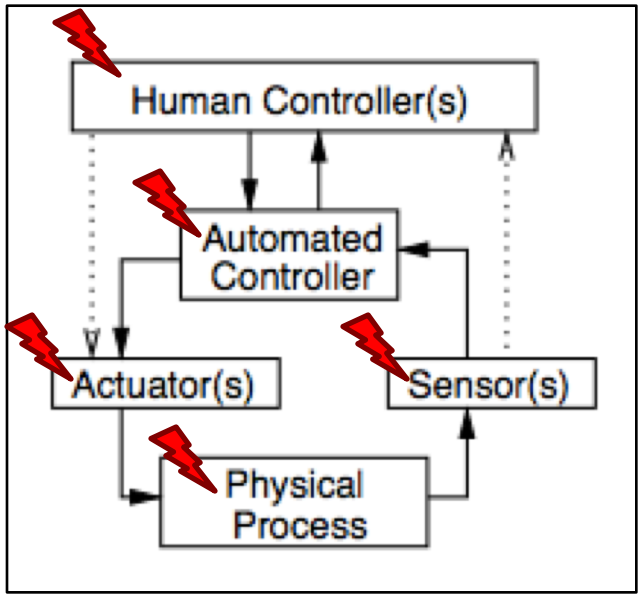




DSN, 2016
HotSoS, 2016
ISMR, 2019

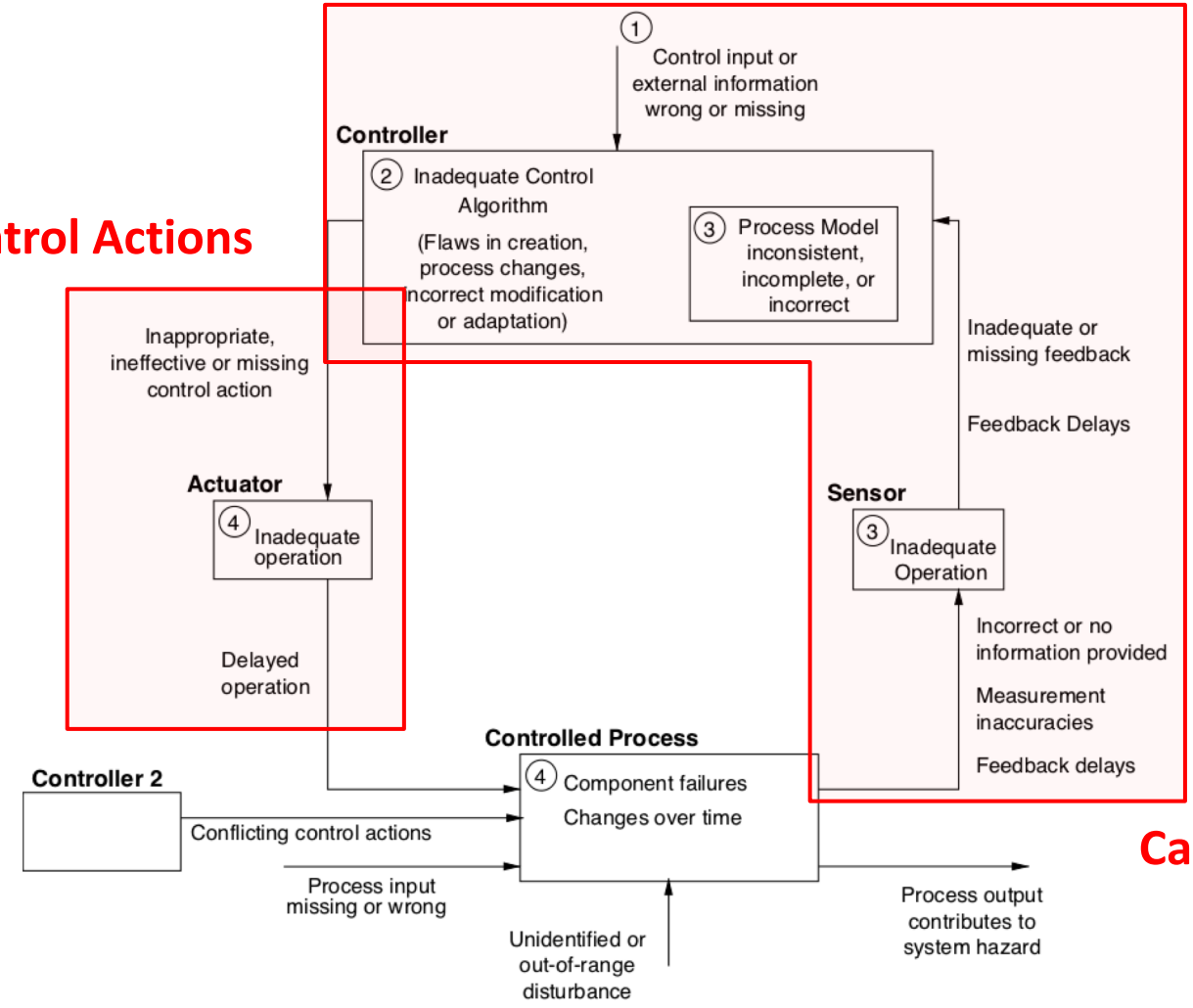
Real Time Safety Monitoring

- **Control-theoretic safety modeling and analysis**
  Violation of safety constraints in the control loops



Unexpected failures or intentional malicious actions leading to unsafe control

Unsafe Control Actions



Unsafe Control Actions

Causal Factors

Nancy Leveson, *Engineering a safer world: Systems thinking applied to safety*. MIT press, 2011.

Unsafe System Context

The set of system conditions under which the control actions could possibly be unsafe and lead to hazards.

i) a required control action was *not performed*

ii) a control action was performed *in a wrong state*

iii) a control action was performed *at an incorrect time*,

iv) a control action was performed *for an incorrect duration*,

v) a control action was provided, but *not followed by the controlled process*

Nancy Leveson, *Engineering a safer world: Systems thinking applied to safety*. MIT press, 2011.

Accidents and Safety Hazards

Accidents:

A-1. Patient expires during or after the procedure.

A-2. Patient is injured or experiences complications during/after the procedure.

A-3. Surgical system or instruments are damaged or lost.

Hazards:

H-1. Robot arms/instruments move:

- to unintended location (H1-1),
- with unintended velocity (H1-2),
- at unintended time (H1-3).

H-2. Robotic arms or instruments are subjected to collision/unintended stress.

H-3. Robotic system becomes unavailable or unresponsive during procedure.

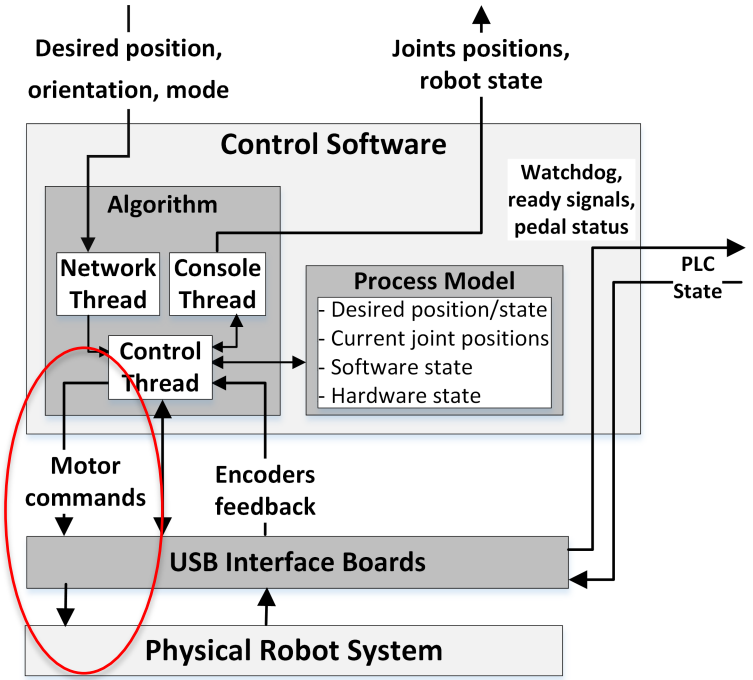
Unsafe System Context

ii) a control action was performed *in a wrong state*

A motor command is ***provided*** by control software when the ***user desired joint position is at a large distance from the current joint position***

Potential hazard: **H1-2**

Robot arms/instruments will move with an unintended velocity



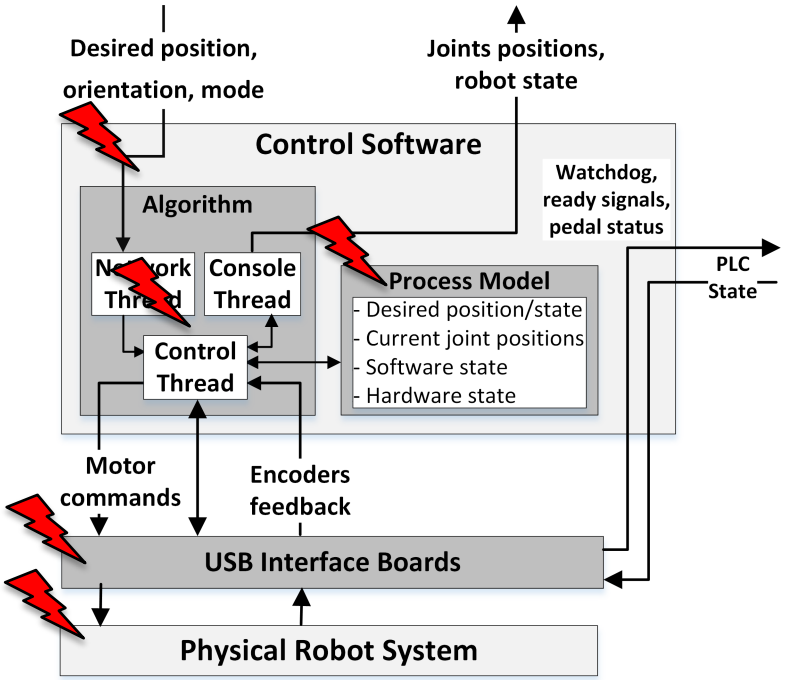
Unsafe Control Actions

ii) a control action was performed *in a wrong state*

A motor command is ***provided*** by control software when the ***user desired joint position is at a large distance from the current joint position***

Potential causes:

- Incorrect console inputs
- Faulty control algorithm
- Incorrect process model
- Faulty USB communication
- Physical system malfunction

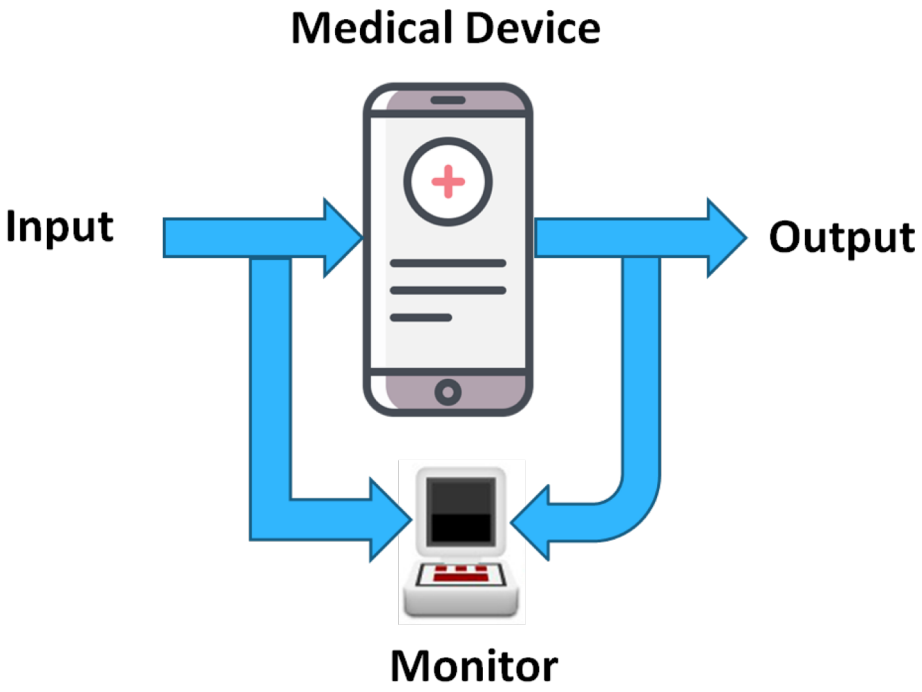


Real Time Safety Monitoring

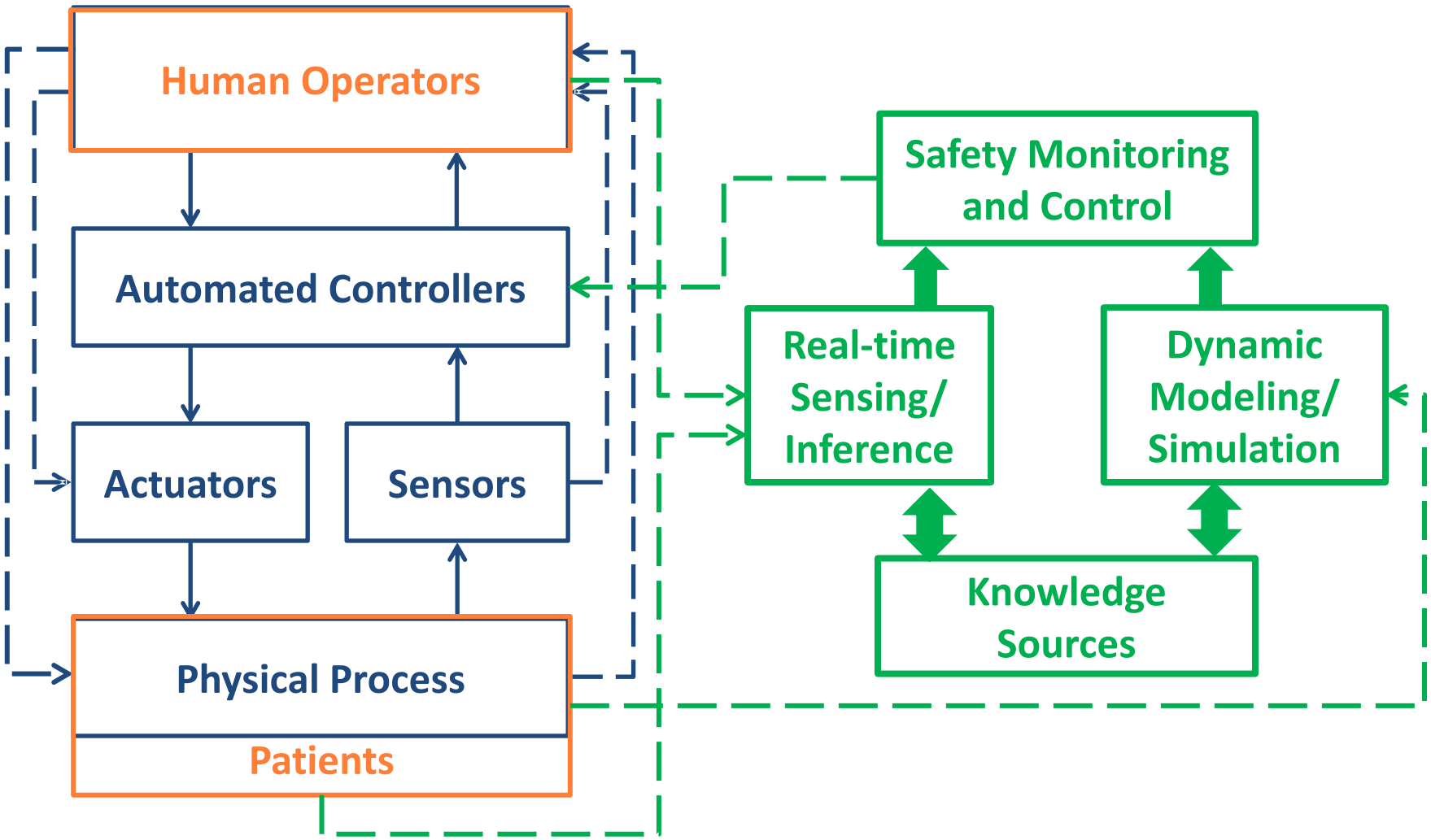
- **Preemptive Detection of Safety Hazards**
  Unsafe system context leading to unsafe control actions

Fault Propagation Timeline

- Fault Occurred
- Fault Activated
- Fault Manifested
- **Unsafe Control Action Detected**
- Unsafe Control Action Issued => Safety Hazard
- Accident



Context-Aware Safety Monitoring



RAVEN II Surgical Robot

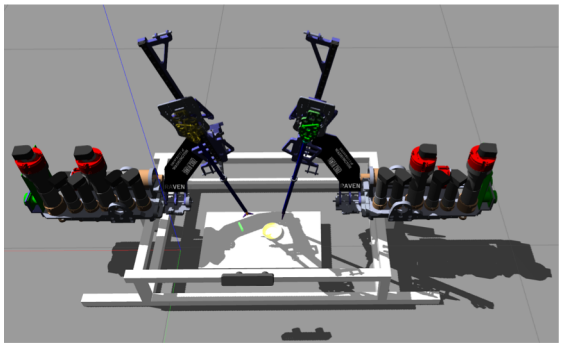
Tele-operation

- Perception
- High-level Task Scheduling
- Motion Planning

Autonomous Agent

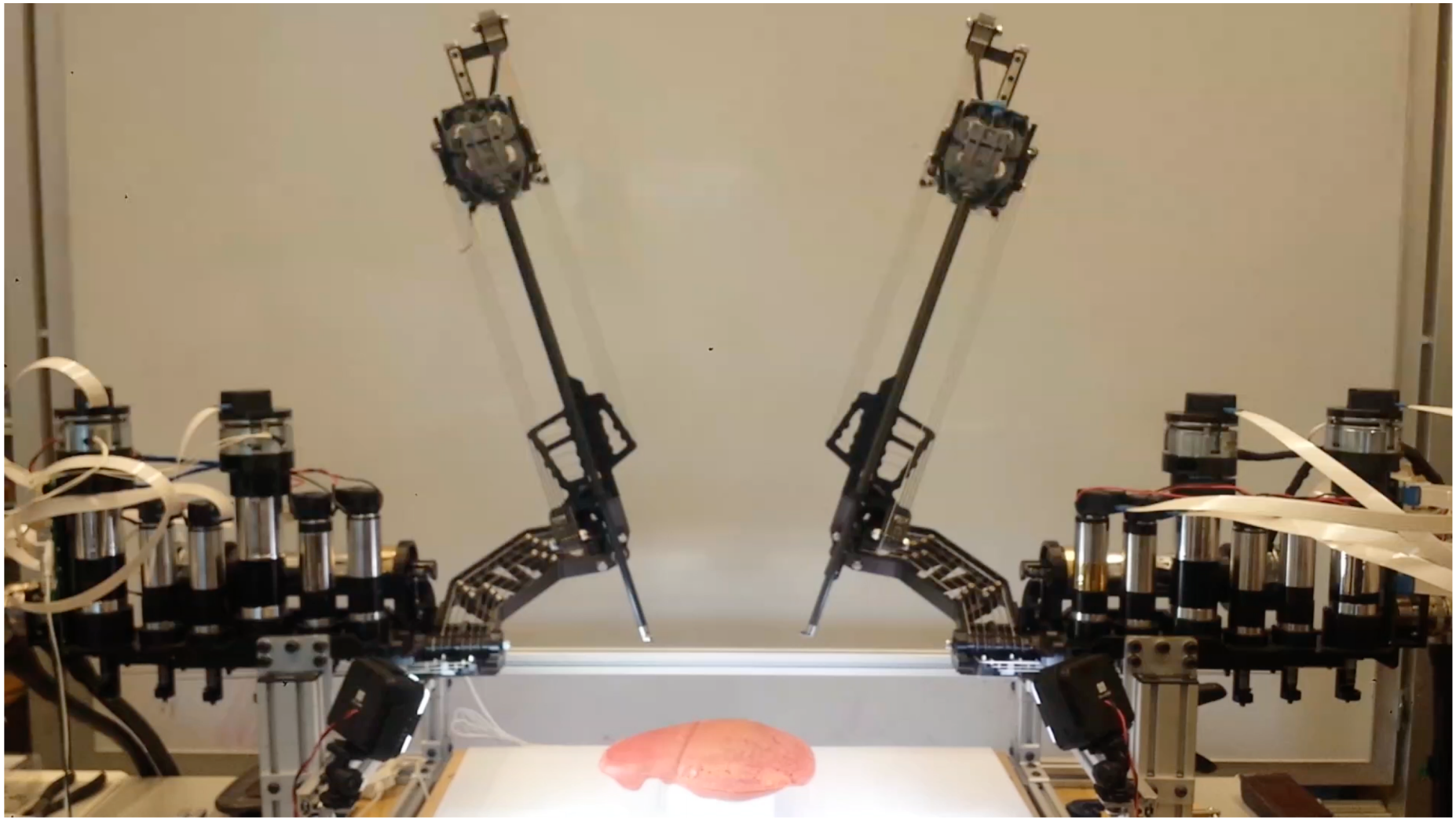
RAVEN Control Box
Software + Hardware

RAVEN Surgical Robot

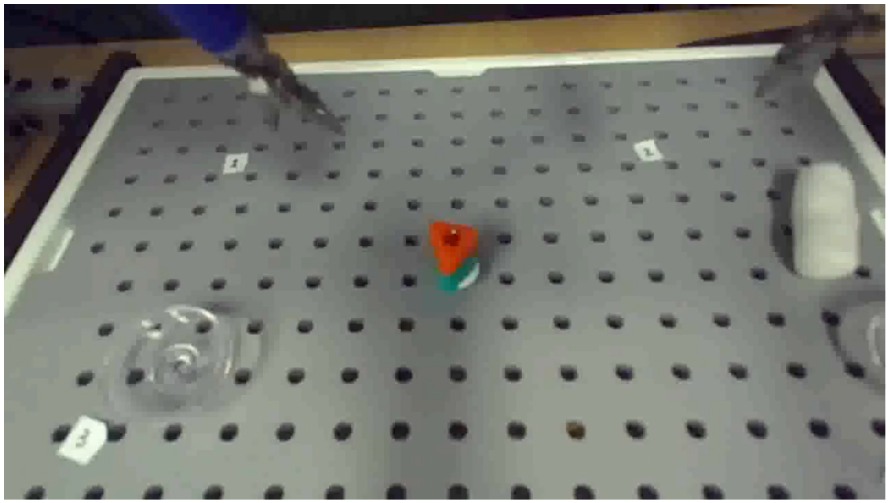


ROS Gazebo
Simulation Environment

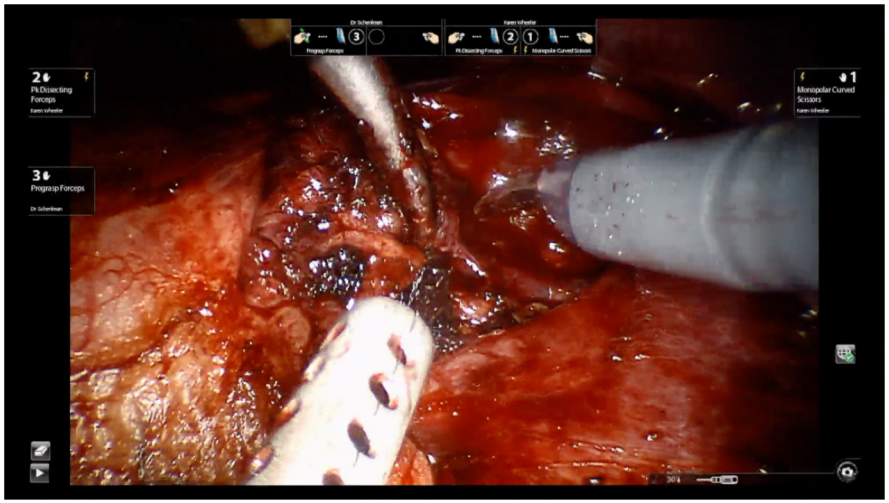
RAVEN II Surgical Robot

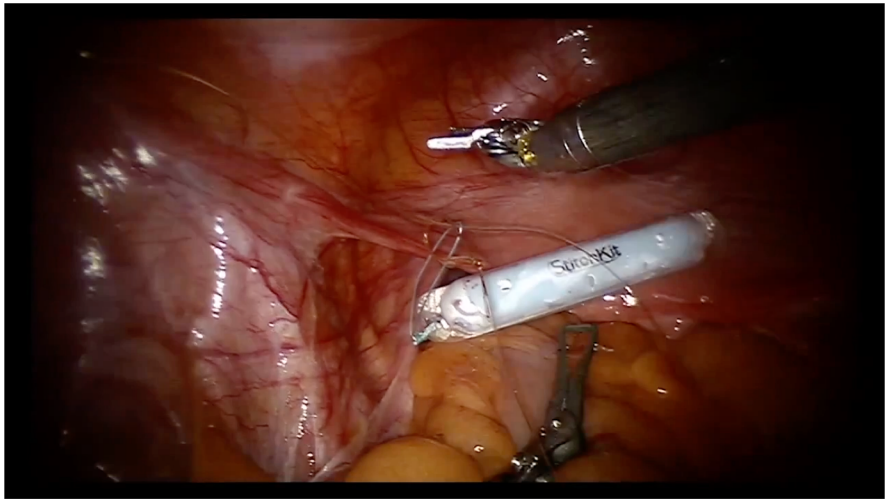


Example Surgical Task: Pick and Place



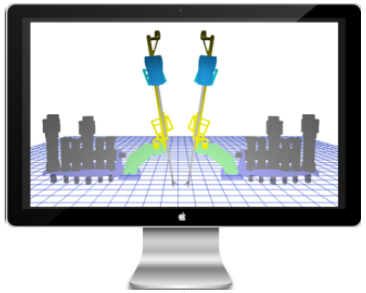
Dry Lab Simulation





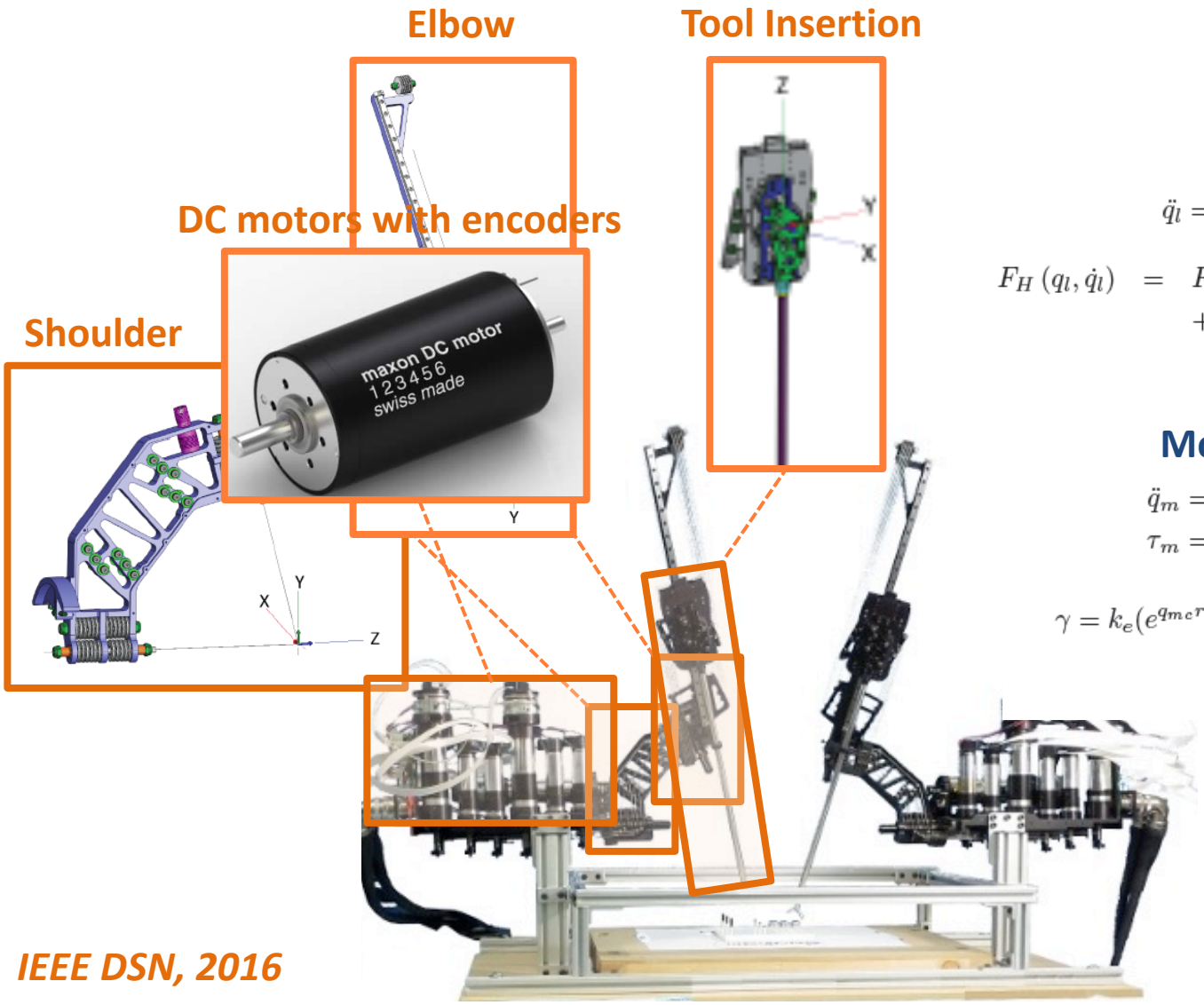
Actual Surgery

Simulator + Hazard Injection Engine



https://github.com/UVA-DSA/raven2_sim

Joint and Motor Dynamics



Joint Dynamics:

$$\ddot{q}_l = I_l^{-1}\left[\Gamma - F_H\left(q_l, \dot{q}_l\right)\right] \quad (1)$$

$$F_H\left(q_l, \dot{q}_l\right) = F_C\left(q_l, \dot{q}_l\right) + F_G + \mathrm{diag}\left(sign(\dot{q}_l)\right) F_{cl} + \mathrm{diag}\left(\dot{q}_l\right) F_{vl} + J^T F_{en} \quad (2)$$

Motor and Cable Dynamics:

$$\ddot{q}_m = (1/I_m)(\tau - \tau_m - \tau_{rn}) \quad (3)$$

$$\tau_m = \tau_{cm}\mathrm{sign}(\dot{q_m}) + \tau_{vm}\dot{q_m} \quad (4)$$

$$\tau_{rn} = r_{mc}\gamma/N \quad (5)$$

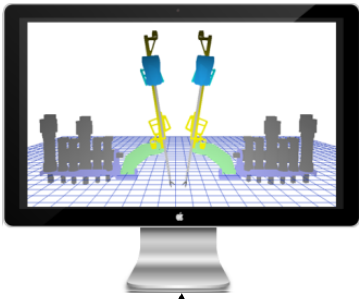
$$\gamma = k_e(e^{q_{mc}r_{mc} - q_l r_l} - e^{q_l r_l - q_{mc} r_{mc}}) + 2b_e(\dot{q_{mc}} r_{mc} - \dot{q_l} r_l) \quad (6)$$

$$\Gamma = r_l \gamma \quad (7)$$

IEEE DSN, 2016

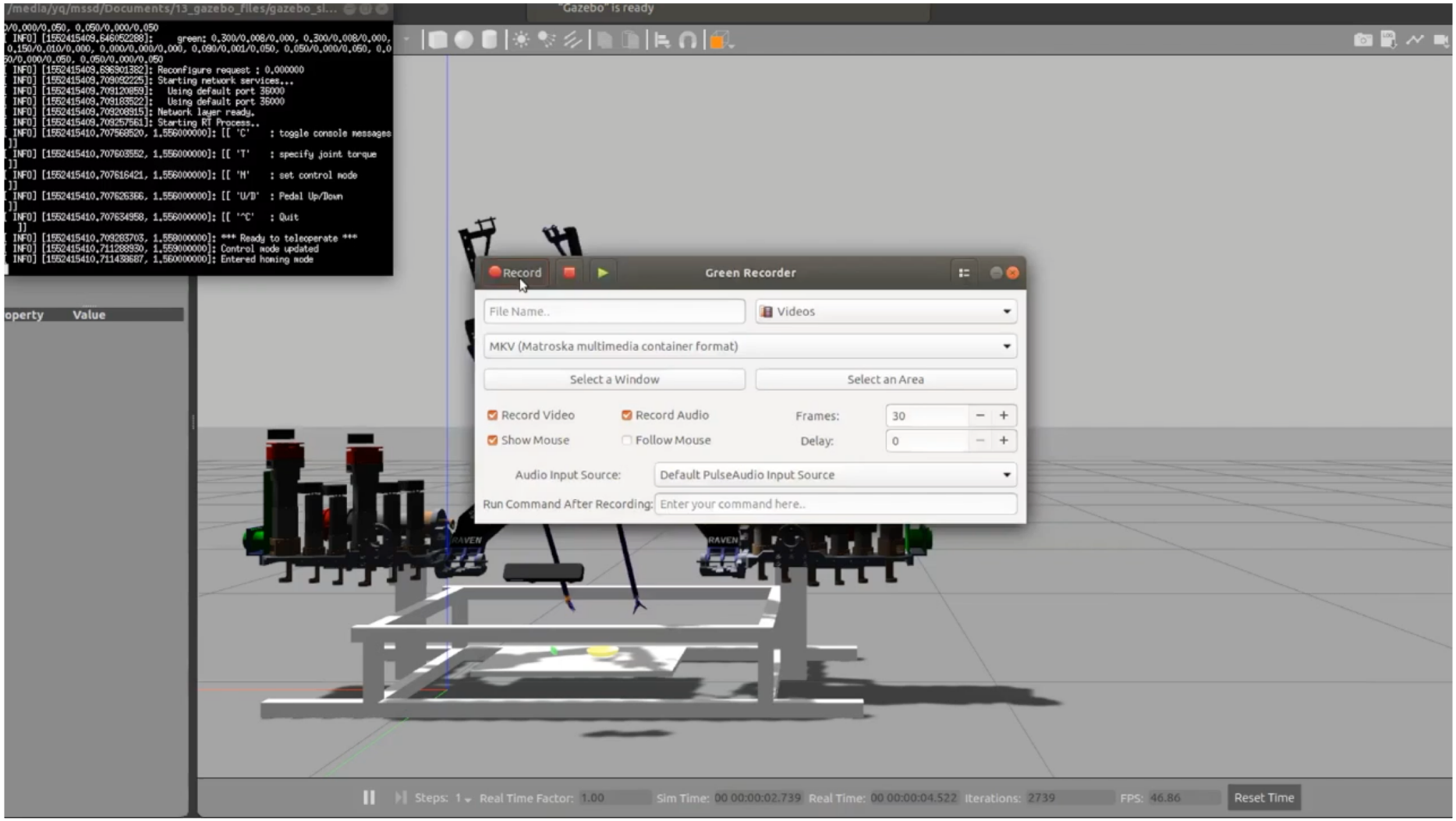
Haghighipanah, et al., 2015

Simulator + Hazard Injection Engine



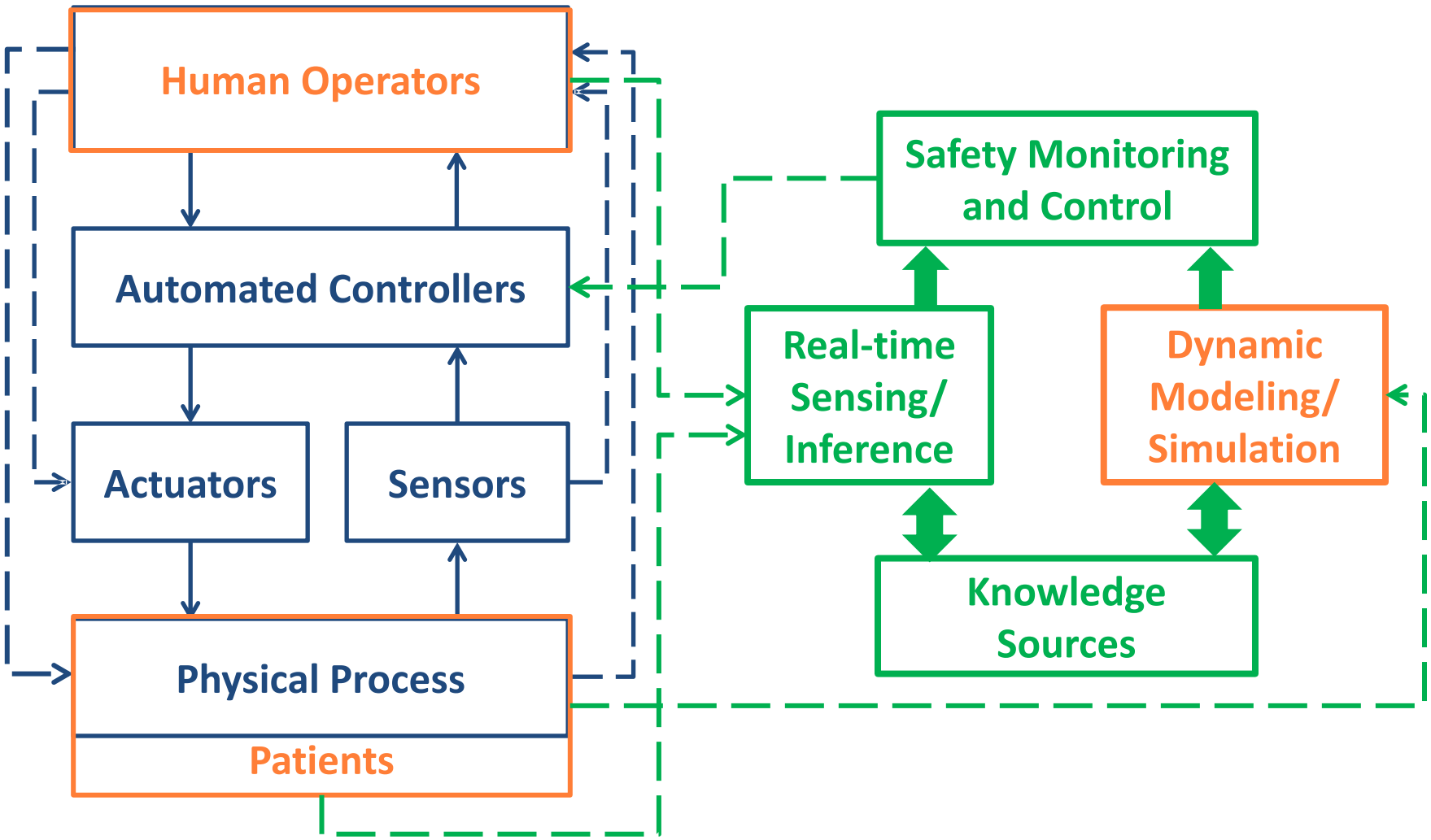
https://github.com/UVA-DSA/raven2_sim

Pick and Place Task in Gazebo Simulator

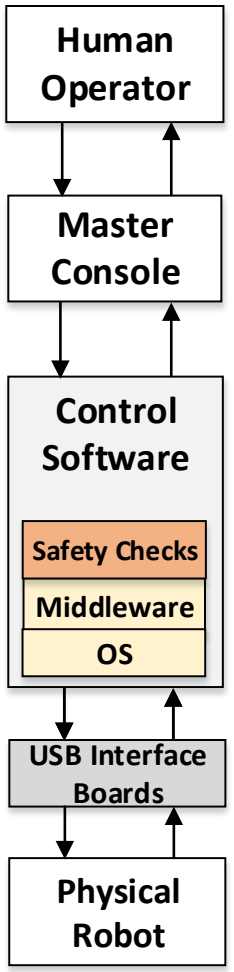


Failure Modes in Gazebo Simulator

Context-Aware Safety Monitoring

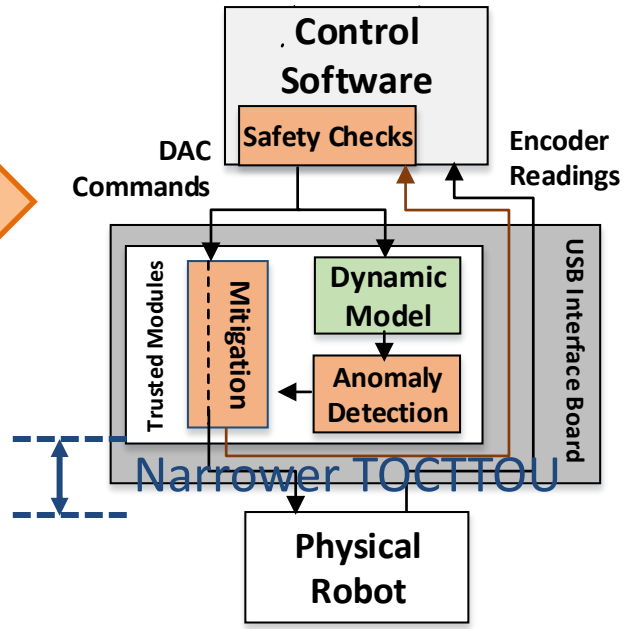


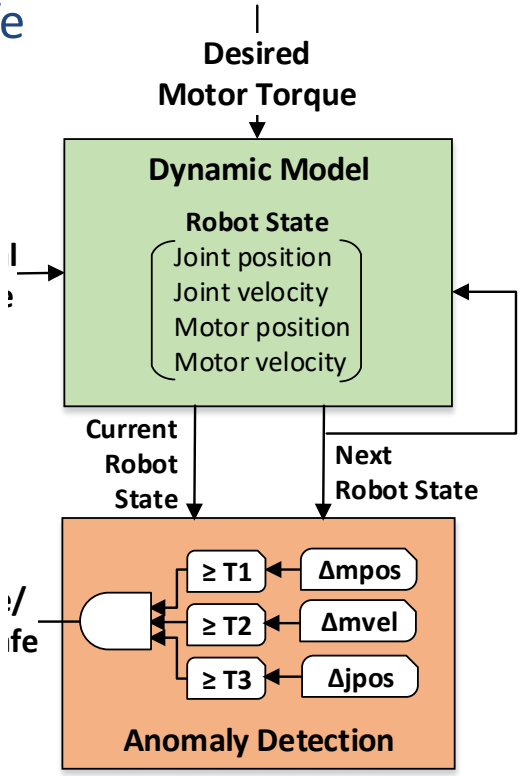
Dynamic Model Based Detection



Preemptive detection of safety hazards

- Real-time computation of joint/motor dynamics
- Estimation of next robot state
- Detect if distance is unsafe





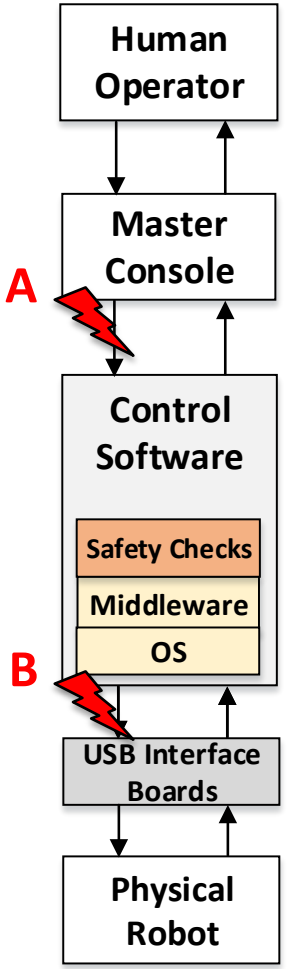
< 1 ms

Safety Hazard Detection Performance

Simulated attack scenarios
- Scenario A: 1,925 runs
- Scenario B: 1,361 runs

Different injected error values and attack activation periods

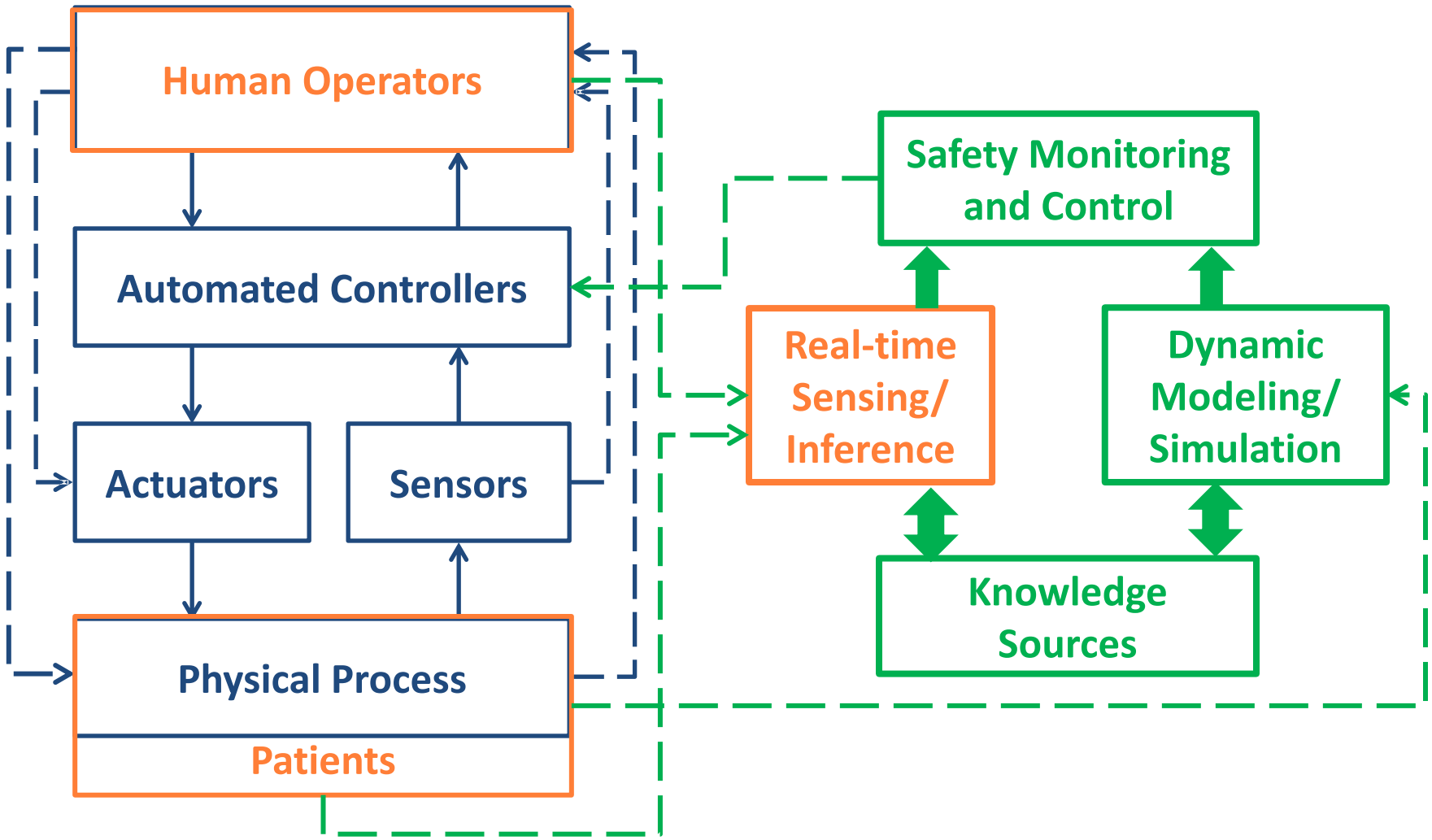
Attack Scenario	Technique	ACC (%)	TPR (%)	FPR (%)	F1 (%)
A (User inputs)	DM	88.0	89.8	12.4	74.8
	RAVEN	84.6	53.3	7.7	57.8
B (Torque commands)	DM	92.0	99.8	11.8	89.1
	RAVEN	90.7	81.0	4.6	85.1



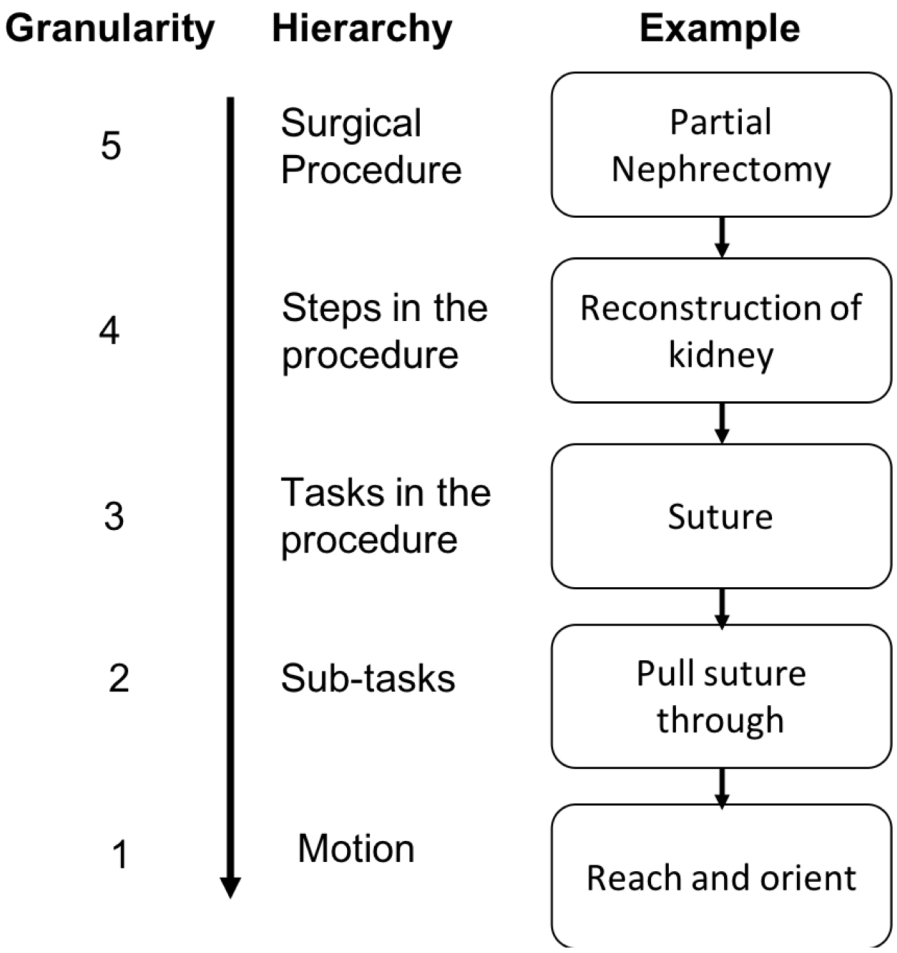
- DM detected **before** hazard manifested in physical layer
- RAVEN detected *at least* 1 cycle **after** safety hazard occurred

Alemzadeh et al., DSN 2016.

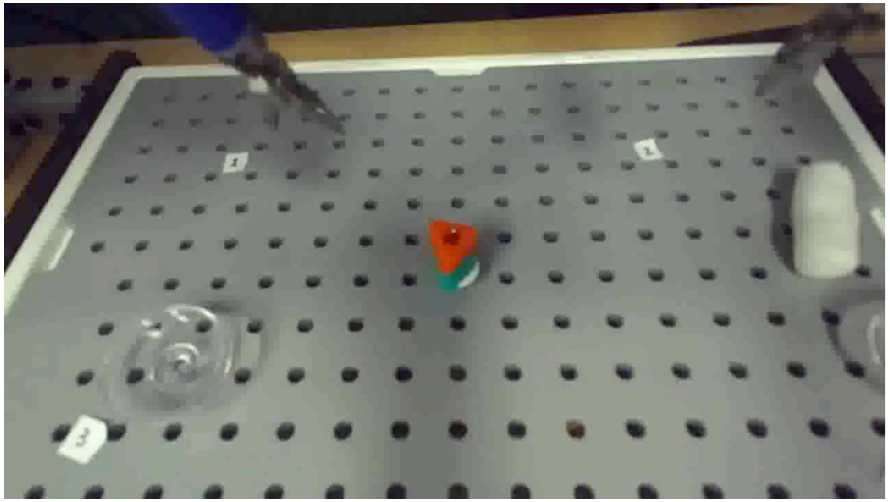
Context-Aware Safety Monitoring



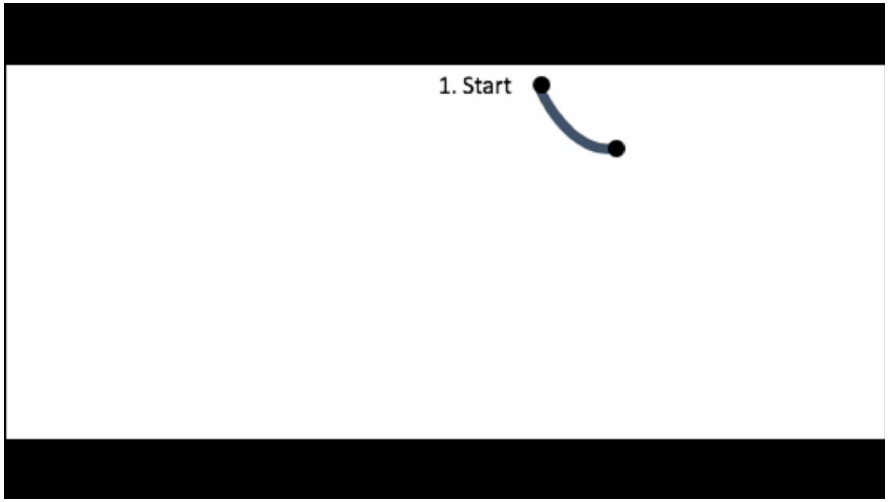
Operational Context in Surgery



Pick and Place Trajectory and Segments



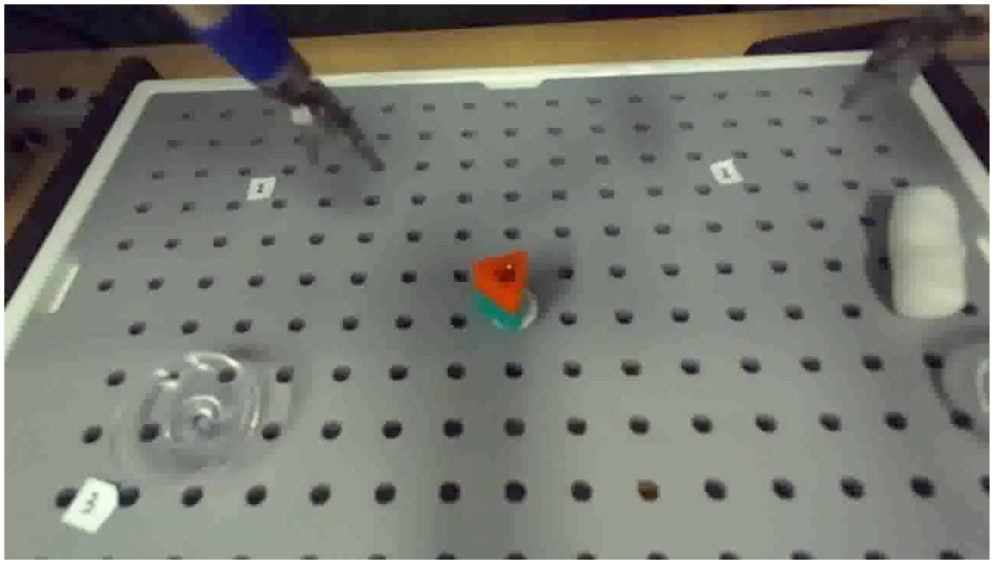
Tracking Surgical Trajectory

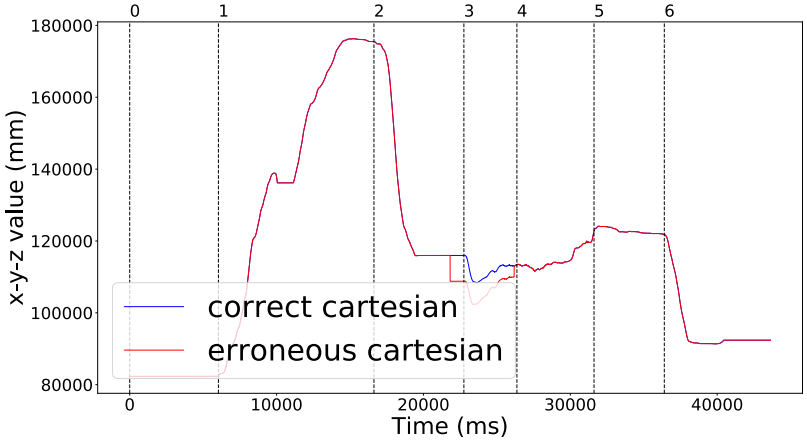


Mapping Trajectory into Segments

Failure Modes in Pick and Place Task

Failure	Cause	Segment
Unintentional release	Grasper angle too high or Wrong scale factor	4
Failure to dropoff	Grasper angle too low	5
Sudden Jump	Wrong Cartesian position Wrong scale factor	all

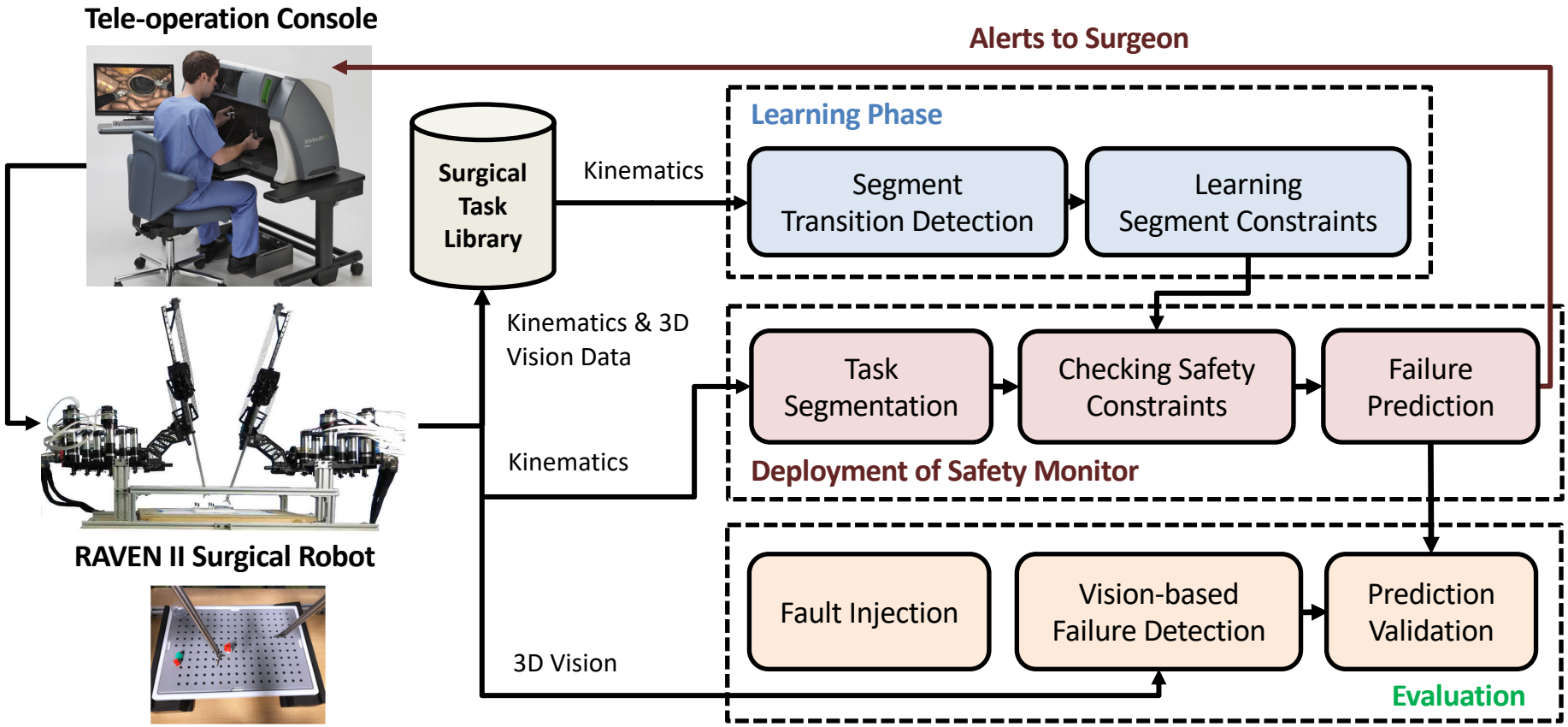




Our Solution

- Actively monitor the movement of the end-effector during fault-free demonstrations of a task
- Learn safety constraints to represent these fault-free behaviors
- Alert the surgeon if we detect a violation of the safety constraints

Context Aware Monitoring, Feedback, Control



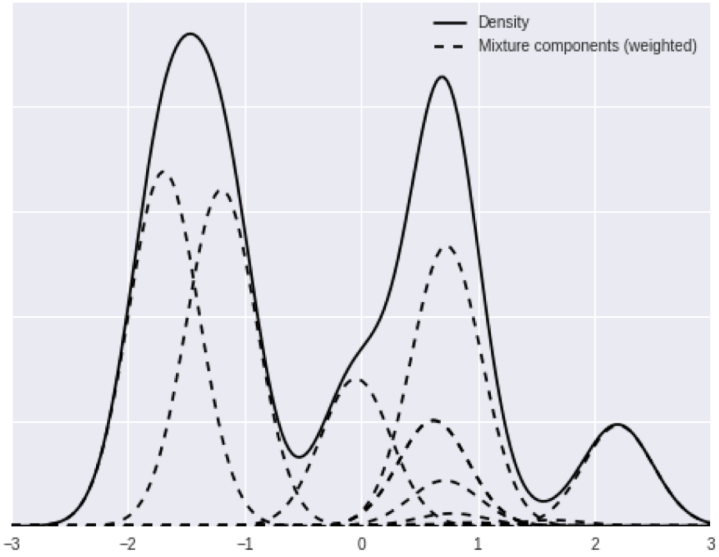
Yasar et al., ISMR 2019.

Detecting Transitions in Segments

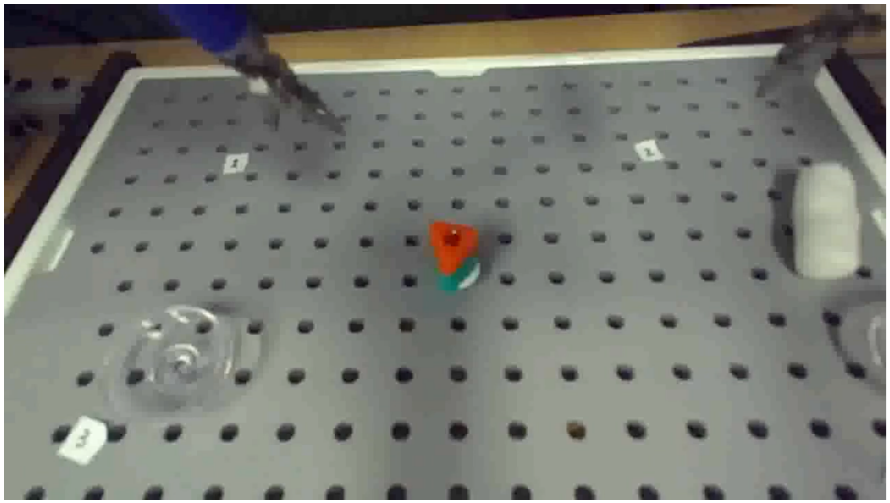
- Find the transitions between the subtasks using unsupervised **Gaussian Mixture Models (GMM)**

- Prior: Number of clusters
- Input to the GMM: Robotic Joint Kinematics values (e.g., position and velocity)

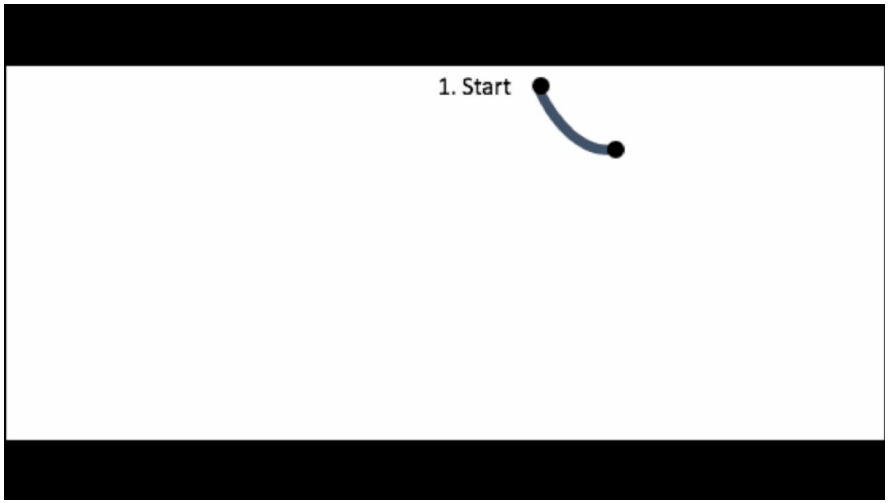
- Previous work (Krishnan et al. 2017) used both kinematics and vision information for detecting segments



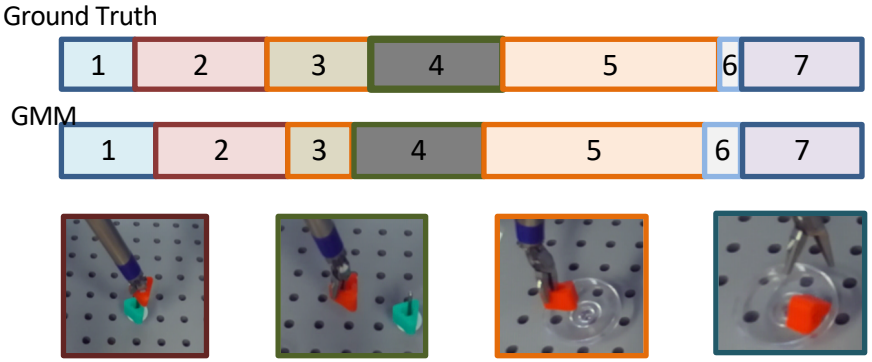
Automated Inference of Context



Tracking Surgical Trajectory



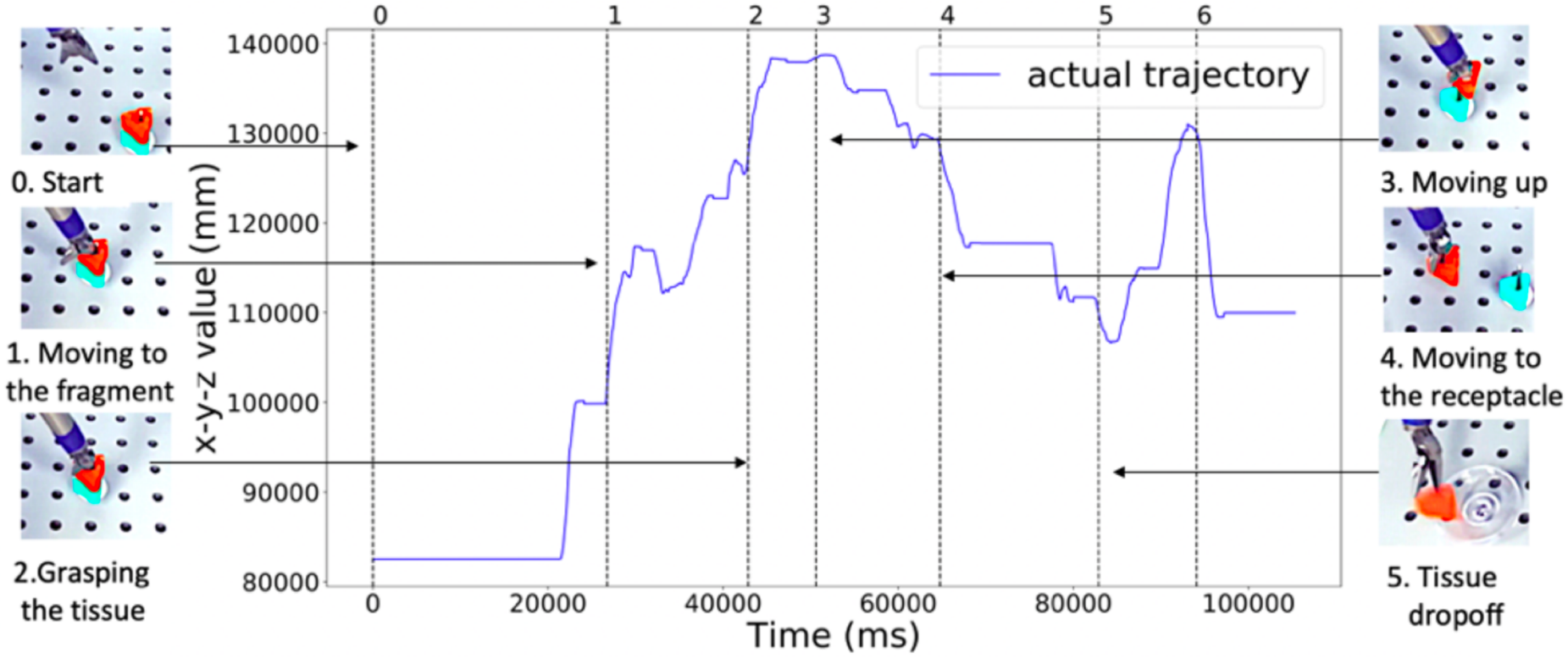
Mapping Trajectory into Segments



Segment 3 Segment 4 Segment 5 Segment 6

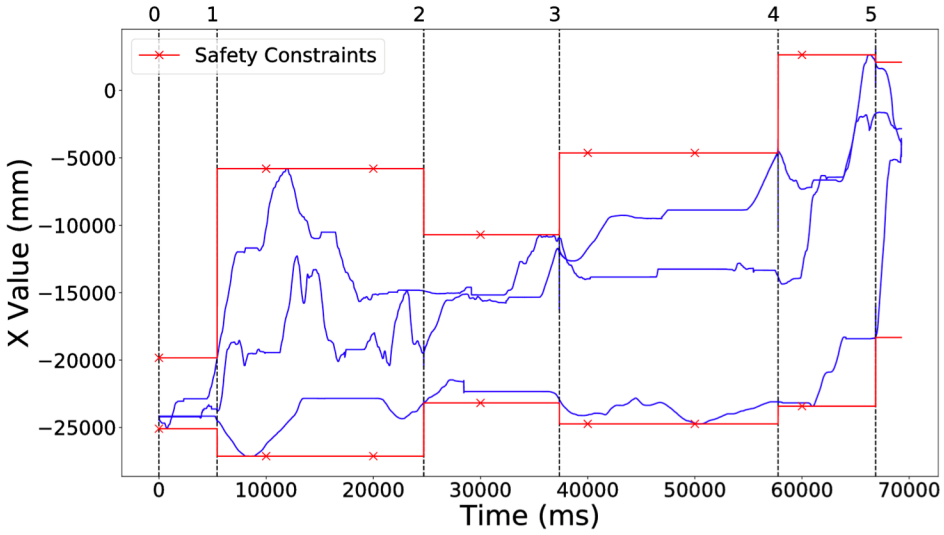
Subtask	Name	Avg. Error (Δt in frames)
0	Start	-56
1	Moving to the block	76
2	Grabbing the block	-69
3	Moving up	-30
4	Moving to the receptacle	-10
5	Dropping the block	-3
6	End	0

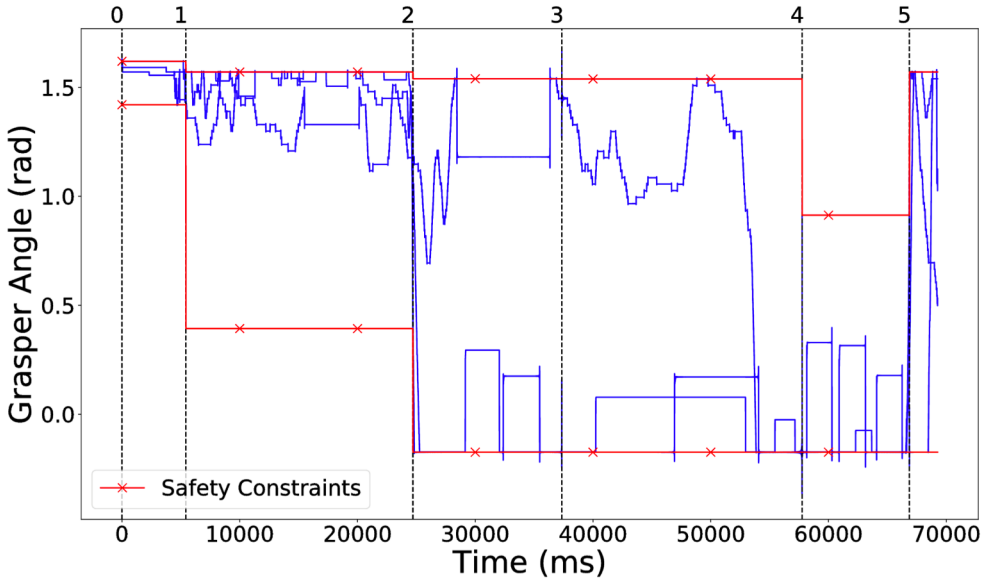
Segment-Specific Safety Constraints



Yasar et al., ISMR 2019.

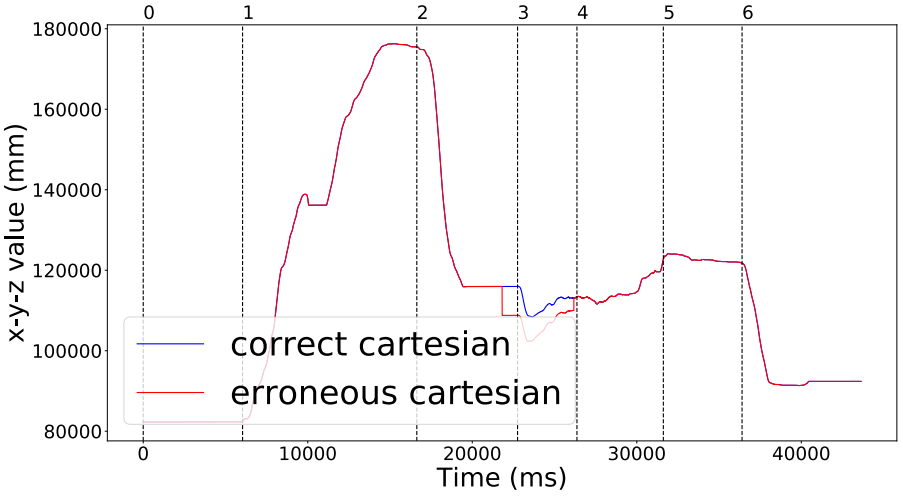
Segment-Specific Safety Constraints



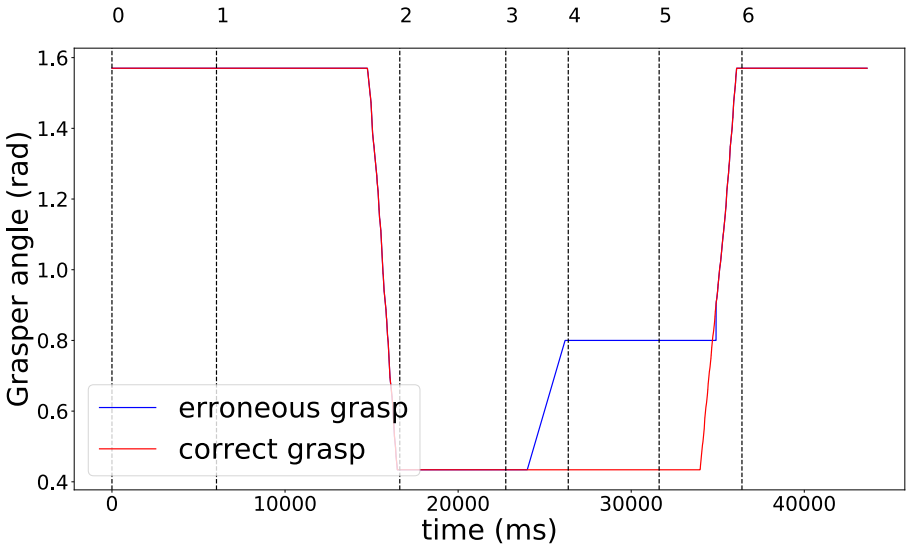


Yasar et al., ISMR 2019.

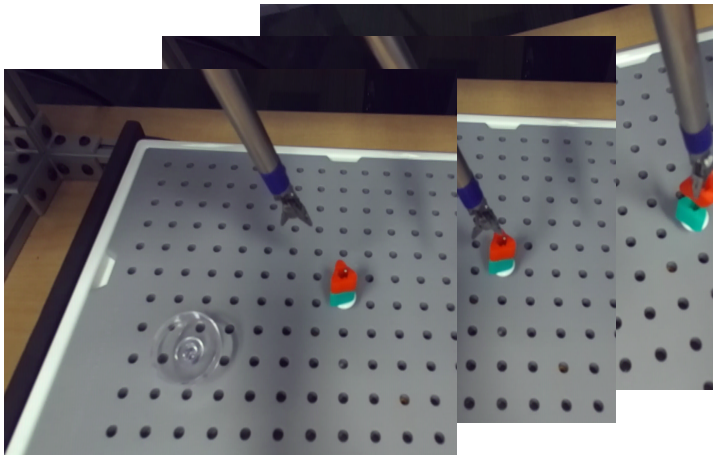
Failure Modes and Fault Injections



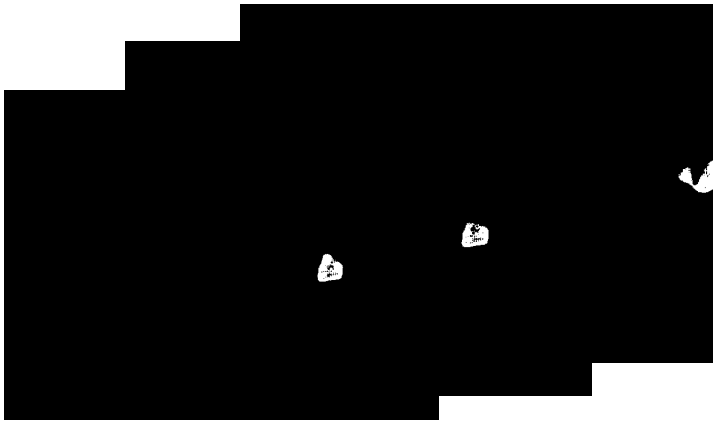
Failure	Cause	Segment
Unintentional release	Grasper angle too high or Wrong scale factor	4
Failure to dropoff	Grasper angle too low	5
Sudden Jump	Wrong Cartesian position Wrong scale factor	all



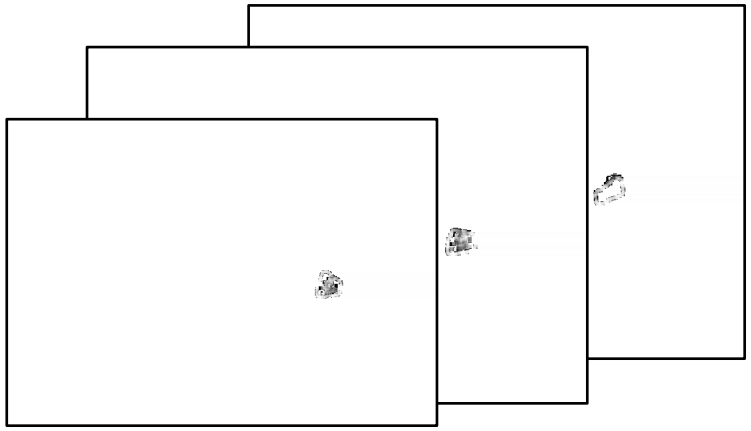
Failure Detection using SSIM



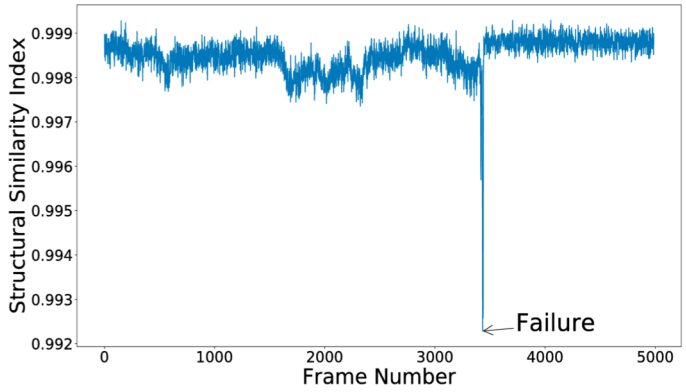
HSV thresholding



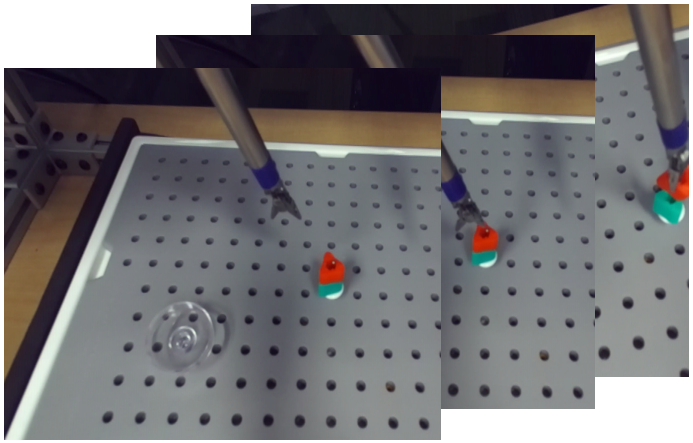
Diff (SSIM)



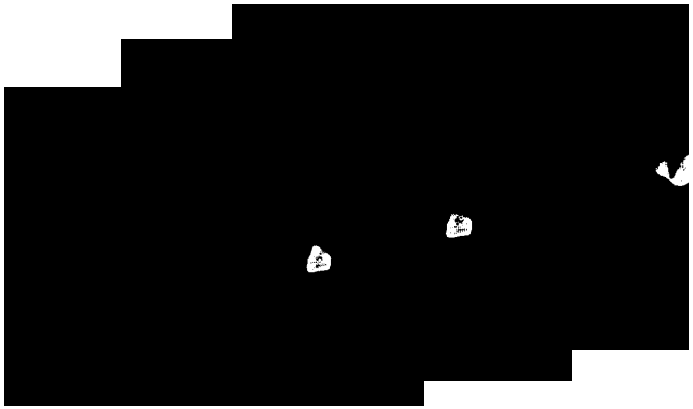
Finding minimum SSIM



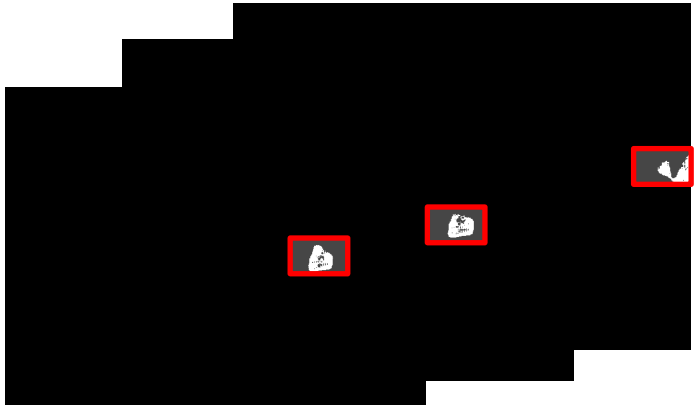
Failure Detection via Dynamic Time Warping



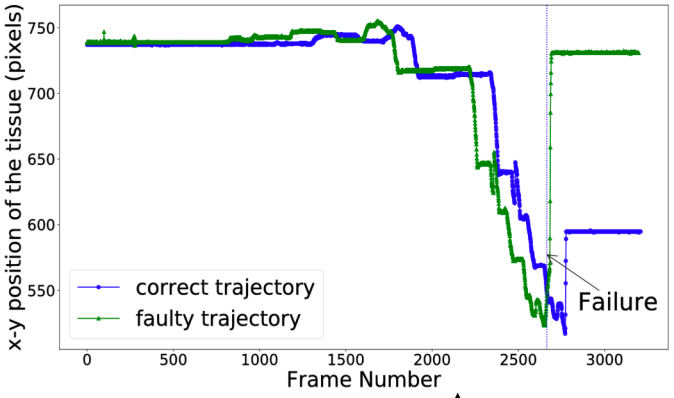
HSV thresholding



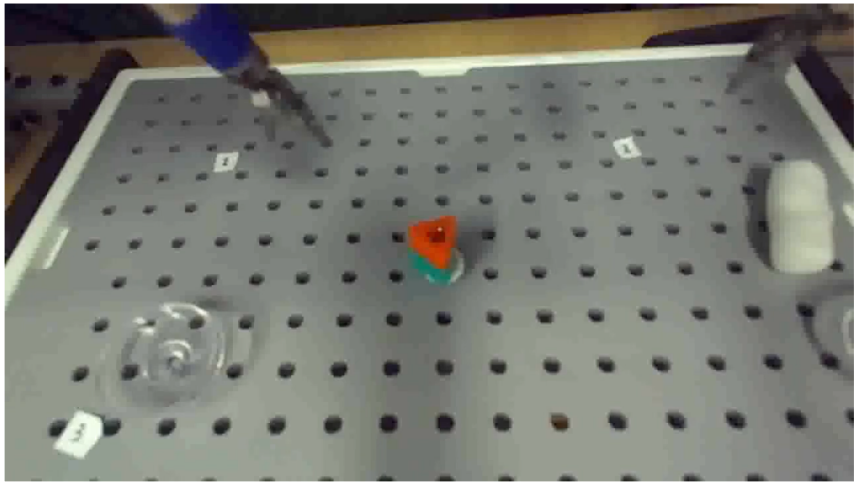
Find contour



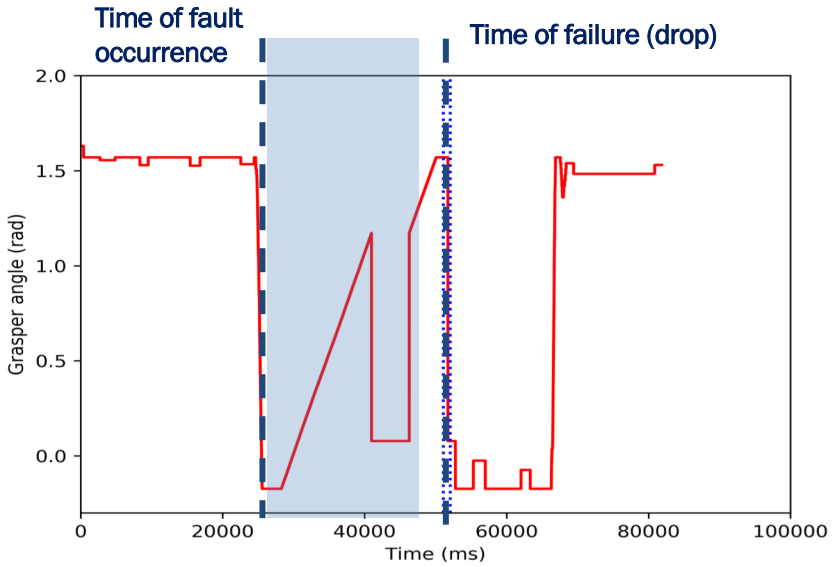
Compute center



Early Detection of Safety-Critical Events



Checking Segment Specific Constraints

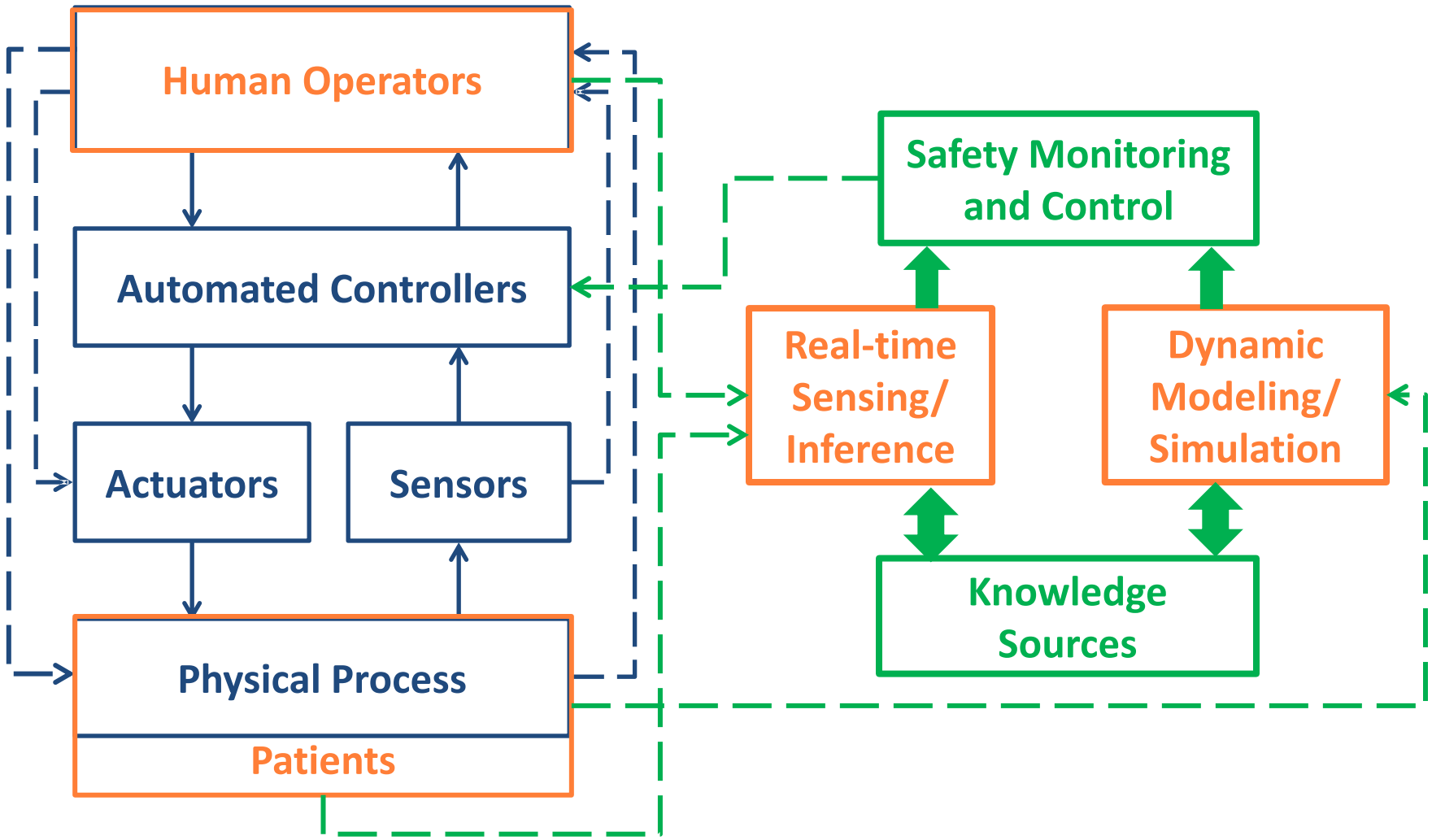


Detection Window

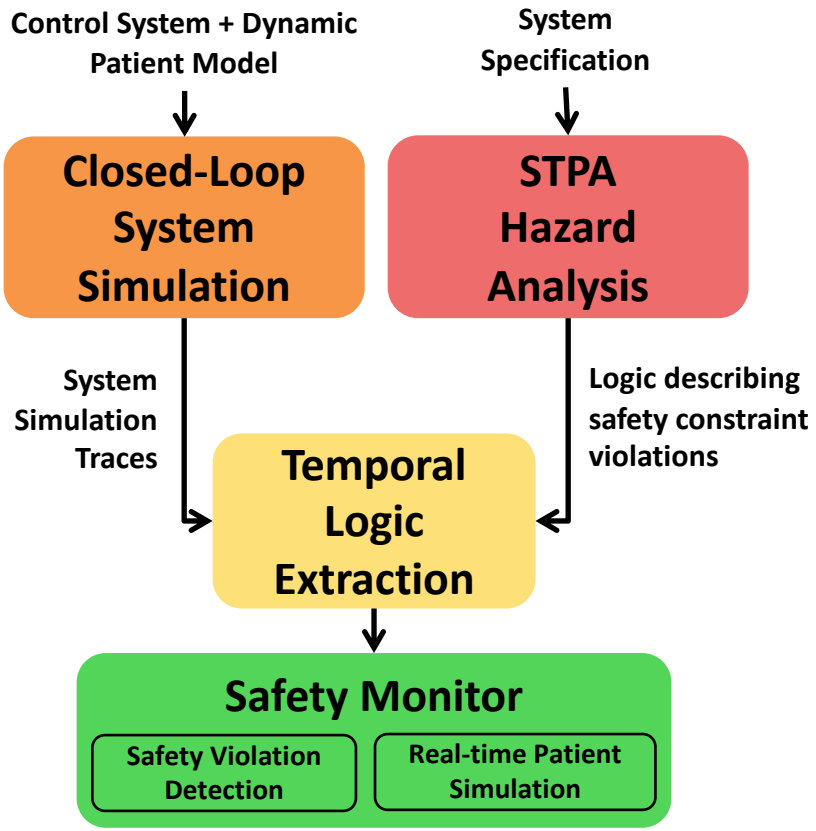
	Overall	Simulation	Dry Lab
Number of Experiments	518	468	50
True Positives	419	398	21
False Positives	95	70	25
True Negatives	3	0	3
False Negatives	1	0	1
Accuracy	**80.8%**		
False Negative Rate	**0.24%**		
False Positive Rate	**96.9%**		

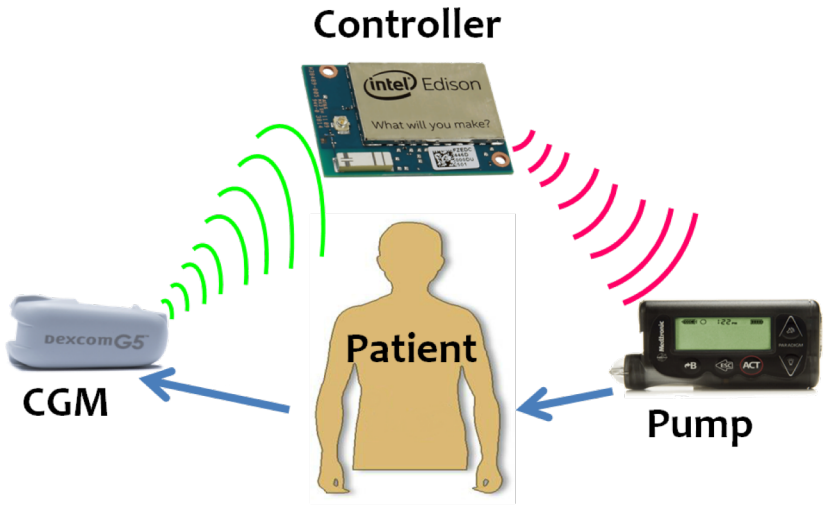
Simulated Failure	Average Reaction Time
Sudden Jump	1.7s
Block Drop	14.4s

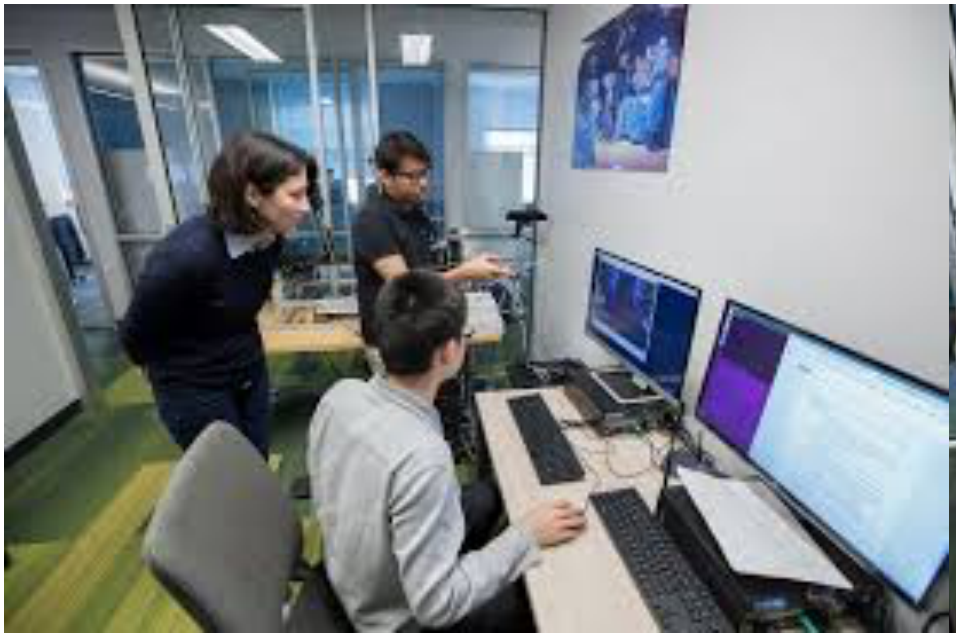
Context-Aware Safety Monitoring



Automated Synthesis of Context-Aware Safety Monitors







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