

Smart Robotic Assistants for Composite Prepreg Sheet Layup

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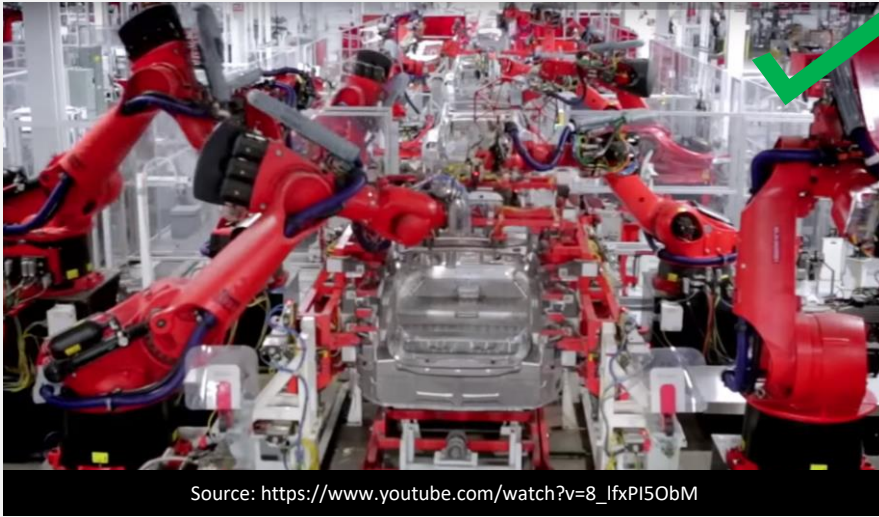
Today's Industrial Robots



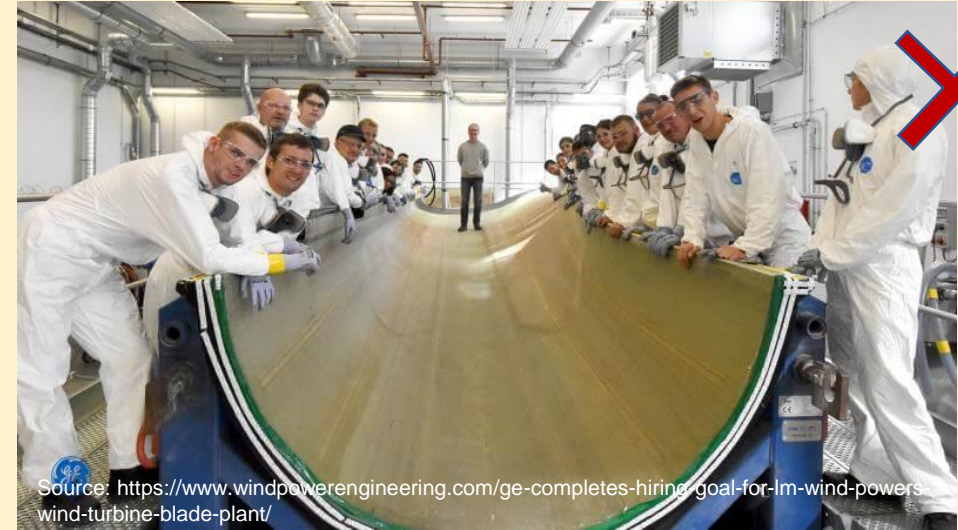
Robots are physically capable of performing highly complex tasks

Where are Robots used in Manufacturing?

Mass Production



High Mix Applications



Humans are still doing these tasks



Source: SnugTop



Source: <https://www.peritustfoam.com/open-cell-spray-foam/>



Source: Fiberglass Part Manufacturer



Source: <https://www.familyhandyman.com/project/drywall-sanding-tips-and-techniques/>



Source: <https://www.aviano.af.mil/News/Articles/News-Display/Article/280626/aviano-unveils-first-locally-painted-f-16/>



Source: https://www.wvnews.com/news/wvnews/wind-energy-has-potential-to-propel-wv-forward/article_c39d6cd5-6ea1-581d-ba3b-ef92203c4278.html

Limitations of Industrial Robots in Manufacturing

- Configuring a robotic cell for a new task takes significant time and effort
- Need human experts to program robots
- Robots repeat preprogrammed motions and cannot automatically adapt to changes in the workspace or tasks
- Recovering from errors requires significant downtime and can be very expensive



Source: <https://www.fanucamerica.com/solutions/applications/material-removal>

Robots are largely used in mass production applications.
Less than 2.5 robots for every 100 manufacturing workers in US.

Challenges Faced by Manufactures

- High labor churn in tedious and ergonomically challenging jobs
- Many unfilled jobs
- Traditional industrial robots cannot be used in high mix applications

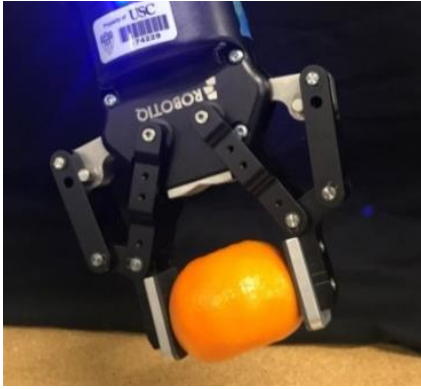


Growth in robotics deployment will mainly come from high mix applications

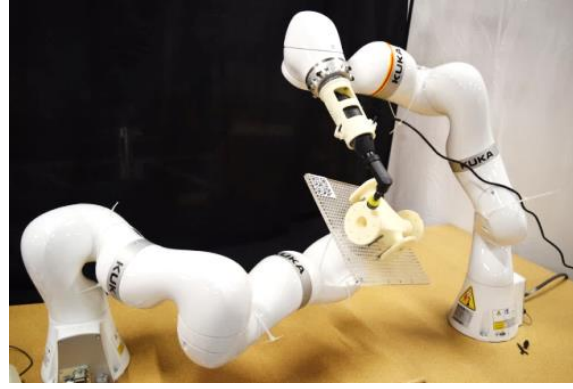
Recent Advances in Robotics



Stereo Vision
Force Sensing
Tactile Sensing



Impedance Control
Visual Servo
Shape Control



Multi-Arm
Manipulation



Mobile
Manipulation



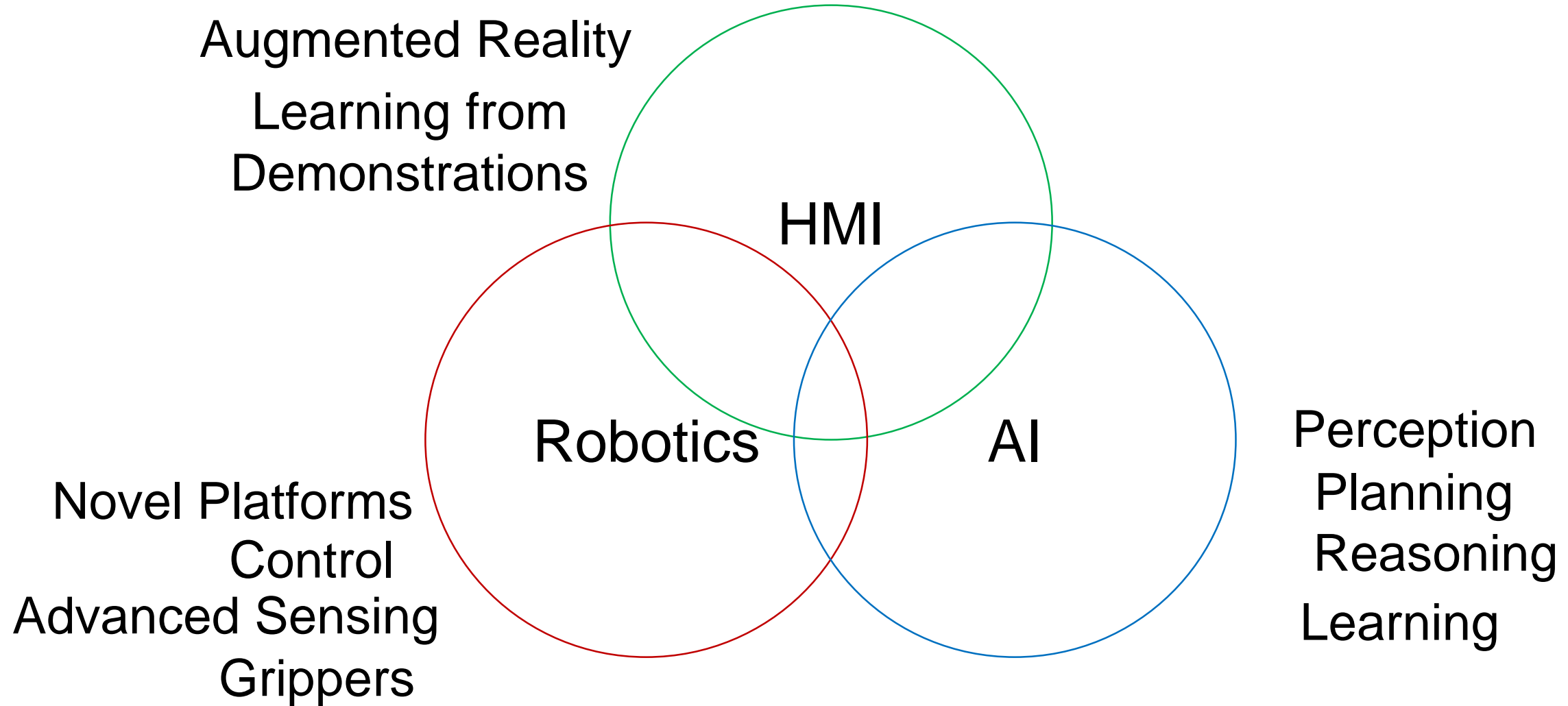
Collaborative
Robots

Relying on humans to program robots is not a viable option as robotic cells get more complex in high mix applications

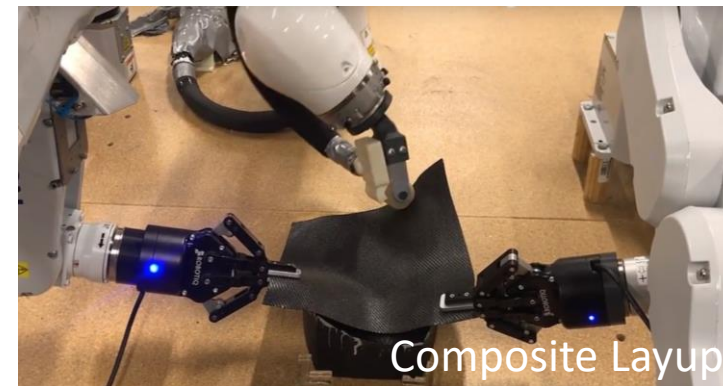
Our Goal

*Develop robotic assistants for
high mix manufacturing applications to
increase human productivity and reduce health risks*

Our Focus: Smart Robotic Assistants



Manufacturing Applications



Current Automation in Composites

- Current automation techniques are focused on tape layup and fiber placement
- These processes are used to manufacture large part with simpler geometries
- More complex and smaller parts are manufactured using prepreg composite layup



Image Source:
<https://www.researchgate.net/profile/Russell-Wanhill/publication/309962841/figure/fig6/AS:781657061736450@1563372974811/Automated-tape-laying-CFRP-tape-layers-being-placed-on-the-lay-up-tool-for-an-Airbus.jpg>

Motivation

- Prepreg layup process is manual and skill dependent
- A typical part consists large number of sheets being stacked on top of each other making the process cumbersome
- Manual process is susceptible to defects and rework
- Carbon fiber prepregs are difficult to recycle and have an adverse effect on the environment, hence reducing rework and waste is important for sustainability

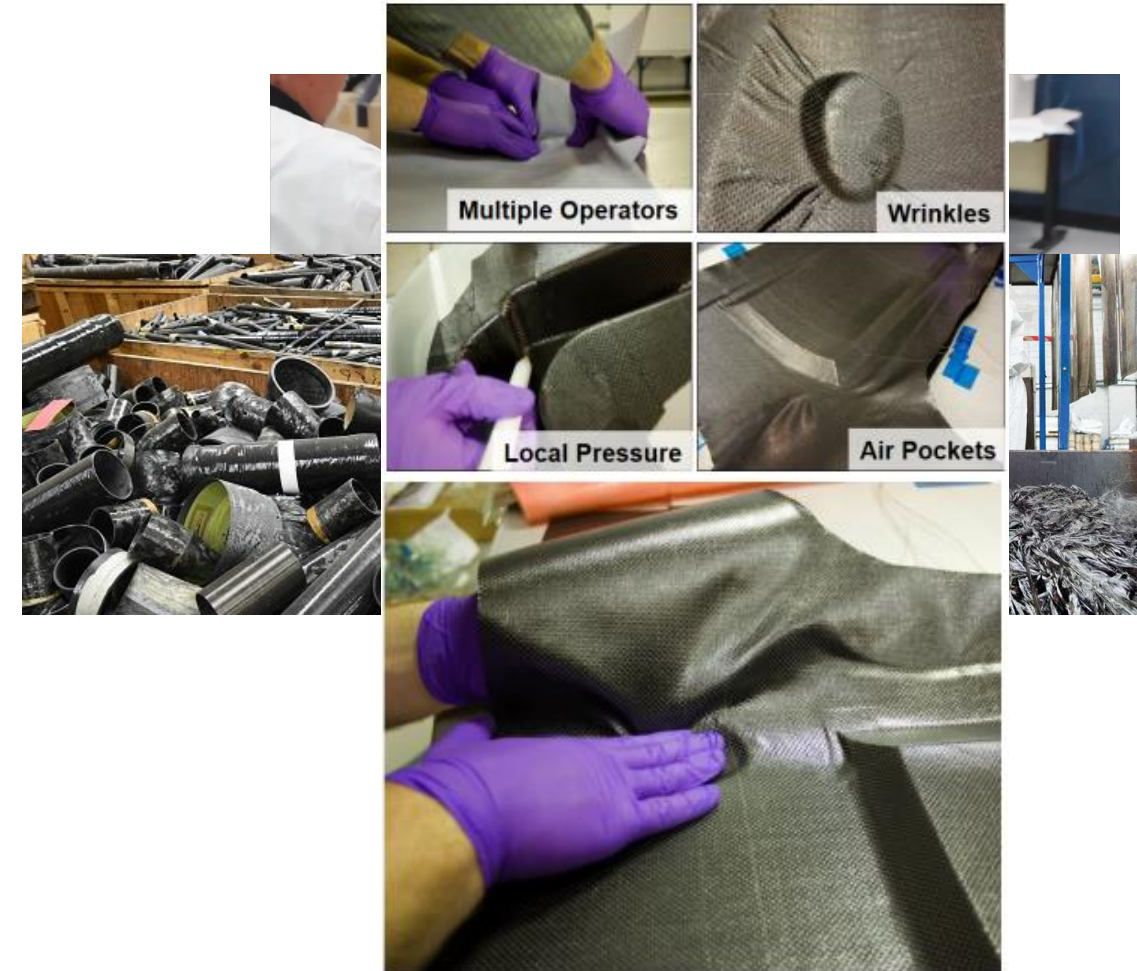
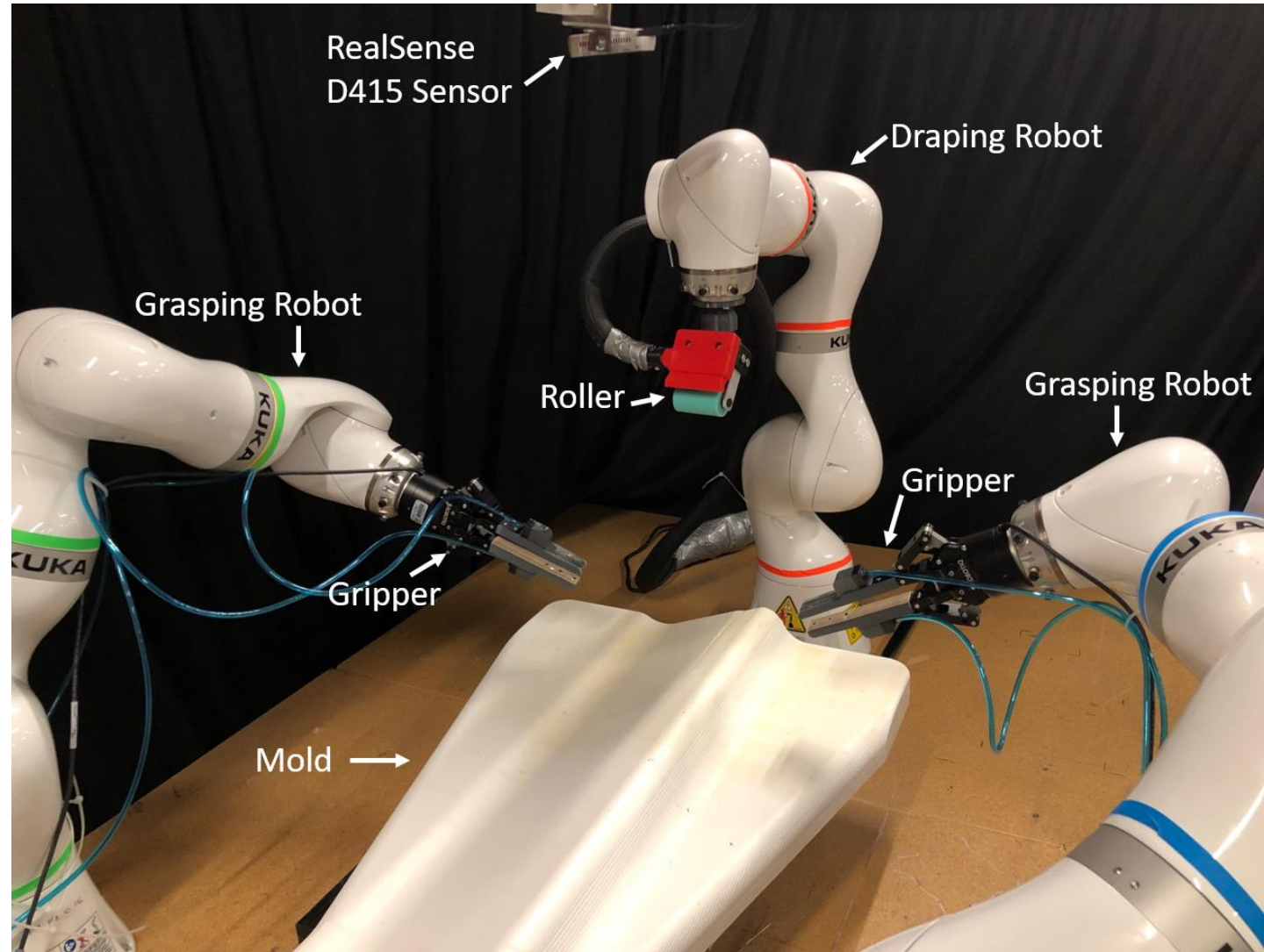
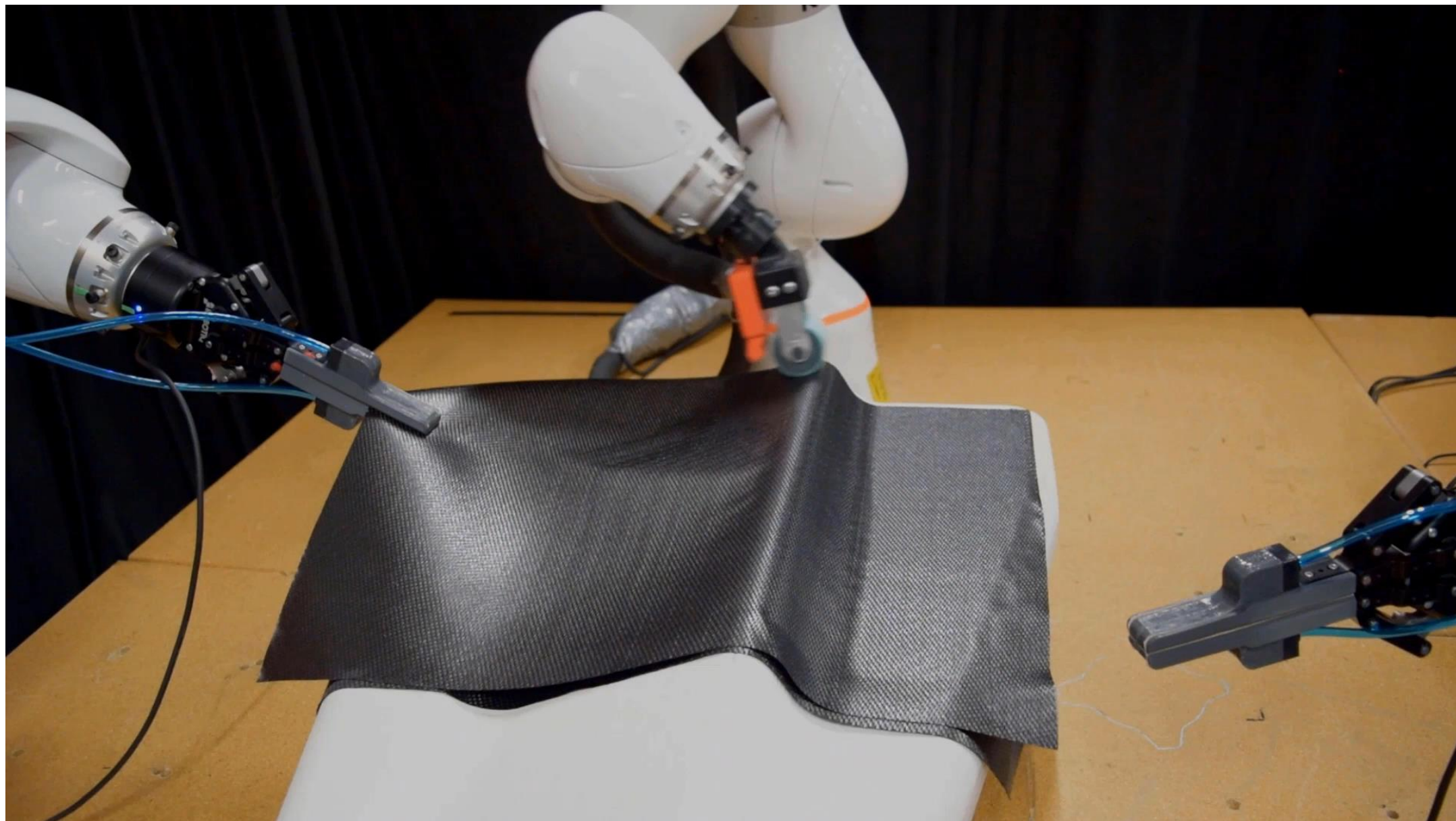


Image Source:
https://d2n4wb9orp1vta.cloudfront.net/cms/brand/CW/2017-CW/0817CW_Recycle_ELG_EG416_5281.jpg?width=550;quality=60
<https://www.innovationintextiles.com/next-step-in-recycling-carbon-thermosets/>

Robotic Cell for Small Part Layup





Results

- Sheet layup operation can be automated using a robotic cell
- Human competitive layup speeds can be achieved
- Robot instructions can be generated automatically without any need for robot programming
- Layup process can be monitored, and interventions can be performed to prevent defect formation



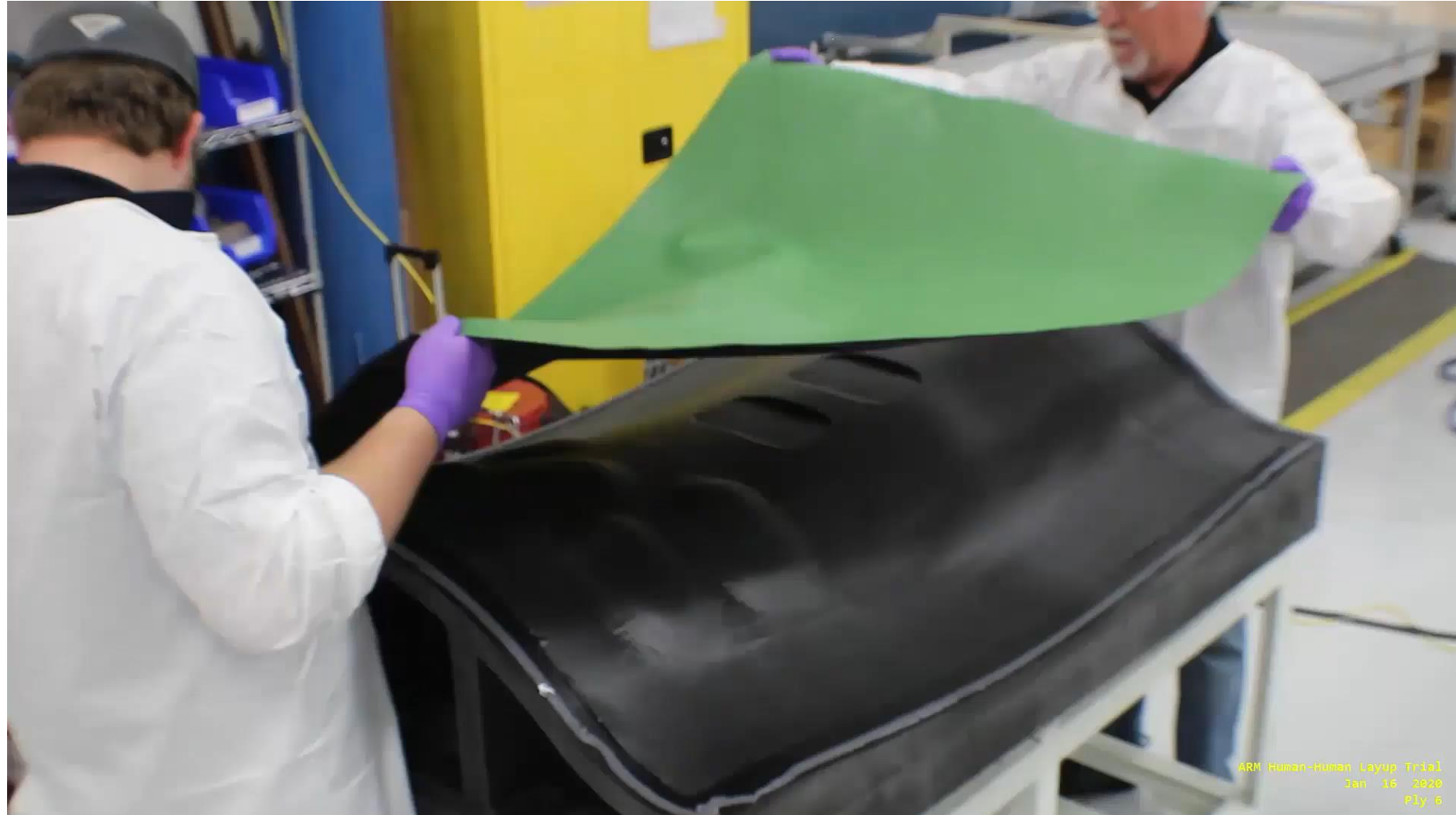
Total Number of Layers 15

Large ply layup time (9 plies): 12 min/ply

Small ply layup time (6 plies): 6 min/ply

Total time: 144 min

Manual Operation Showing Large Part Layup



Hybrid Cell for Large Part Layup

RealSense
D415 Cameras

Draping
Robot

Tool

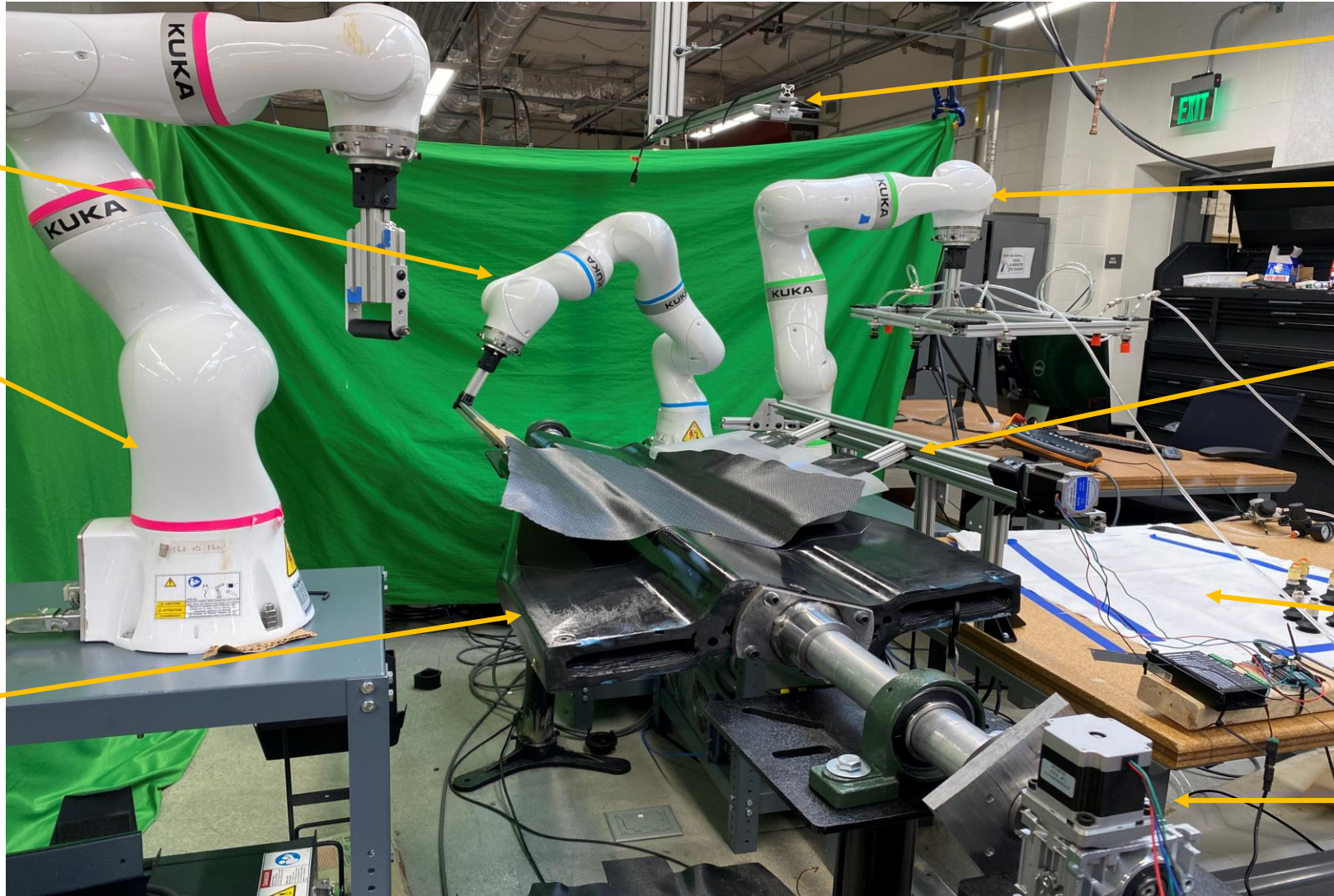
Grasping Robots

Kinect for
Human
Gesture
Recognition





Layup on Complex Rotating Tools



Sheet Support
Robot

Draping Robot

Rotating Tool

RealSense
Setup

Sheet Transport
Robot

Auxiliary Sheet
Support Tool

Prepreg
Station

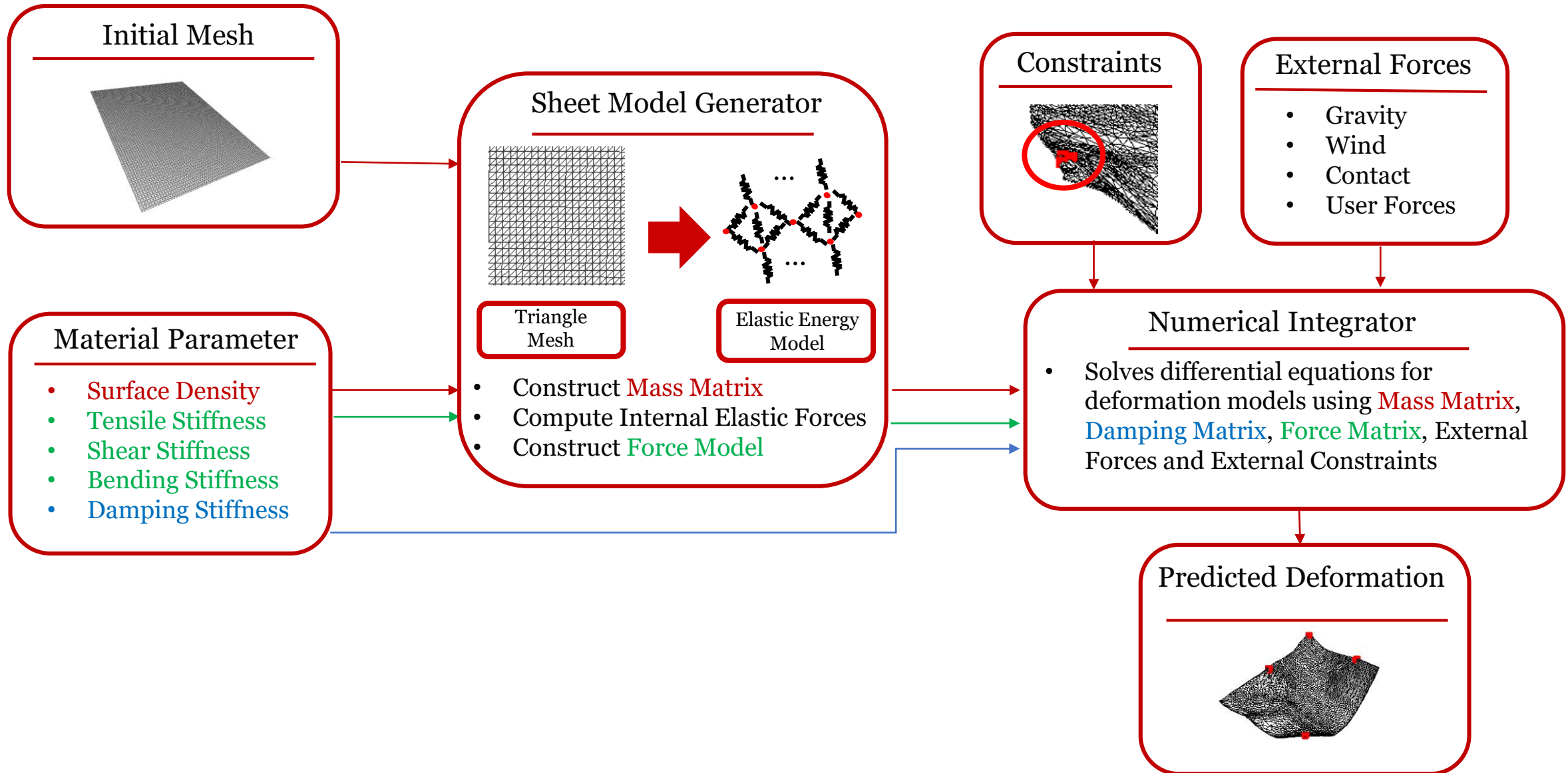
Tool Rotary
Mechanism

Key Technologies

- Digital Twins
- Robot Trajectory Planning
- Deep Learning Based Defect Detection
- Visual Servo for Sheet Transport

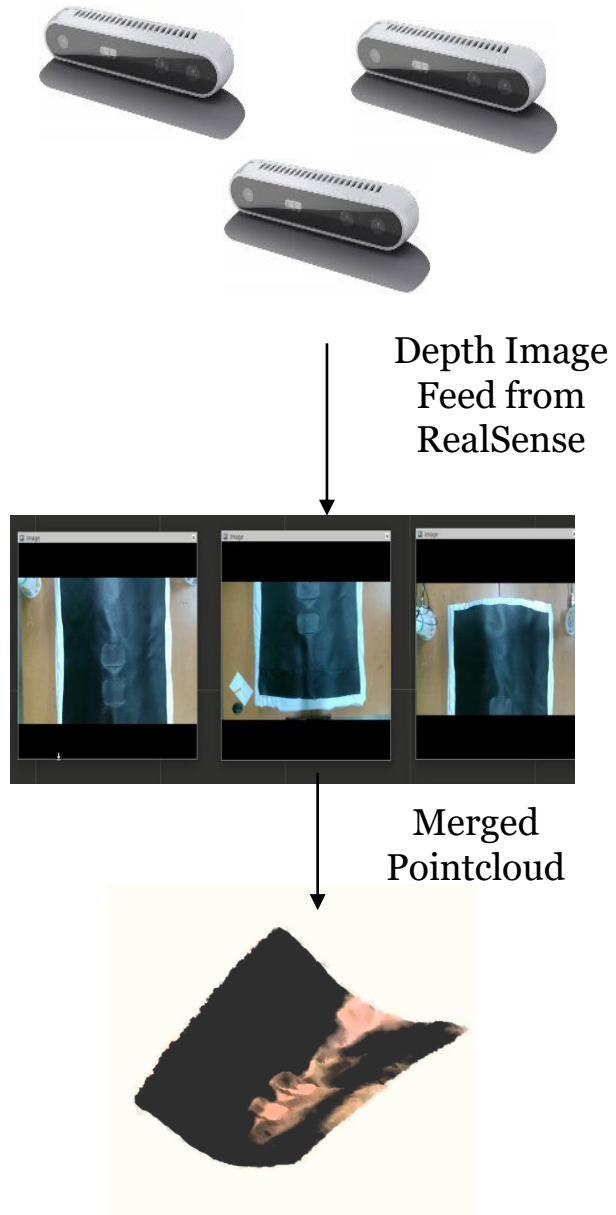
Creating Digital Twin for Prepreg Sheet Handling

Sheet Simulation

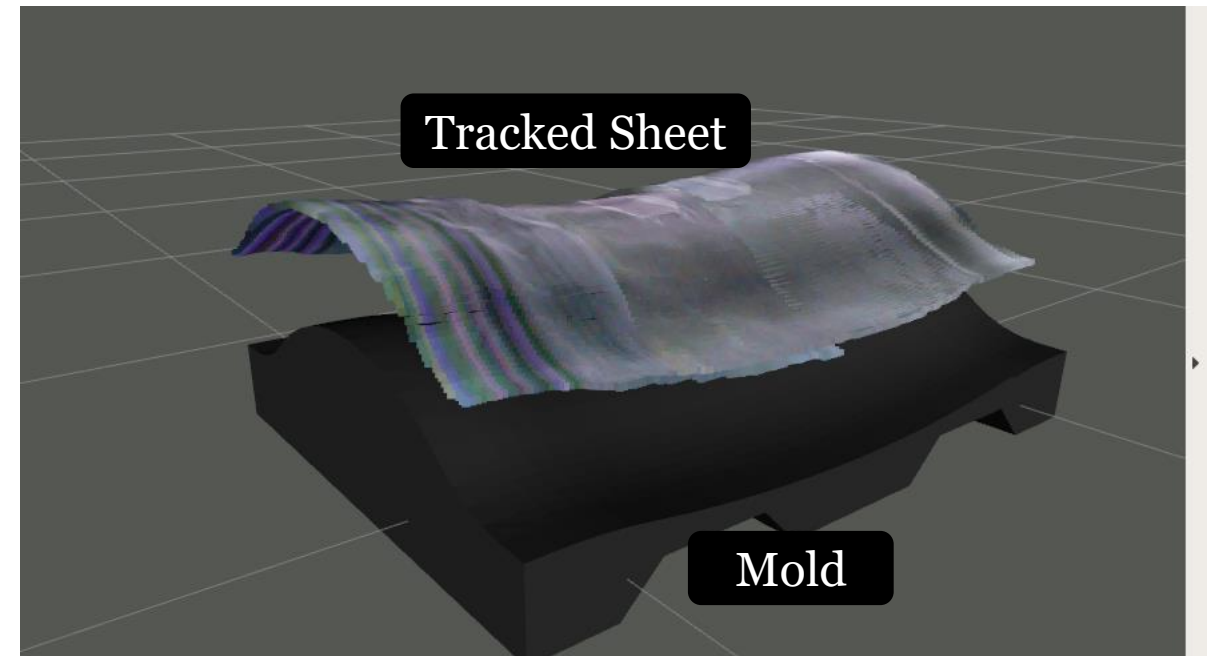


Sheet Simulation Demonstration

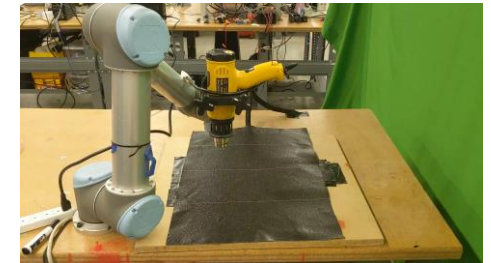
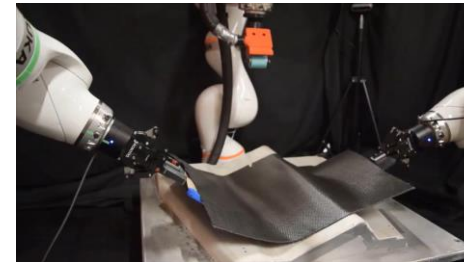
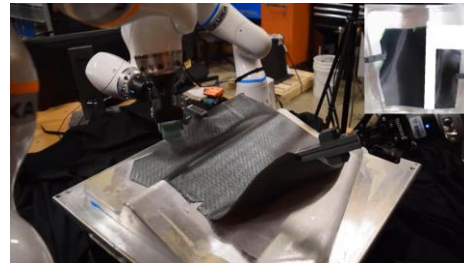
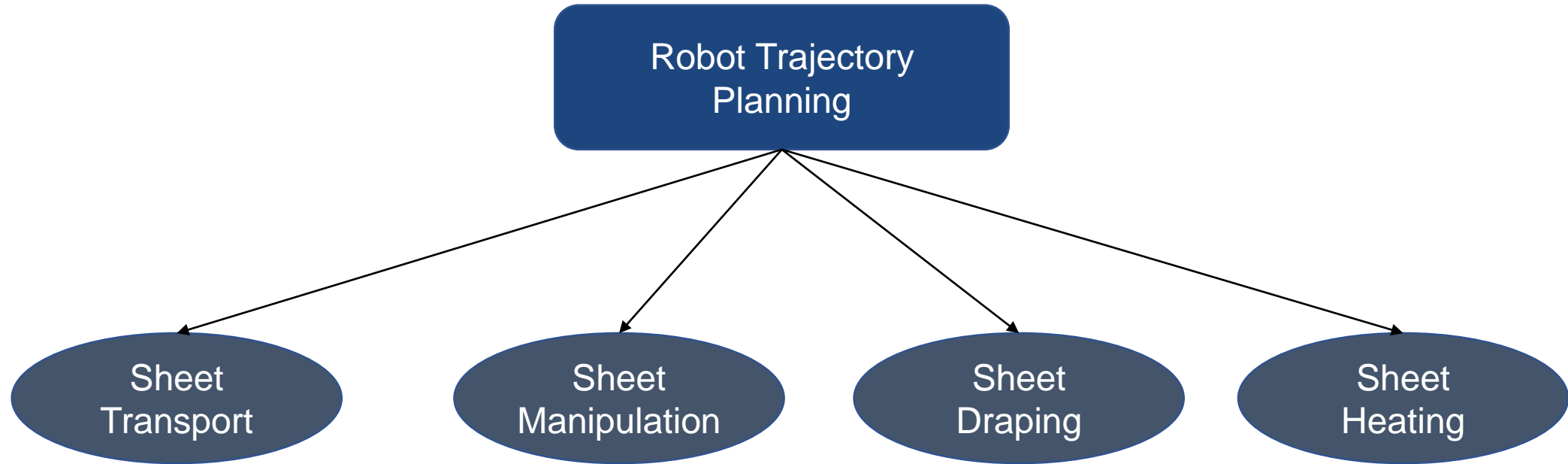
Real-Time Sheet Tracking



Tracked Data of the Composite Sheet in RVIZ Over Mold

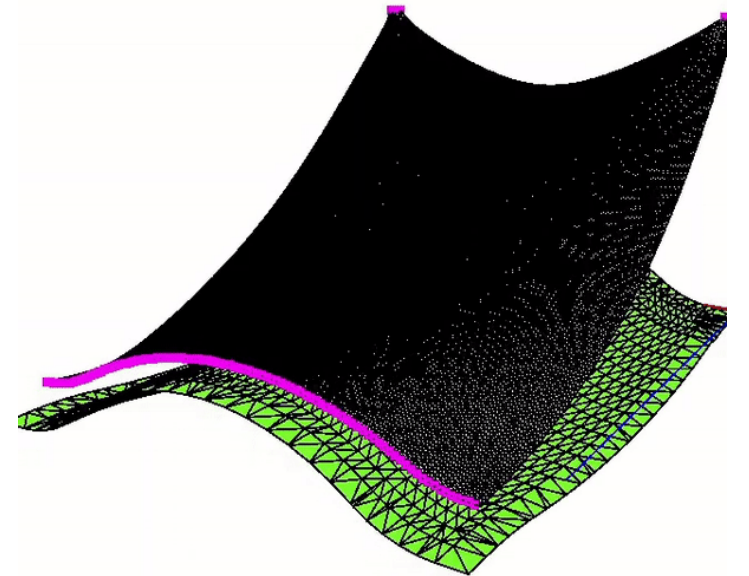
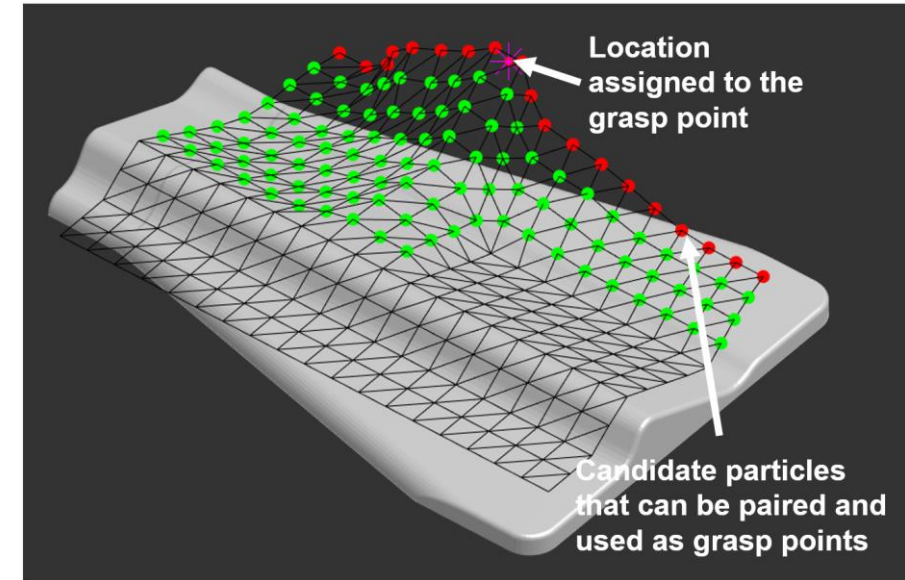


AI Enabled Robot Trajectory Planning



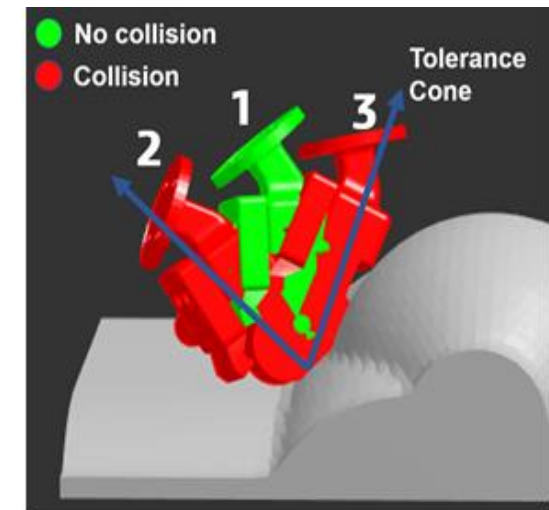
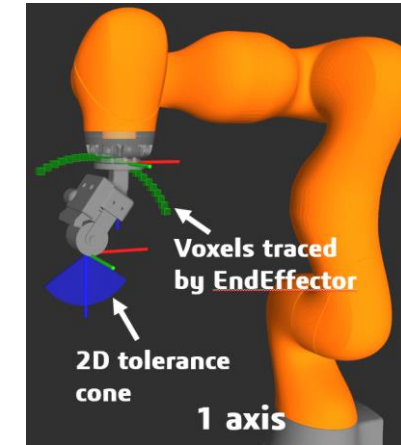
Grasp Planning

- Prepreg sheet is approximated by a mesh defined by position of vertices
- Vertices shown in green are the free points on the undraped sheet
- Vertices shown in red are grasping points
- The robot uses a pair of consecutive grasp points to move the sheet
- This planner implements a heuristic-based searching approach to find grasping locations that meet the criteria
- These simulated locations are subjected to certain process constraints to generate which is further optimized for shortest time path



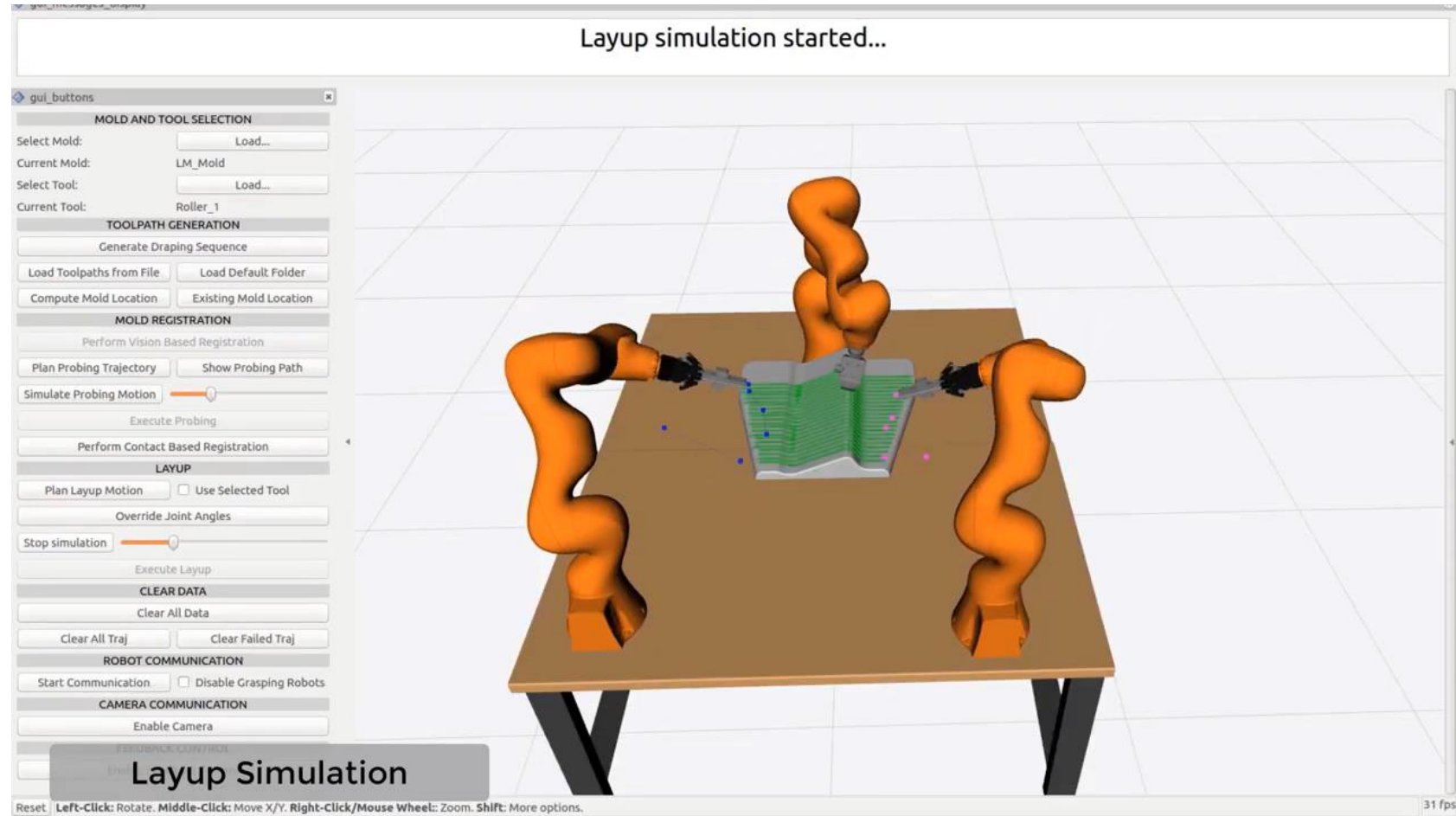
Sheet Draping Planning

- Human operators provides the draping paths over the mold by interacting with the User Interface
- These paths need to be used to compute robot trajectories, such that the process constraints are met
- Process constraints involve
 - Continuity constraints while executing each path
 - End-effector velocity constraints
 - Force constraints
 - Collision constraint between draping and grasping robots as well as the human
- Generate variable density waypoints by using surface normal information to increase the time of compaction at high curvature surfaces



Three orientations illustrated in the tolerance cone

Sheet Draping Planning



Sheet Heating Planning

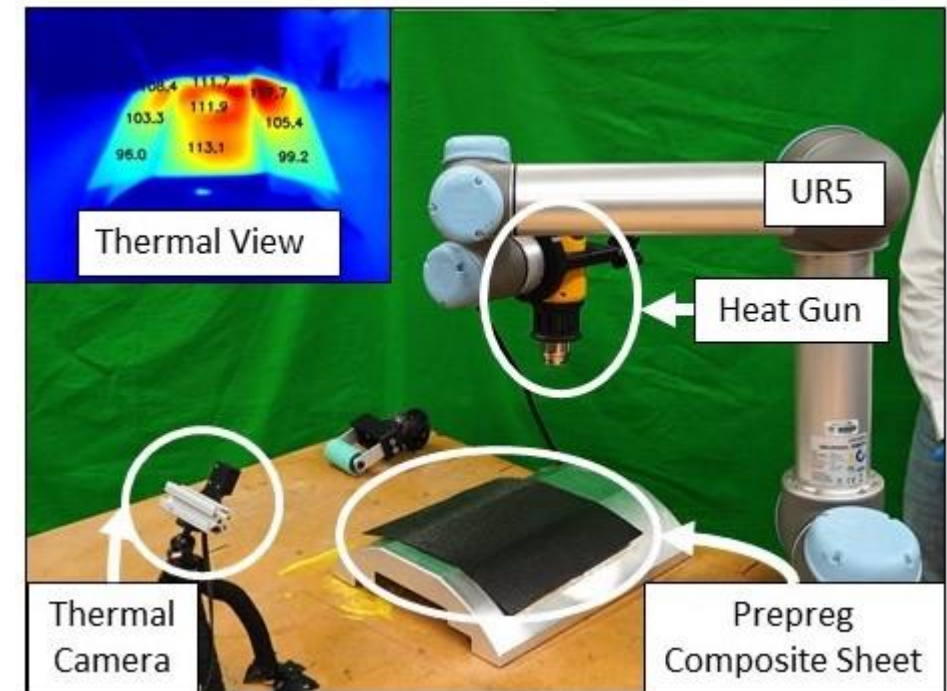
Objective: Maintain the temperature of the prepreg sheet within a desirable range conducive for draping without thermally damaging the prepreg

Challenges:

- Important to account for thermal characteristics of the sheet for robot planning
- Improper heating can exacerbate defect formation

Solution:

- Learn a temperature evolution model for the sheet
- Look-ahead search to find feasible state transitions for robot to maintain temperature constraints
- Use sensors to monitor the sheet state and achieve adaptive control



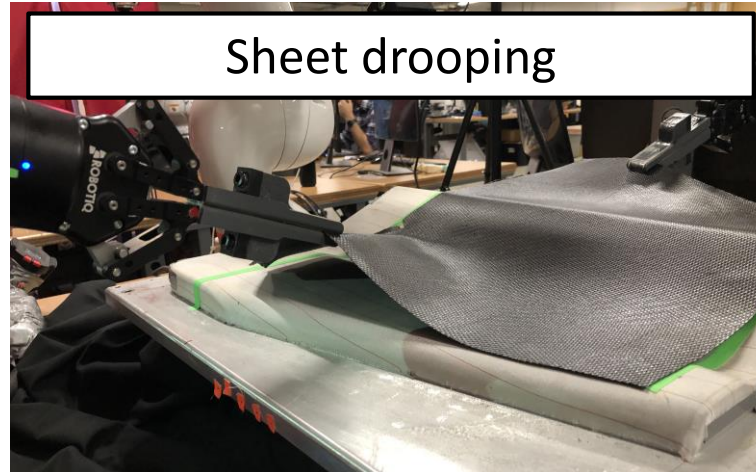
Deep Learning Based Defect Detection



Common Defects: Drooping and Wrinkles



Sheet without drooping



Sheet drooping



Partial sheet drooping



Wrinkle due to sagging of
prepreg plies



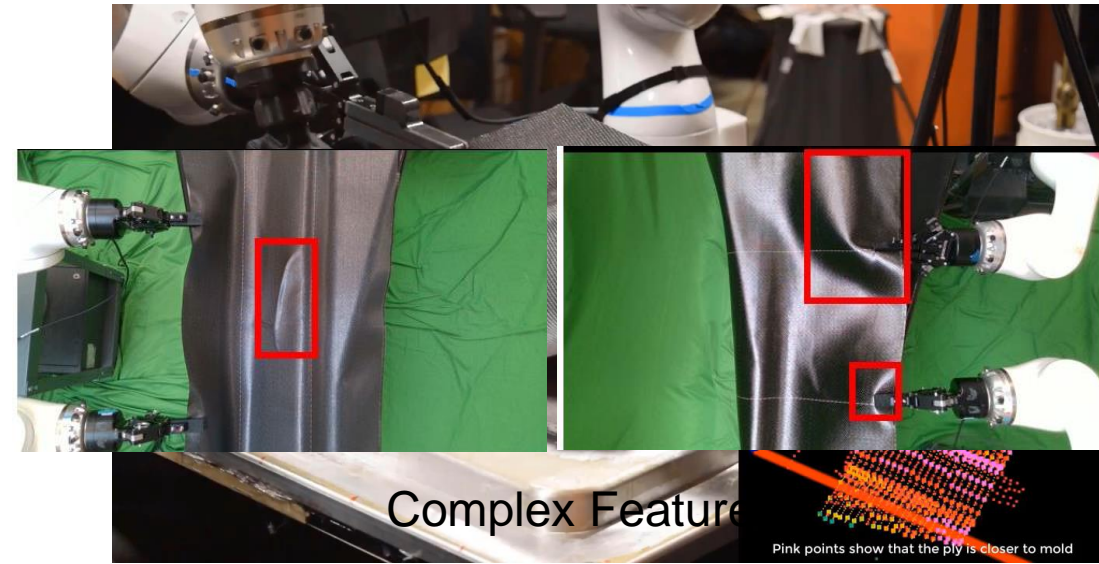
Wrinkle due to inappropriate
grasping robot location



Unresolvable wrinkle in
draped region

Deep Learning Based Defect Detection

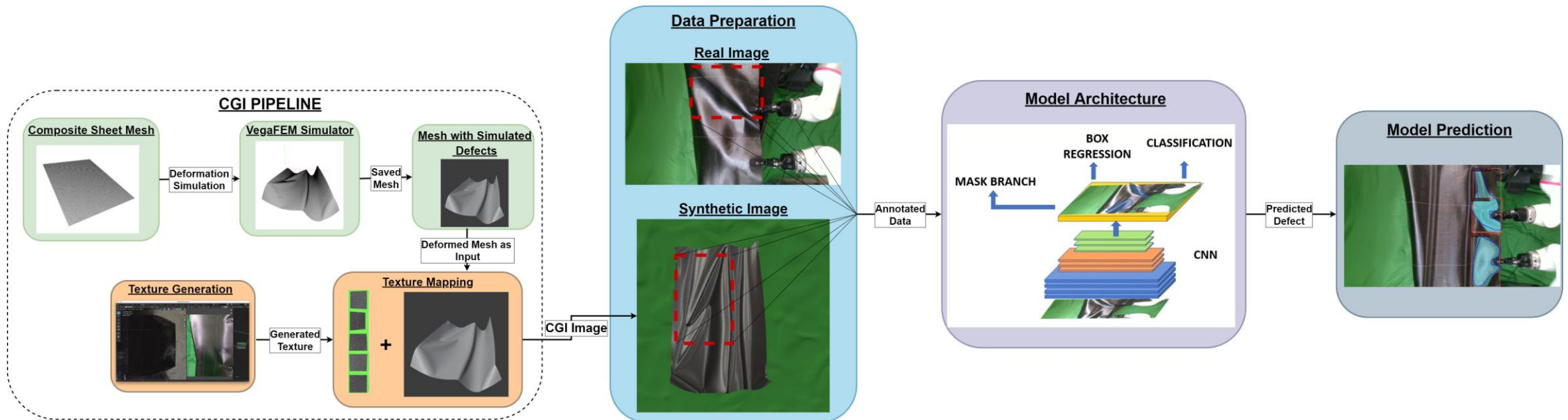
- Open loop execution of the robot motion plans can lead to poor quality and defect formation
- Detecting anomalous configurations and defects in-process is an important feature for any smart robotic cell
- Defects in a composite prepreg have complex features making them difficult to detect using traditional vision method
- Deep learning methods are more robust but pose the challenge of collecting large amounts of data



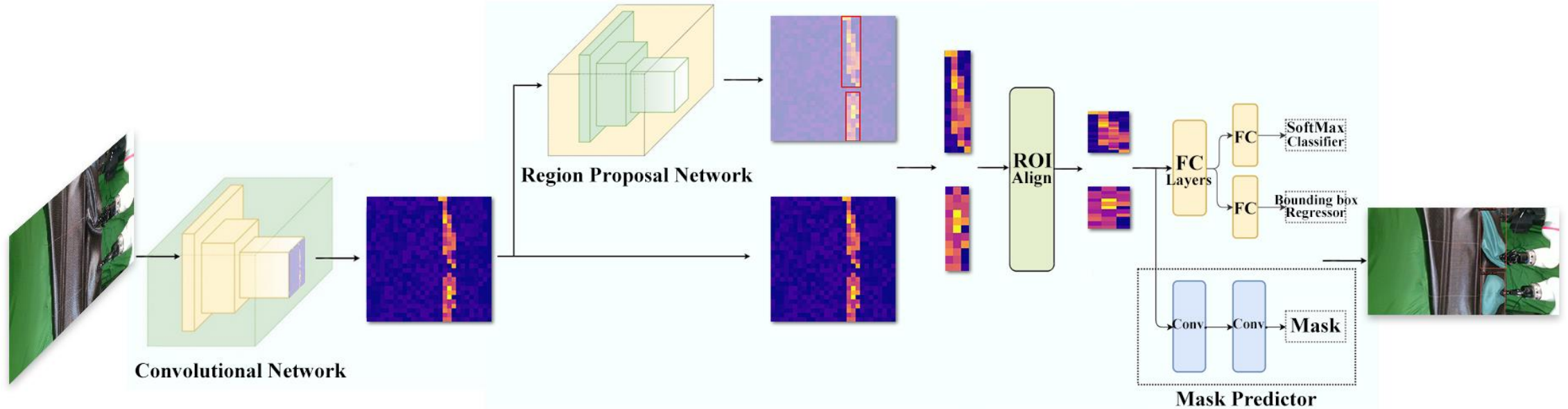
Solution: Physics Aware Synthetic Image Assisted Defect Detection

Deep Learning Based Defect Detection

Objective: Use the digital twin of the sheet combined with advanced CGI textures to generate feature aware synthetic images to train a deep learning model for defect detection



Deep Learning Network Architecture



- We use 2-Stage training for the network
- Mask-RCNN architecture is used
- On performing ablation studies, ResNet-50 is used as the backbone

Deep Learning Based Defect Detection

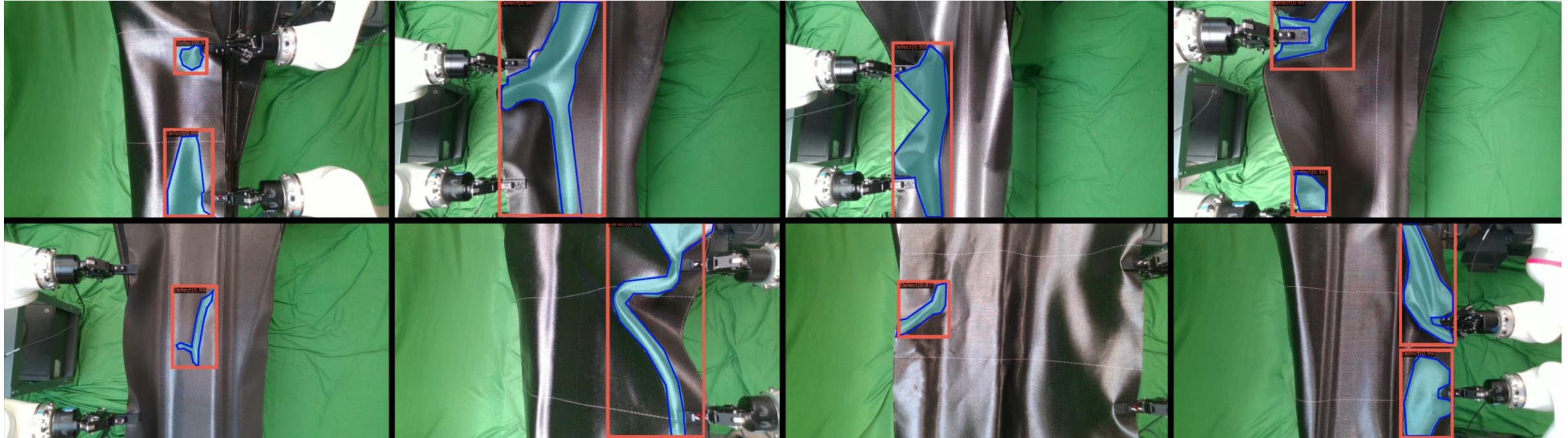
SYNTHETIC IMAGES



REAL IMAGES



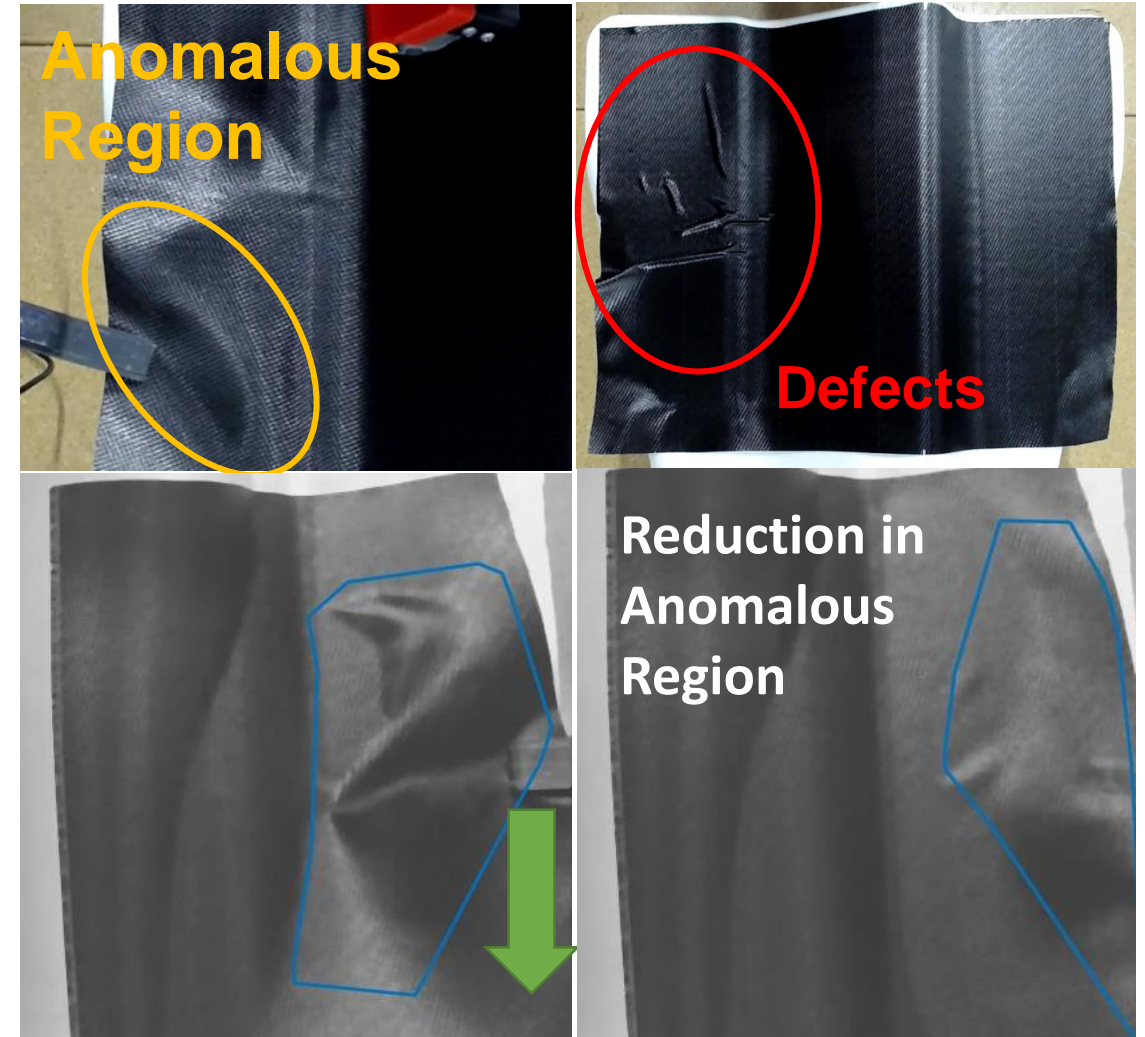
Results



- The deep learning-based defect detection system can detect wrinkle-based defects with 0.98 mean average precision (mAP)

Preventing Defect Formation

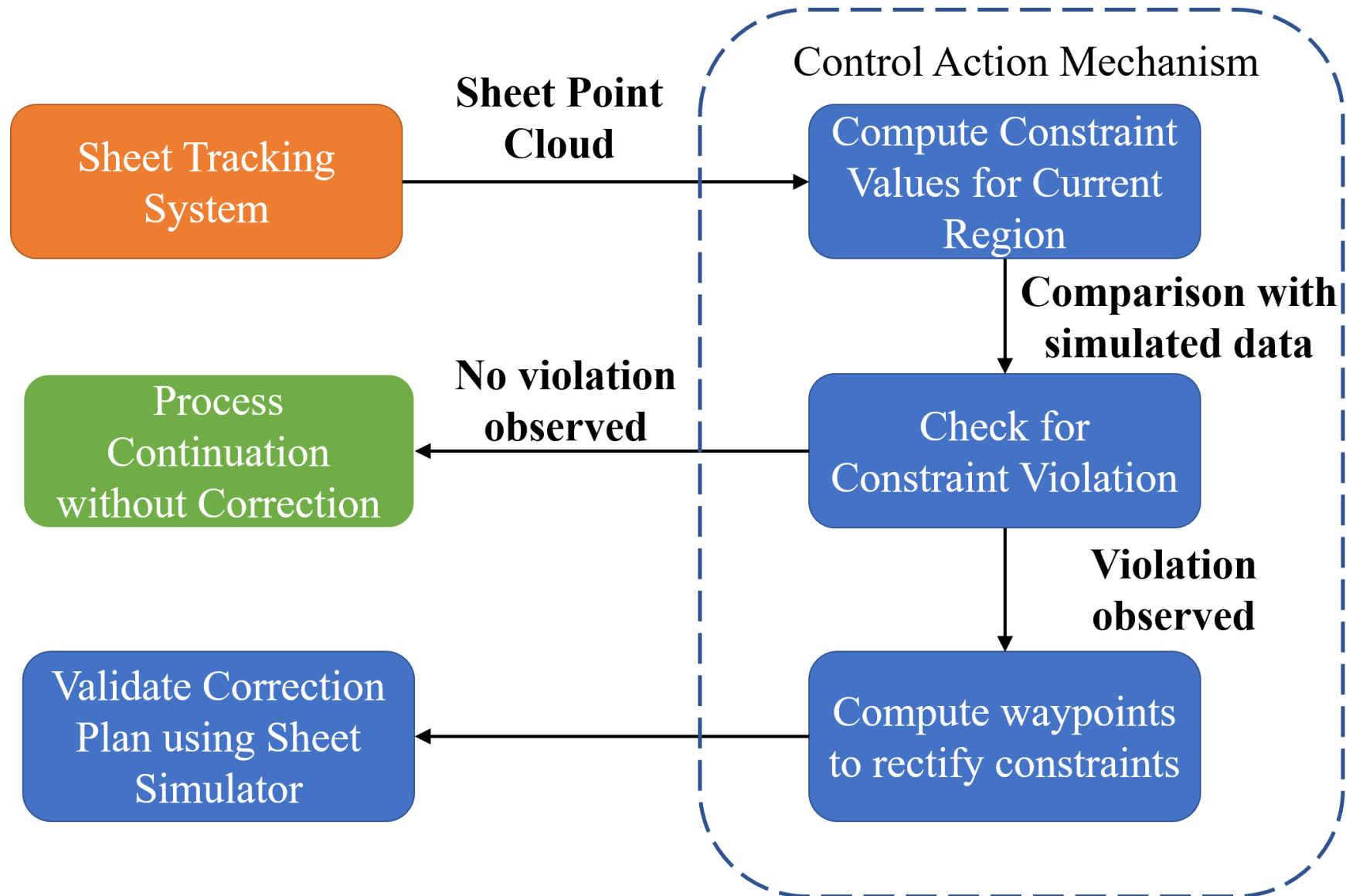
- Uncertainties in simulation model used for generating grasp plans can cause defects during robotic layup
- System uses multi-modal sensing based on vision and force feedback for on-line process monitoring
- AI based algorithms use the feedback to identify risky anomalous regions which can generate defects
- Analogous to hand layup, a pulling motion at the gripper reduces such anomalous regions



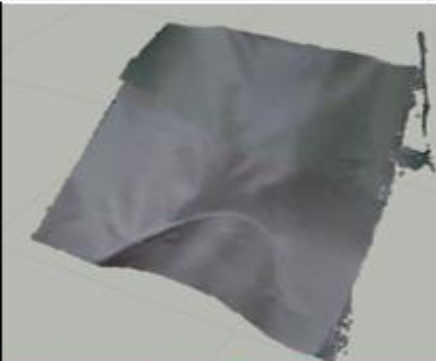

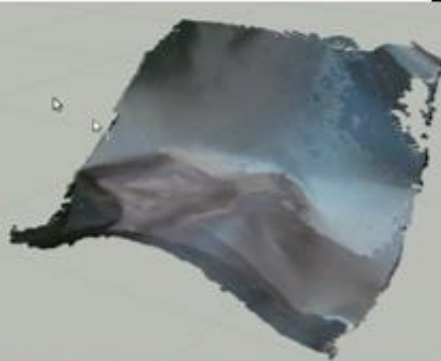





Intervention Controller

- Grasp plans may showcase some deviations from ideal scenario.
- Grasp planning system has inherent dependence on material model which maybe erroneous
- This entails for a need of an intervention system that can detect anomalous behavior and take appropriate action
- We utilize the sheet tracking system to evaluate grasping locations in real-time and take corrective actions

Intervention Controller: Basic Idea

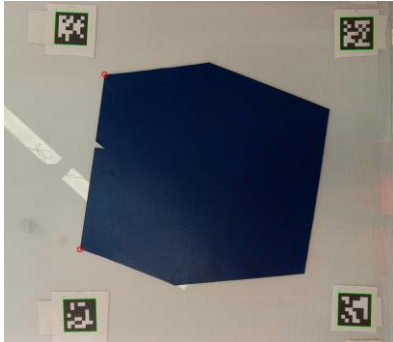


Results

	Case 1	Case 2 A	Case 2 B	Case 3
Sheet Tracking				
Actual Sheet Configuration				
	No Deviation	Minor Deviation		Major Deviation

Visual Servo Based Sheet Transport

Sheet Transport in Prepreg Composite Layup



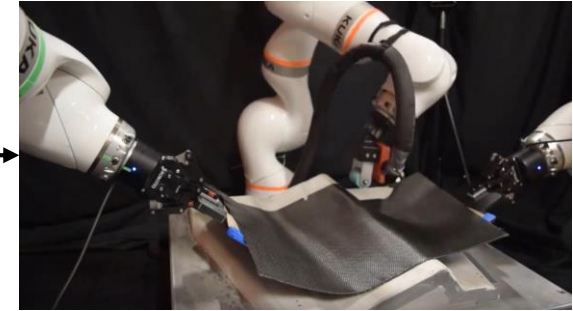
Sheet Loading
Station



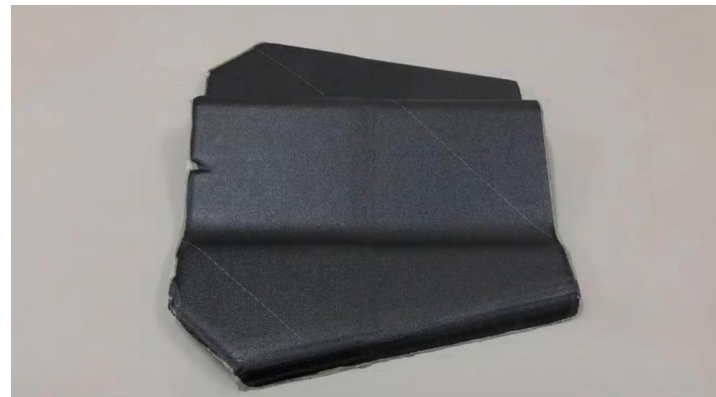
Sheet Pick and
Place Robot



Sheet Placed at
the Desired
Location



Sheet Draping
with Appropriate
Grasping



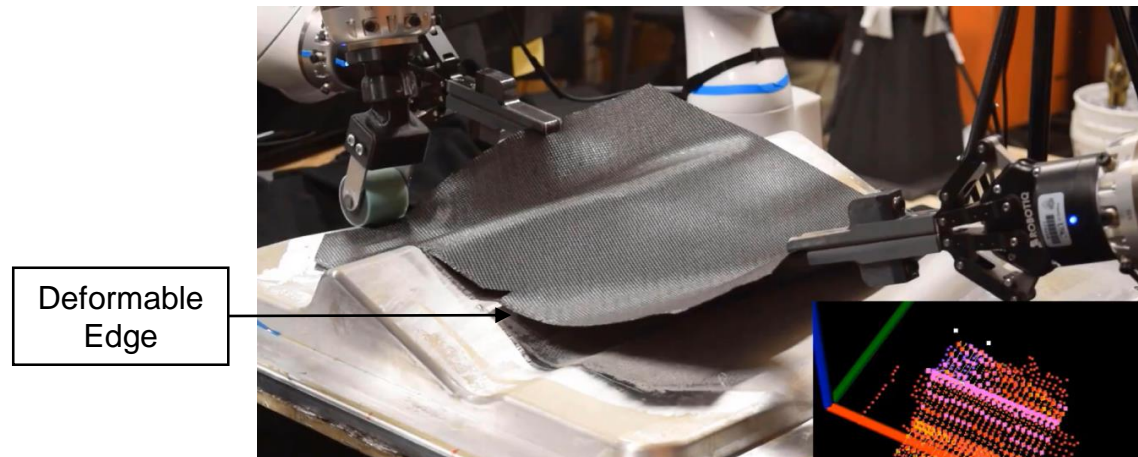
Final Cured Part

Challenges with Sheet Transport

- Deformable nature of the sheet can cause the sheet to assume different shape under suction-based grasping
- Deformable features are difficult to align with classical methods



Misalignment while using classical open loop vision-based techniques



Deformable features more difficult to align

Visual Servo a Potential Solution



Initial Misalignment of the
Feature

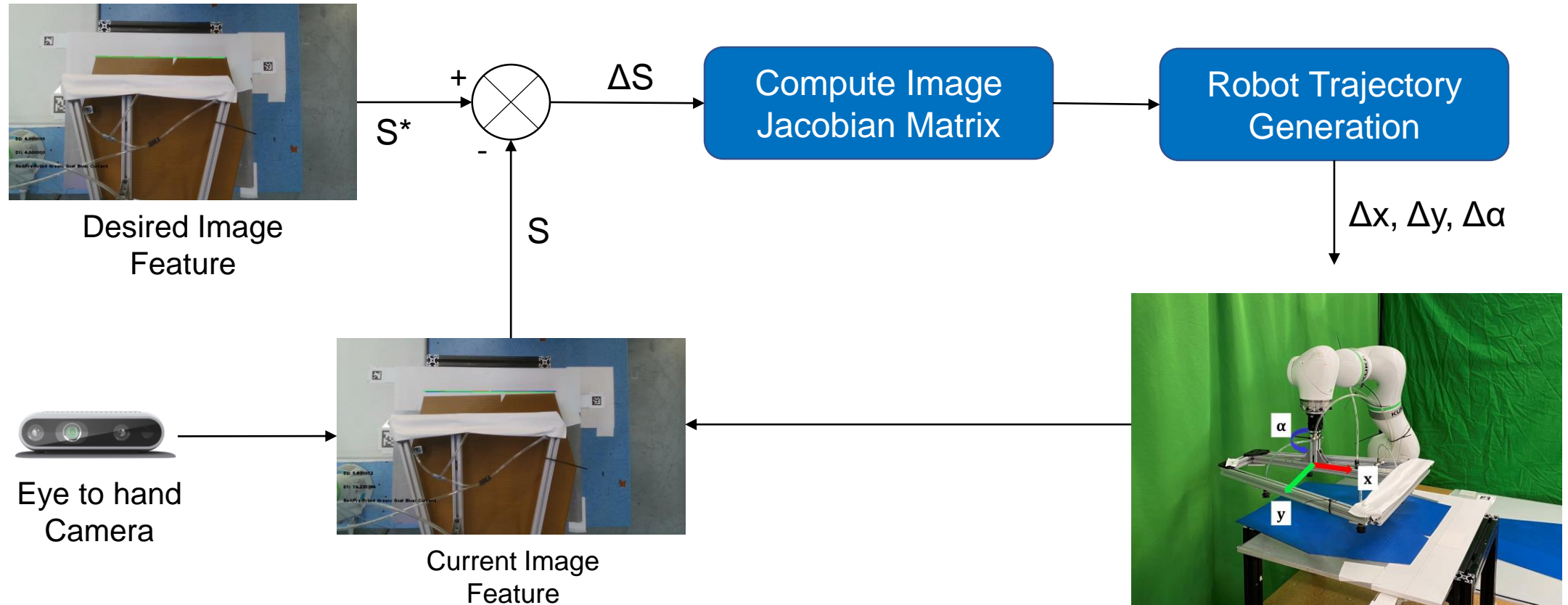
Compute
Trajectory



Final Aligned Feature

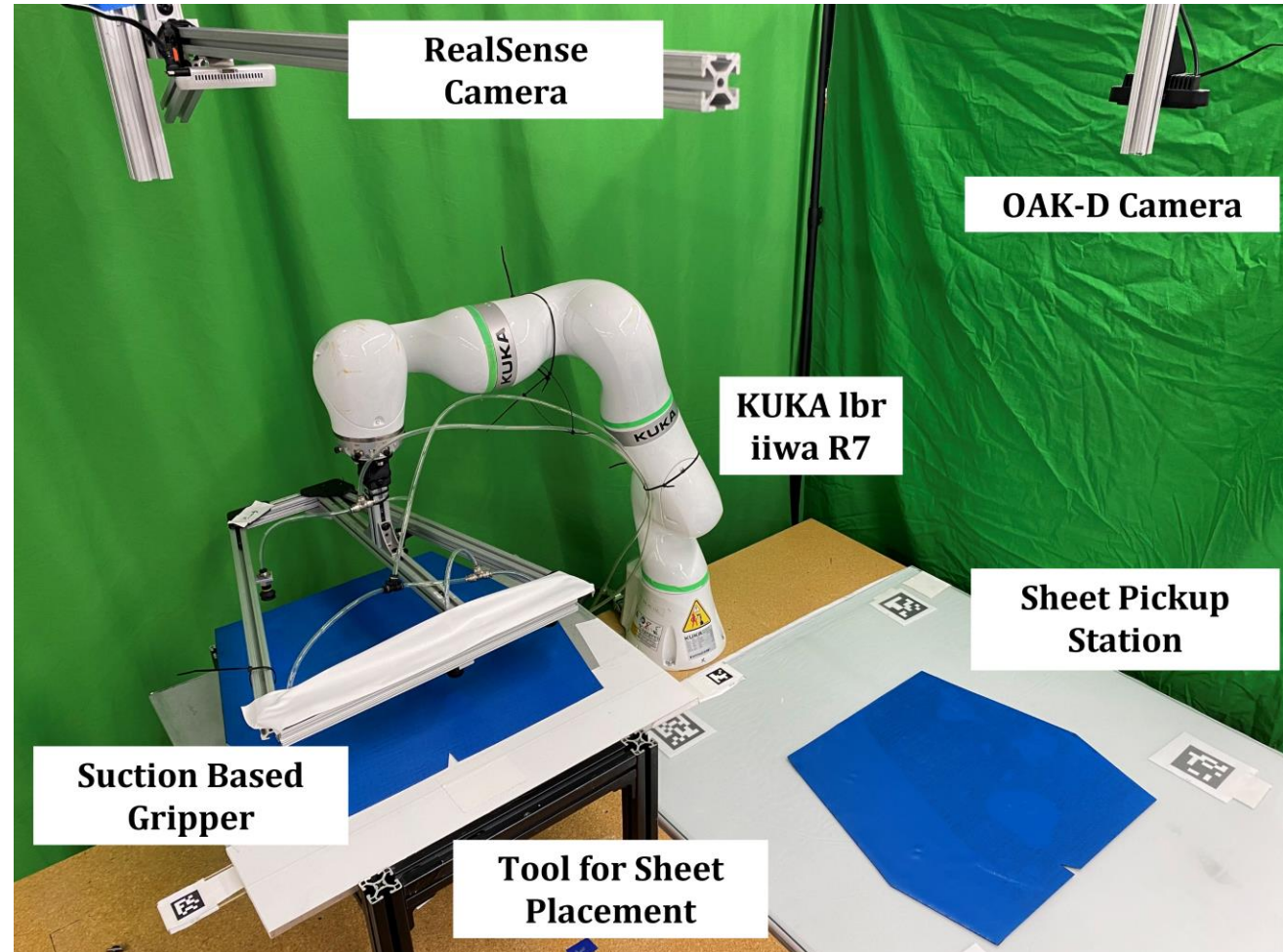
- Visual Servo is a classical method where the robot motion is controlled based on some alignment features in the image space
- Classical visual servo can be slow due to following reasons:
 - Slow convergence of the system
 - Feature recognition system
 - Estimation of Image Jacobian Matrix

Visual Servo Architecture

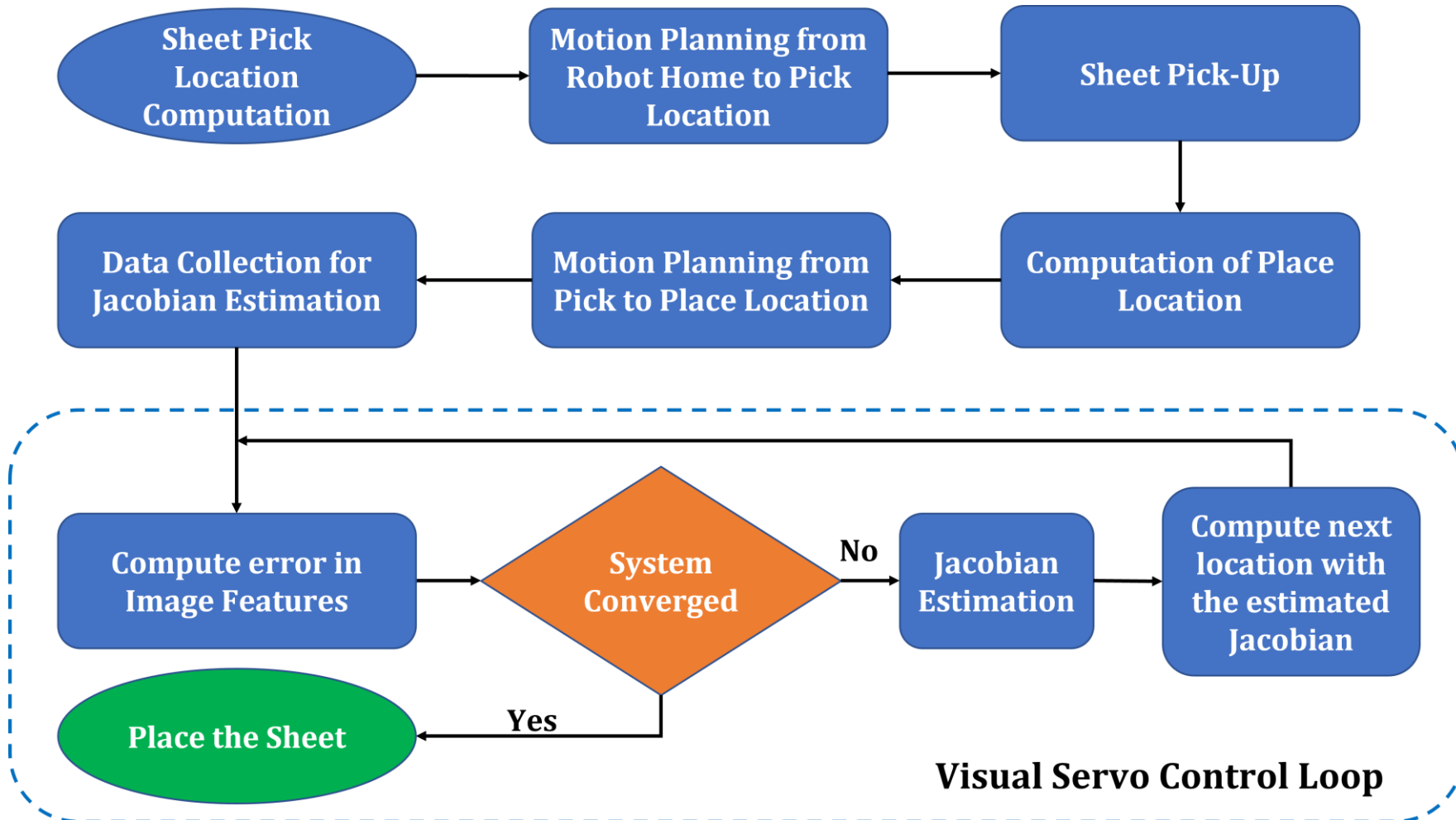


Cell Setup

- KUKA LBR Manipulator as Sheet Transport Camera
- RealSense and OAK-D cameras for vision capabilities
- Suction based tool for sheet pickup



Visual Servo Process Flow



System Overview

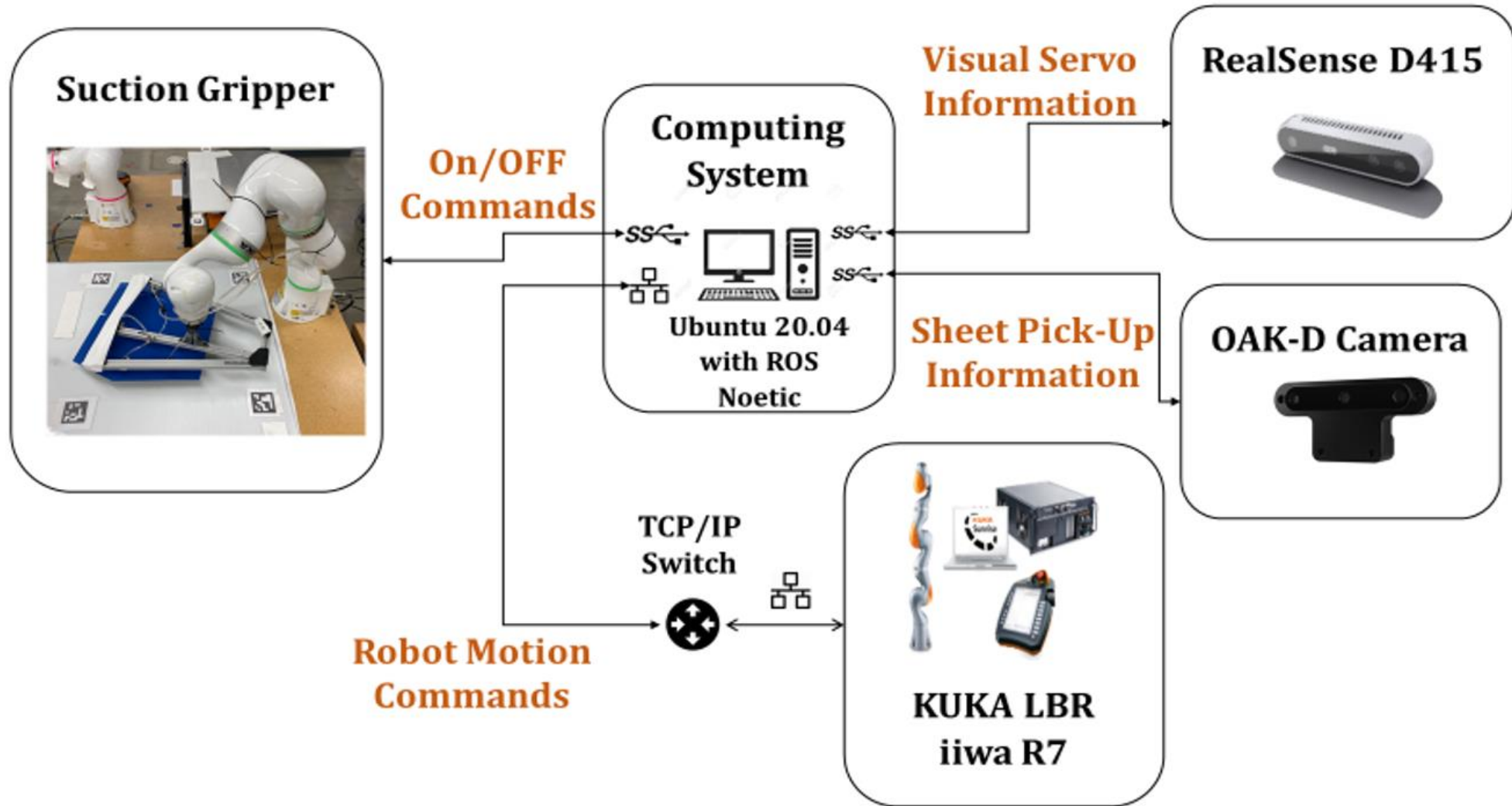
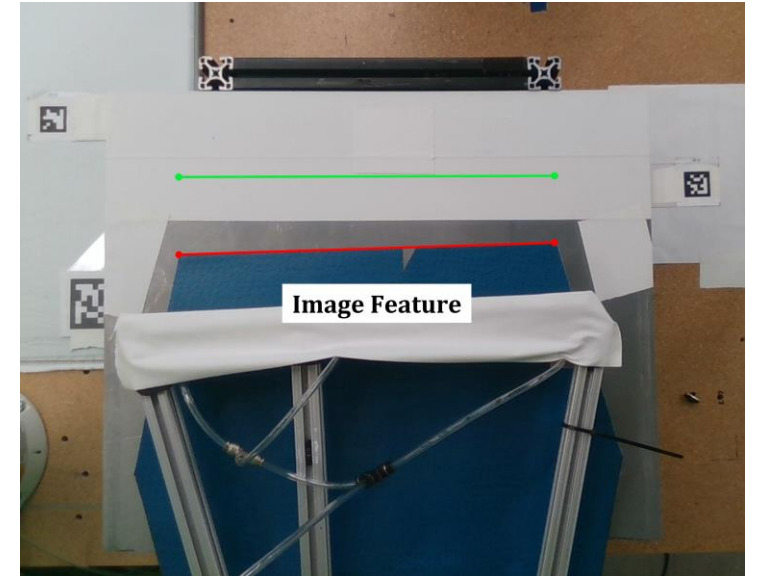


Image Feature Recognition

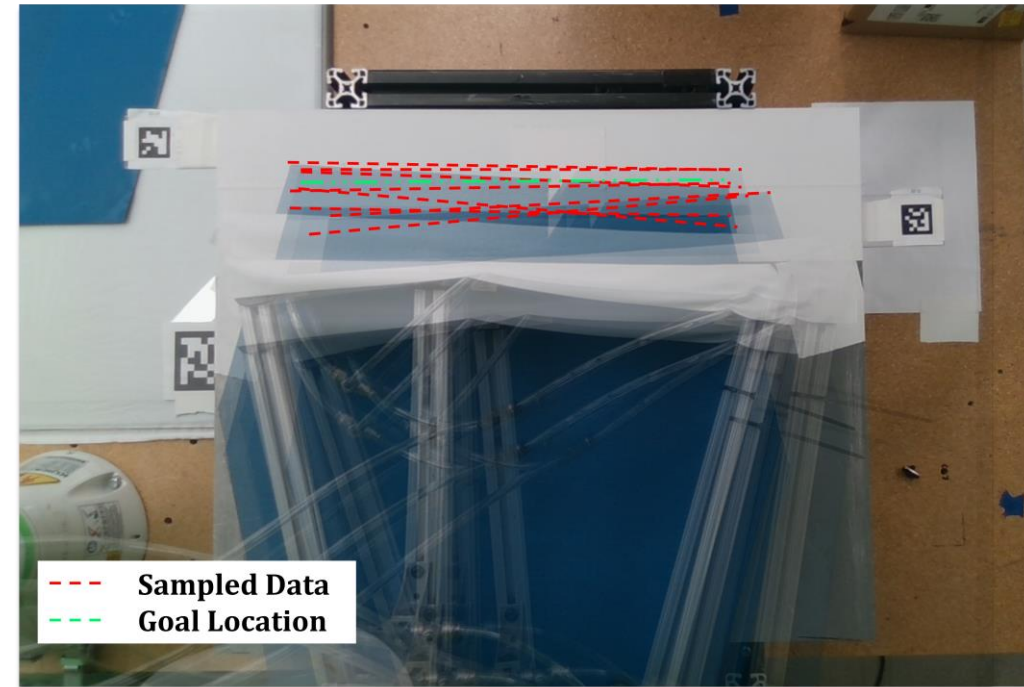
- Image feature is a crucial element of the visual servo control loop
- The objective is to choose a feature that is least affected by occlusions for easier detection
- The feature should also be unaffected by the deformability of the sheet
- The feature we chose was an edge of the sheet and the corners are the ones we track for feature alignment



Desired Image
Feature in Green

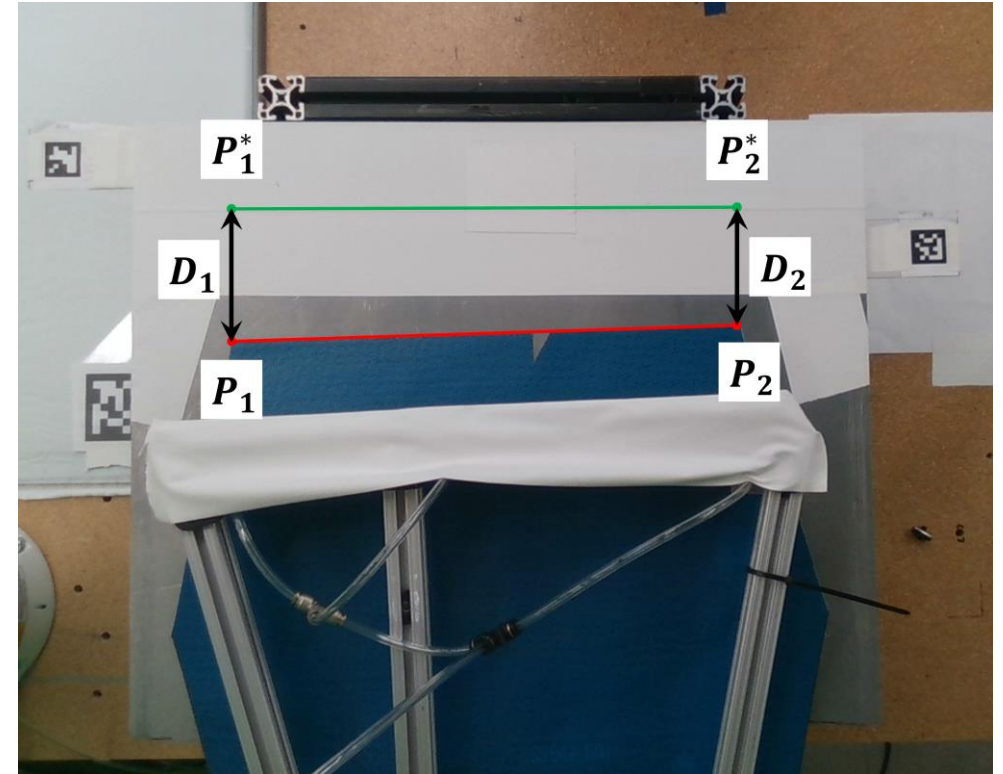
Image Jacobian Matrix Estimation

- In order to overcome the issues with the uncertainty in image feature due to sheet's deformable nature we adopt a sampling-based learning scheme
- We initially sample the space for varying values of cartesian poses and record corresponding pixel value
- We then use k-nearest neighbor search Algorithm to estimate the Image Jacobian Matrix



Trajectory Planning

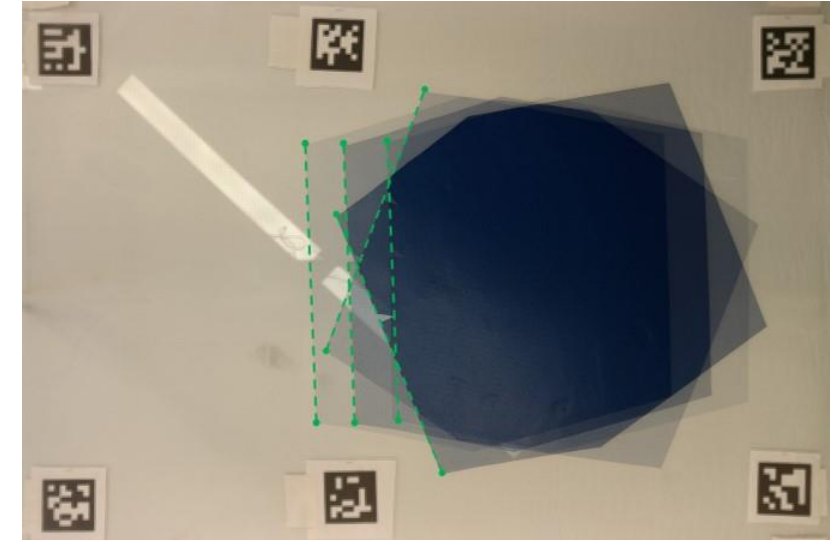
- The Image Jacobian is used to compute the robot trajectory such that the image feature error ΔS is reduced
- We run a BFGS based optimization technique to compute the path of reducing error
- The Robot was operated in Position Control mode for faster speeds and lesser vibrations of the tool



The error for the optimizer is proportional to the sum of D_1 and D_2

Experimental Results

- We conducted 50 trials for sheet pick and place
- Parameters Varied:
 - Sheet Loading Position ($\pm 10\text{ cm}$) and Orientation ($\pm 10^\circ$)
- The desired results are within the tolerance in observed in the industry for composite sheet placement applications

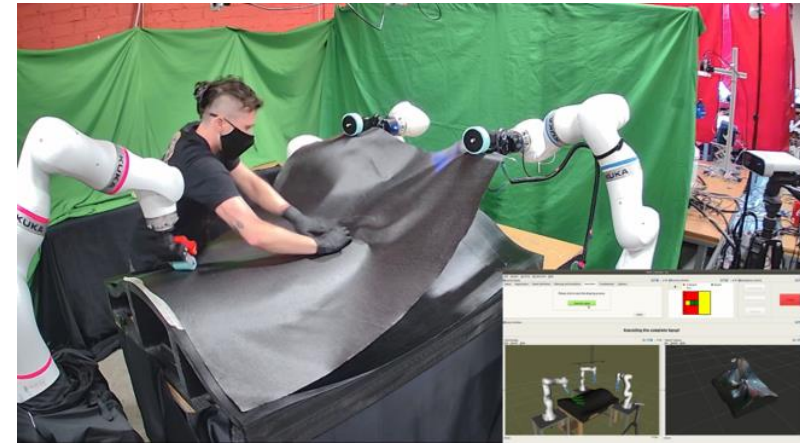
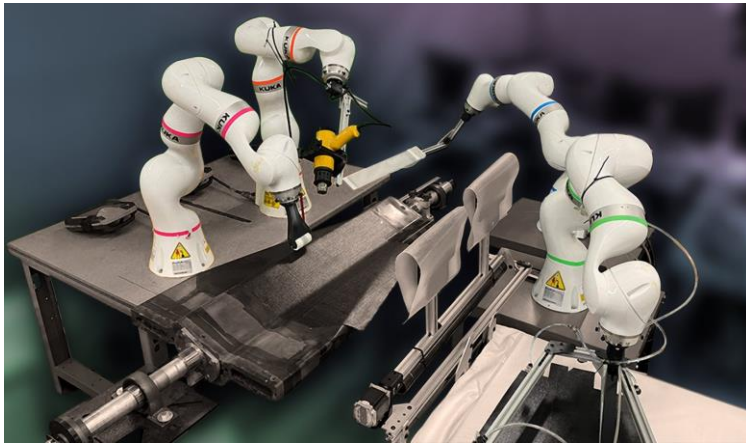


Variations in Sheet Loading Position

No of Trials	Position Error		Orientation Error		Image Jacobian Estimation Iterations	
	Avg	Max	Avg	Max	Avg	Max
50	0.8 mm	1.6 mm	0.2°	0.5°	2	4

Summary

- Composite prepreg layup is an important process in manufacturing high performance parts and will revolutionize the industry
- We have demonstrated how AI and machine learning tools can be deployed to achieve a smart robotic cell for automation of composite prepreg layup



Conclusions

- The advent of human-safe robots is enabling robots to work on ergonomically challenging tasks and amplify human productivity
- *Physics-Informed AI* is a key to realizing smart robotic assistants



Load Part



Task Robot



Let Robot to Do Work

Contributors



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Ying Cai



Jingsong Chu



Neel Dhanaraj



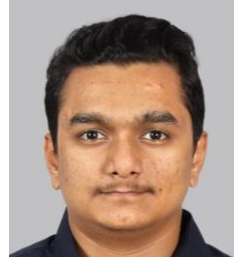
Liwei Fang



Ariyan Kabir



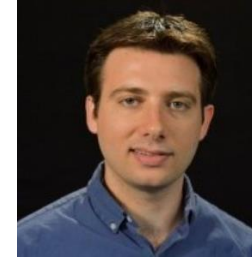
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Keng-Yu Lin



Rishi Malhan



Omey Manyar



Siddhant Nadkarni



Pradeep Rajendran



Dr. Brual Shah



Yash Shahapurkar



Aniruddha Shembekar



Shantanu Thakar



Yeo Jung Yoon

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Publications

- O. M. Manyar, J. Cheng, R. Levine, V. Krishnan, S. K. Gupta. Synthetic image assisted deep learning framework for detecting defects during composite sheet layup. *ASME IDETC-CIE*, St. Louis, MO, August 2022.
- O. M. Manyar, B. Deshkulkarni, A. Kanyuck, S. K. Gupta. Visual servo-based trajectory planning for fast and accurate sheet pick and place operations. *ASME Manufacturing Science and Engineering Conference*, West Lafayette, IN, June 2022
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- R.K. Malhan, A.V. Shembekar, R.J. Joseph, A.M. Kabir, P.M. Bhatt, B.C. Shah, W. Zhao, S. Nutt, and S.K. Gupta. A smart robotic cell for automating composite prepreg layup. *North America Society for the Advancement of Material and Process Engineering (SAMPE)*, May 2020
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