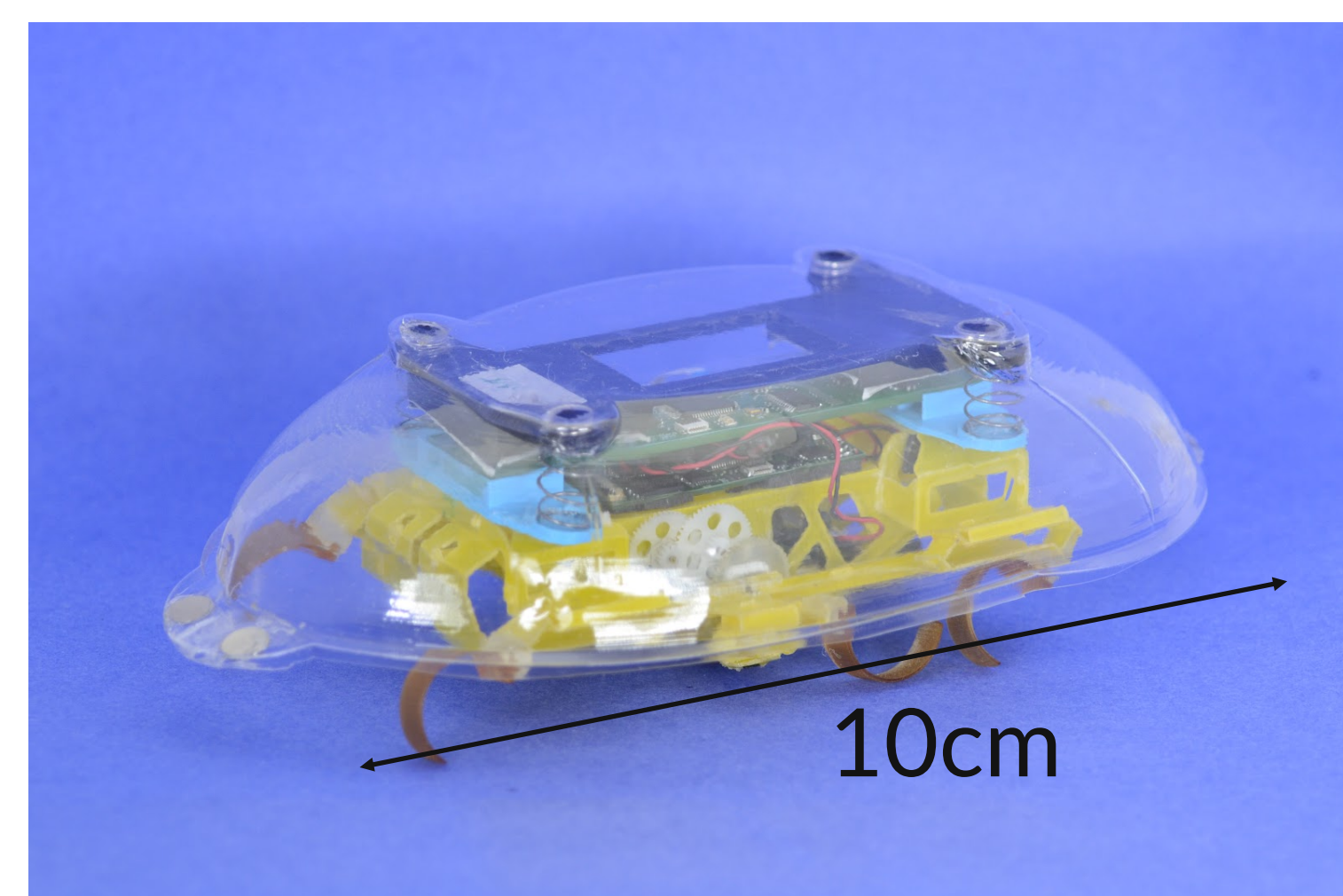




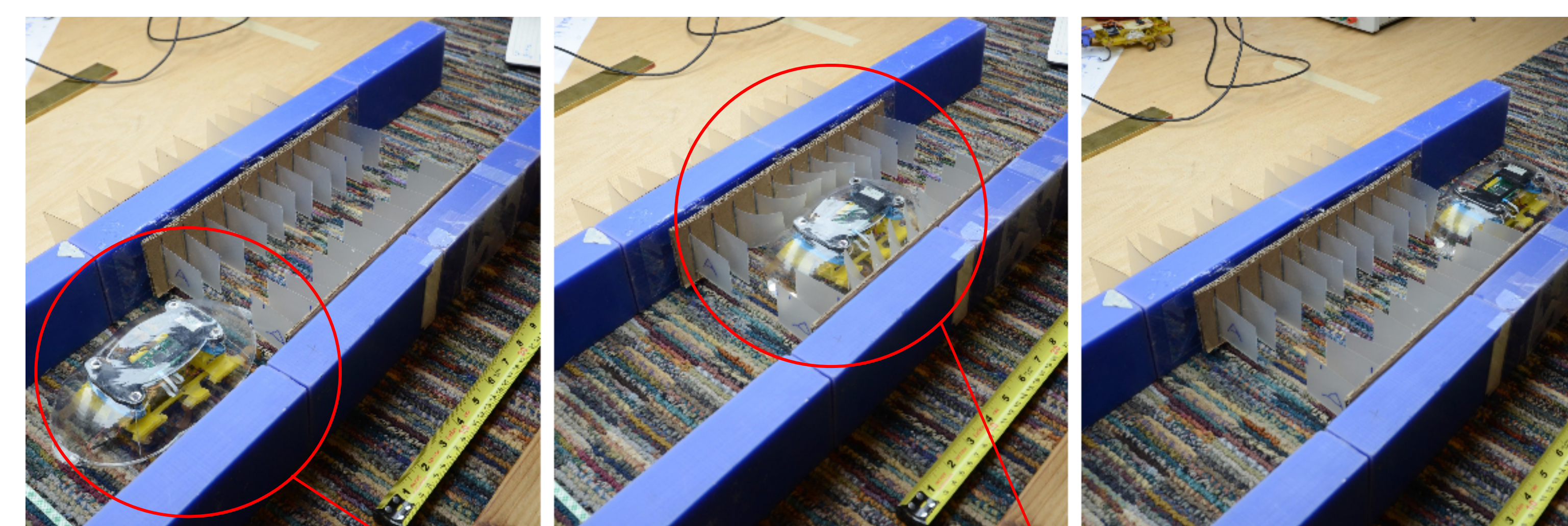
## Terrain Classification with Force-Torque Sensor Equipped Millirobot



Tactile robot in natural environment.



Robot model with tactile shell.



No Touch (Label: 0)

Touch (Label: 1)

### Objective

**Goal:**  
Distinguish low and high resistance regions.

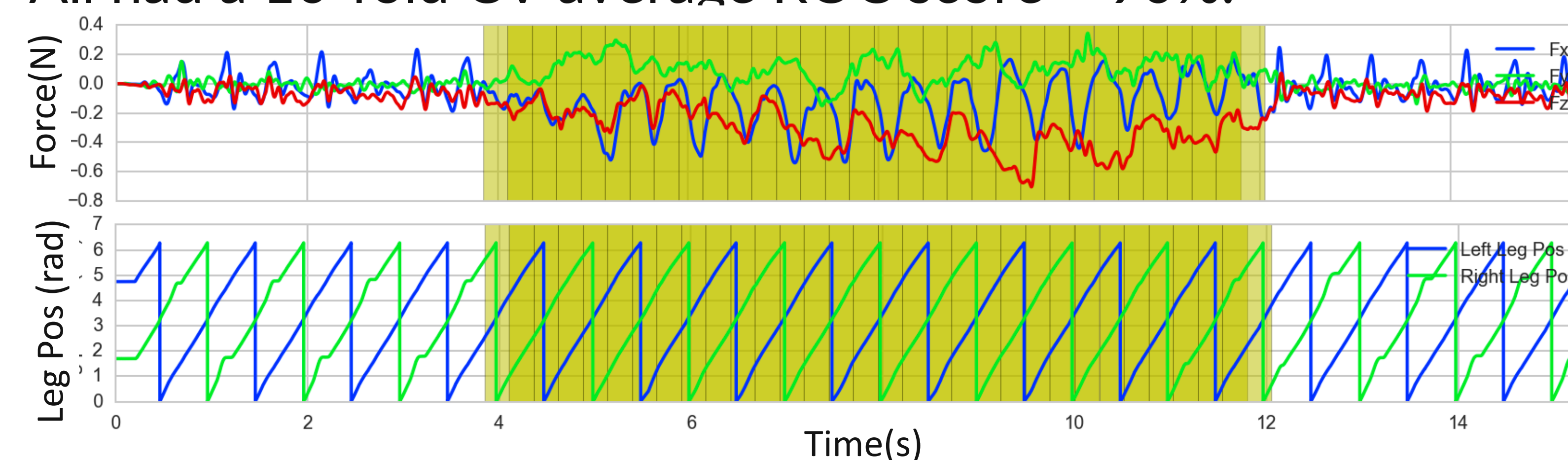
- Why?**
- Low vision environments, e.g. Dirt, moisture, dust.
  - Navigating through brush or dense compliant obstacles

### Method

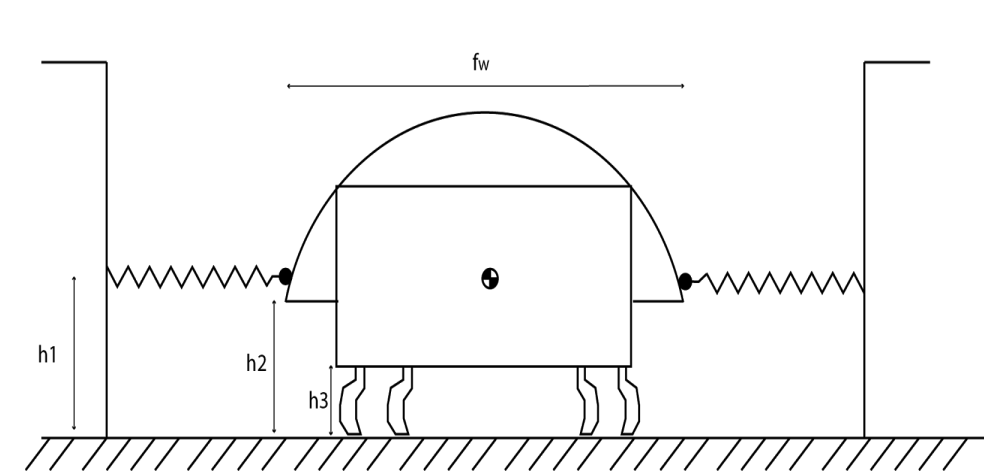
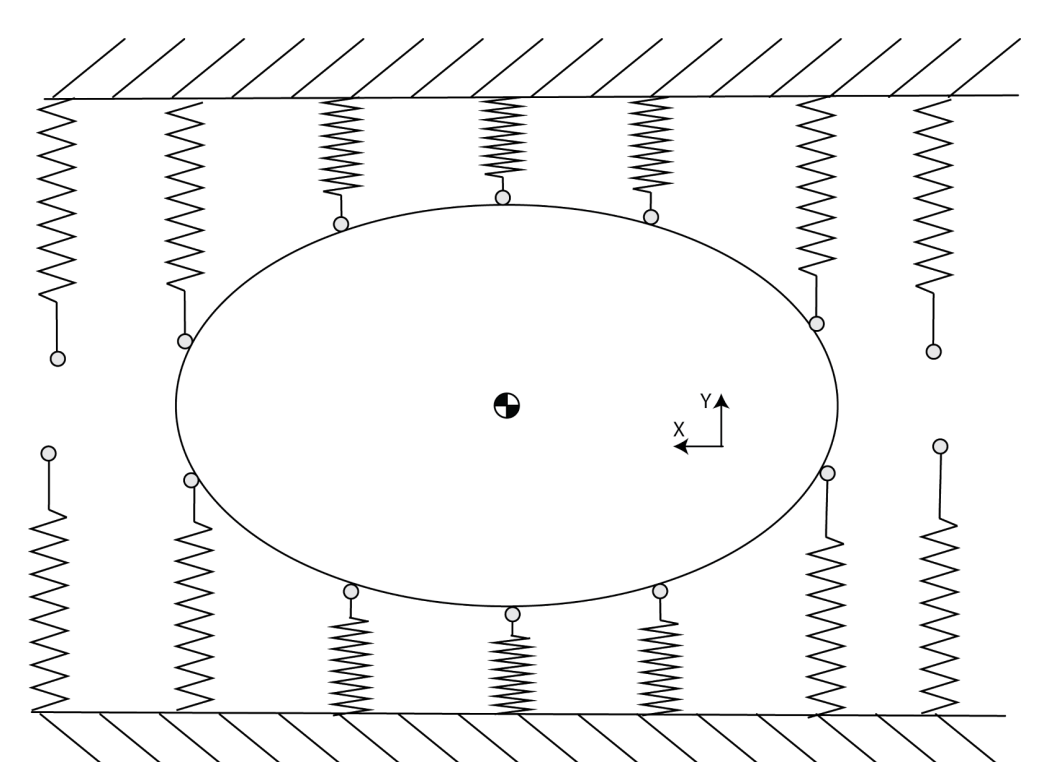
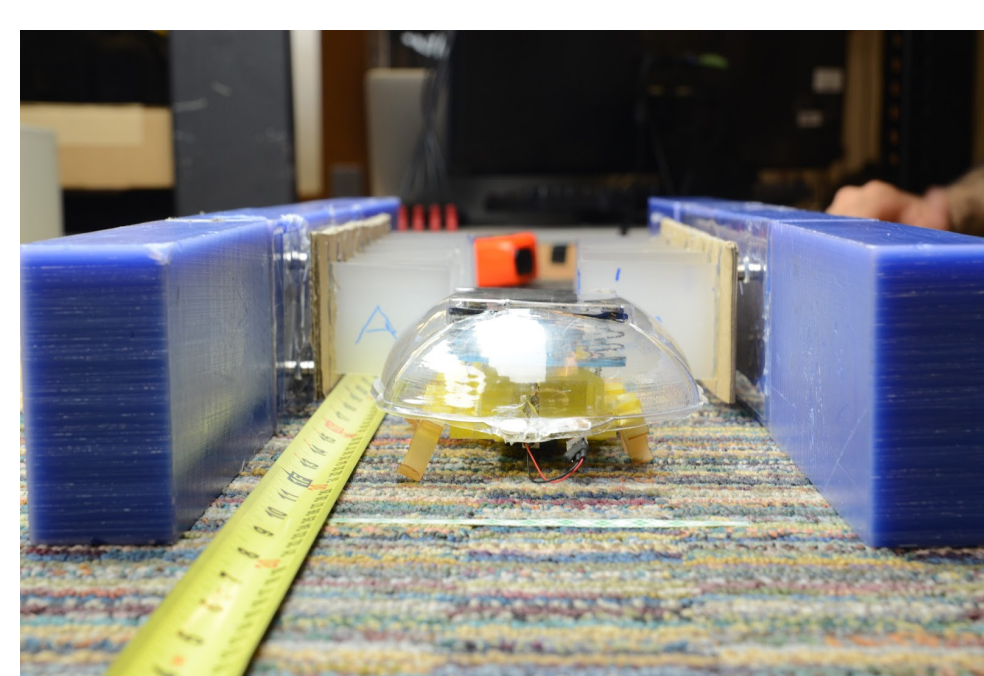
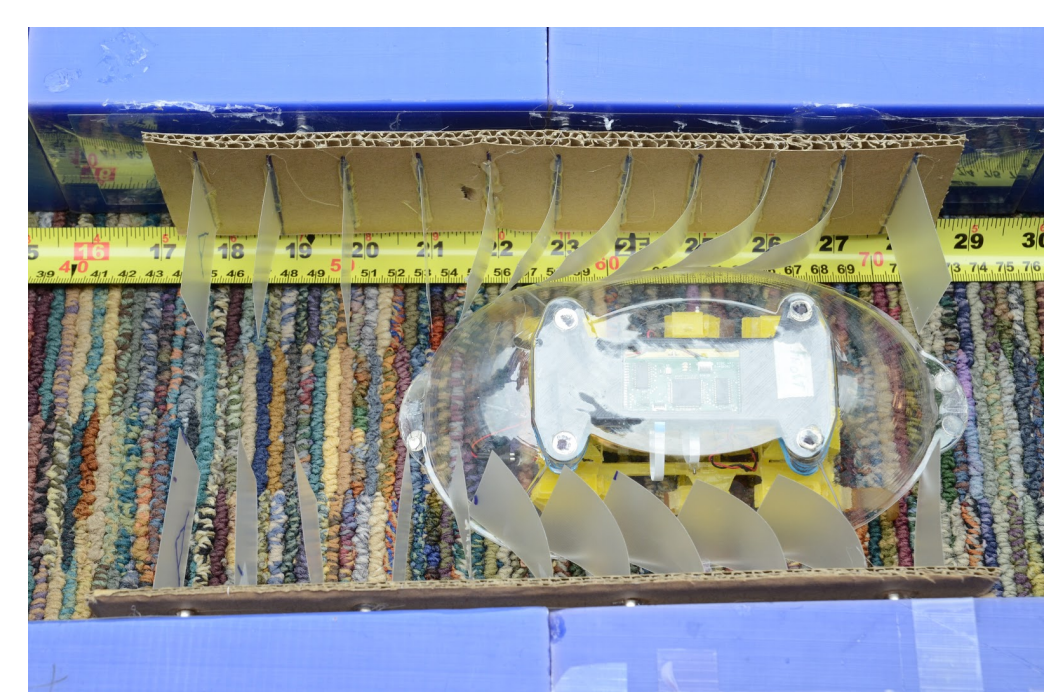
- Our experiment setup has 3 states: start (no touch), traversal (touch) and exit (no touch).
- Dataset consists of all the experiments from 4,6,8 and 10cm flap widths.
- After segmenting and applying windows we labeled the data as touch (1) or no touch (0).

### Results

- Input: shell forces,  $F_x$ ,  $F_y$ ,  $F_z$ , Gyro, Accelerometer, Leg Positions and Velocity.
- Trained random forests, gradient boosted trees, RBF SVM, and Neural Network
- All had a 10-fold CV average ROC score > 90%.



- Training Set: 772 samples (1) + 853 samples (0).
- Test Set: Single experiment with 4cm.



Classifier	10-fold CV avg ROC
Random Forests	97.4%
<b>Gradient Boosted Trees</b>	<b>97.6%</b>
RBF SVM	97.3%
Neural Network	92%

Gradient Trees Top Features	Random Forests Top Features
$F_z$ _max	$F_x$ _skew
$F_x$ _skew	$F$ _mag_energy
$F$ _mag_max	$F$ _mag_mean
$F_z$ _mean	$M_x$ _std
$M_y$ _min	$F$ _mag_max