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# Terrain Classification with Force-Torque Sensor Equipped Millirobot



Tactile robot in natural environment.

## Objective

#### Goal:

Distinguish low and high resistance regions.

#### Why?

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







Website: bml.eecs.berkeley.edu



Robot model with tactile shell. 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).

**Training Set:** 772 samples (1) + 853 samples (0).

Test Set:  $\bullet$ Single experiment with 4cm.



### Results

- and Velocity.
- Neural Network



#### Classifier

Random Forests

**Gradient Boosted** Trees **RBF SVM** Neural Network



No Touch (Label: 0)

Touch (Label: 1)

Input: shell forces, Fx, Fy,Fz, Gyro, Accelerometer, Leg Positions

Trained random forests, gradient boosted trees, RBF SVM, and

All had a 10-fold CV average ROC score > 90%.

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Image: Constraint of the state of		
4 6 10 12 14 Time(s)		
	<b>Gradient Trees Top</b>	<b>Random For</b>
10-fold CV avg ROC	Features	<b>Top Features</b>
<b>10-fold CV avg ROC</b> 97.4%	Features Fz_max	Top Features Fx_skew
<b>10-fold CV avg ROC</b> 97.4%	Features Fz_max Fx_skew	Top Features Fx_skew F_mag_ener
97.4% 97.6%	Features Fz_max Fx_skew F_mag_max	Top Features Fx_skew F_mag_ener F_mag_mea
<b>10-fold CV avg ROC</b> 97.4% <b>97.6%</b>	Features Fz_max Fx_skew F_mag_max Fz_mean	Top Features Fx_skew F_mag_ener F_mag_mea Mx_std

### YouTube: BiomimeticMillisys







