



# Tactile perception method development for a mobile walking robot

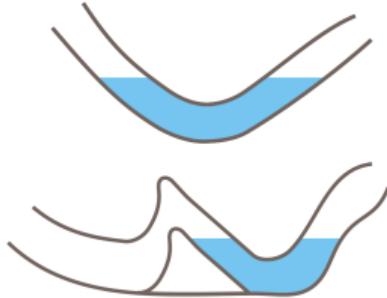
Student: Oleg Bulichev

Supervisor: Alexander Maloletov

# Motivation: why do we need to explore caves by robots



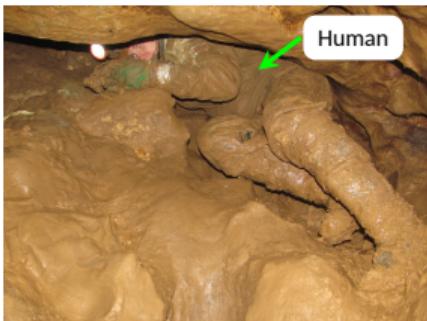
Salt



Siphon



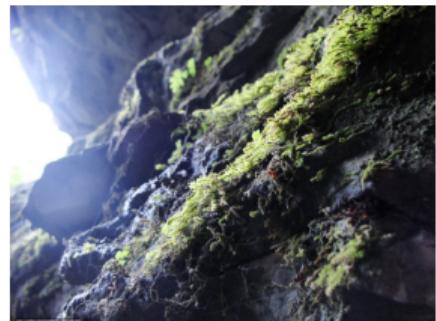
Glacier cave



Clay



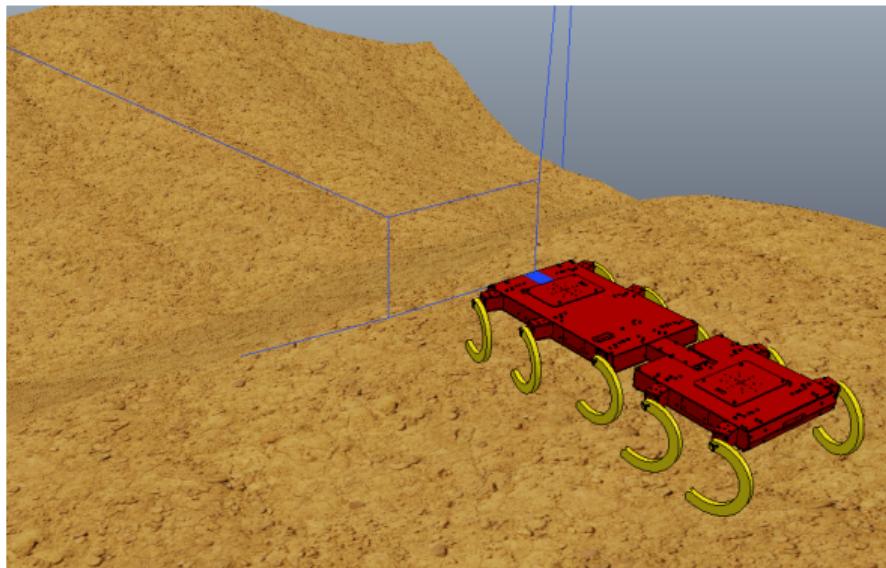
Puddle



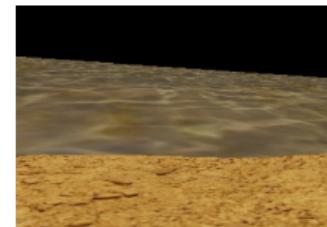
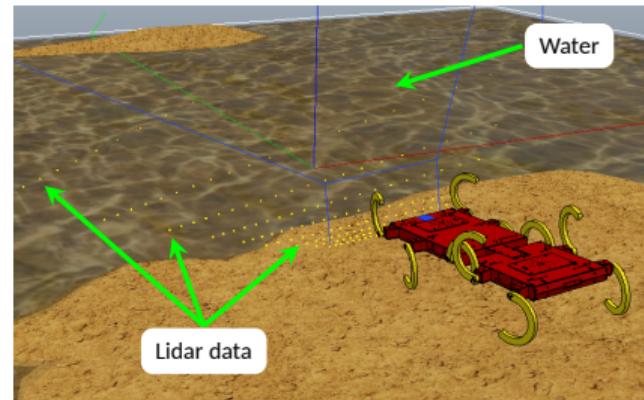
Moss

# Motivation: unsolvable problem for cameras and lidars

Question: how to make a map when you have a puddle above a surface?



Terrain without water



Camera view



# Problem statement

## Problem 1

How to obtain a useful information on terrain, **when we have a SLAM based on lidars, cameras?**



# Problem statement

## Problem 1

How to obtain a useful information on terrain, **when we have a SLAM based on lidars, cameras?**

Obtain map and terrain type



# Proposed solutions

## Problem 1

*Map can be built using tactile sensors on each leg of the robot and create a **dense point cloud** using sampling from generated mesh from **modified Delaunay triangulation**.*

*Terrain type can be obtained solving **terrain classification** problem using **machine learning**.*



# Literature review

Consider issues:

- Cave environment: obstacles, dimensions.
- Robots for cave exploration: from zeppelins, to quadruped robots.
- Methods for map creation: using classical and haptic. Based on cameras, lidars, tactile sensors, etc.

## Existing problems

- Robotics systems for exploring loose caves
- Object mesh creation using tactile sensor installed on manipulator
- Map creation using lidars and cameras



# Literature review

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## Existing problems

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New problem is proposed



# Robot design

## *Requirements*

**Problem** – choose robot mover type. This robot should:

- should have *small dimensions* to sneak through holes;
- have enough *off-road passability* to pass a granular terrain;
- should *traverse through small water obstacles*;
- can *climb on big stones*;



# Robot design

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Cycling robot with 1 DoF leg



# Robot design

*Structural synthesis problem*

## Question

What the optimal number of legs should be in such robot?



# Robot design

*Structural synthesis problem*

## Question

What the optimal number of legs should be in such robot?

## Answer

Robot should have **8-14 legs** in total!

# Robot design

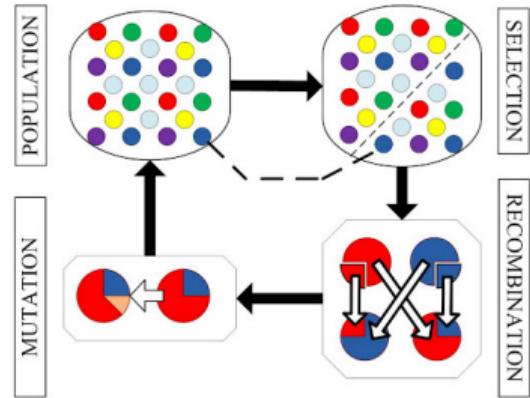
*Technological stack*



**Generating terrain approach  
(Robot traverse an *artificial terrain* based on *generating parameters*)**



**GAZEBO**  
*Robot simulator*



**Genetic algorithm  
(DEAP Library)**



# Robot design

## Assumptions

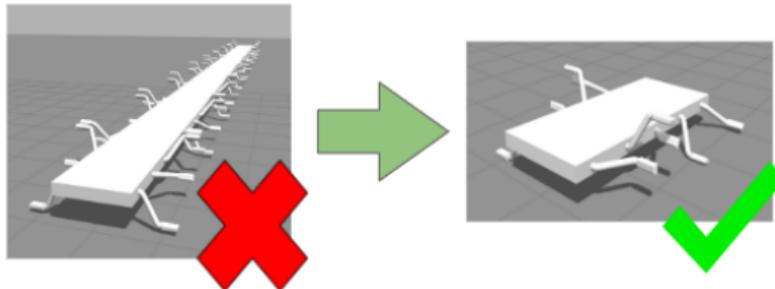
- Generated terrain family with the same constants has the same complexity.  
Parameters:

- Cell width and length
- Cell height range
- Distribution parameter

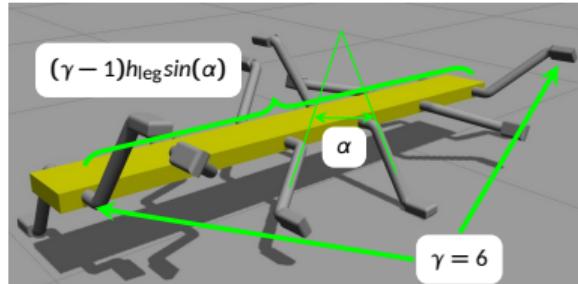


# Robot design

## Proposed solution



**Idea:** Minimize number of legs without losing off-road passability



$$F \rightarrow \max = \beta \left( \omega_1 \cdot \underbrace{\delta}_{\text{Distance}} + \omega_2 \cdot \frac{\overbrace{1}{\text{Simplified body length}}}{(\gamma - 1)h_{\text{leg}}\sin(\alpha)} \right) + \\ + (1 - \beta)\delta^{\omega_1} \left( \frac{1}{(\gamma - 1)h_{\text{leg}}\sin(\alpha)} \right)^{\omega_2}$$

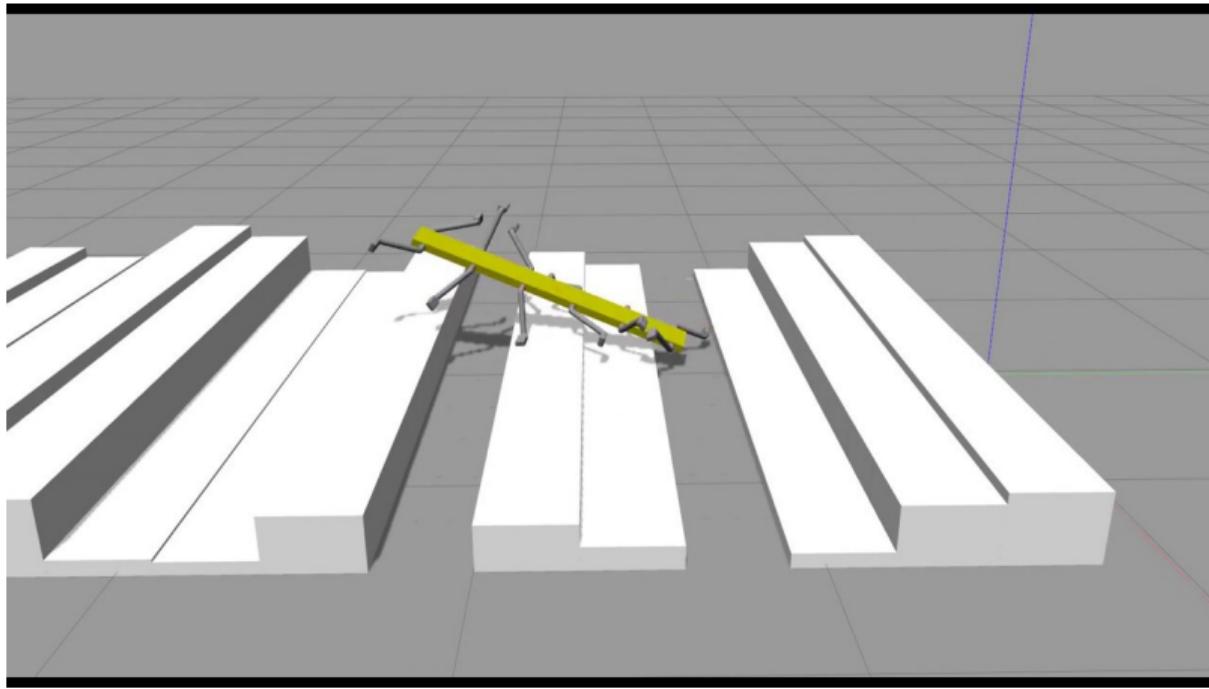
$\beta$  is adaptive parameter,

$\omega_{1,2} \in [0..1]$  are the weight coefficients.



# Robot design

*Video: the story of one generated robot*





# Robot design

Particular results:  $\omega_1 = 0.6$ ,  $\omega_2 = 0.4$

	Terrain types	No. Legs	Angle b/w neighbor legs	No. individuals
Phase 1		12	73	200
		12	72	
Phase 2		10	68	55
		12	77	

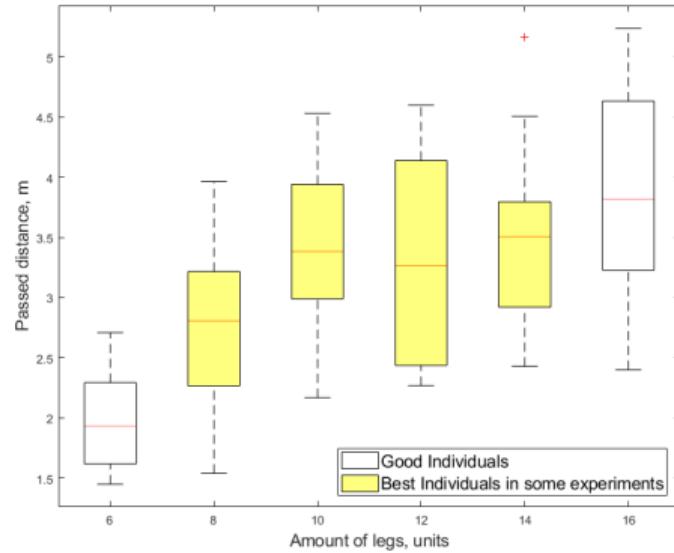


# Robot design

## Global results

Based on fitness function the number of legs range starts from 8 till 14 for different  $\omega$  values.

It can be explained by static stability criteria.  
In such case 4 legs will touch the ground.



*Correlation between amount of legs and passed distance by best robot individuals from several experiments*



# Robot design

*Increasing maneuverability*

## Question

1. Long robot can stuck in a cave hole, while rotating. How to avoid it?
2. How to climb on big stones?



# Robot design

*Increasing maneuverability*

## Question

1. Long robot can stuck in a cave hole, while rotating. How to avoid it?
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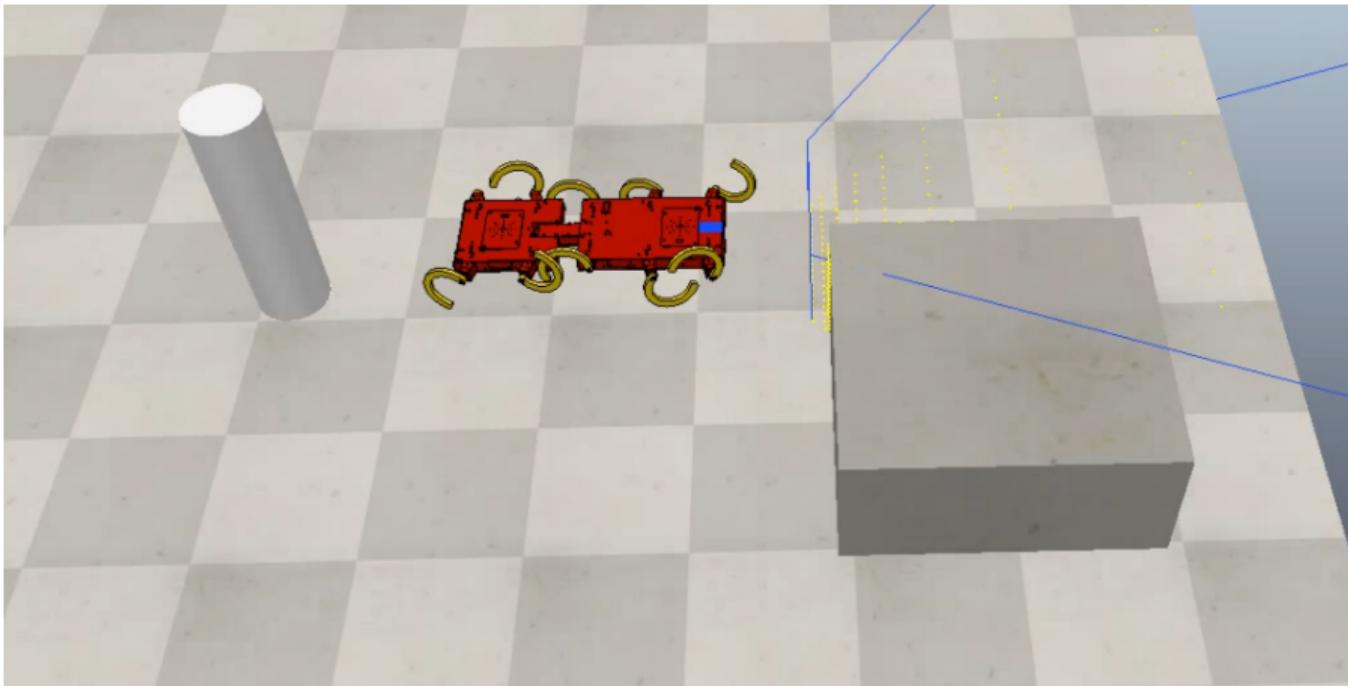
## Answer

1. Add an ability to sidestep without changing orientation.
2. Make a segmented body.



# Robot design

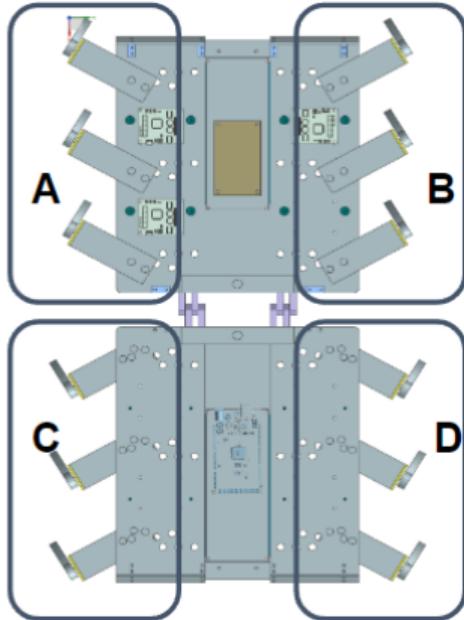
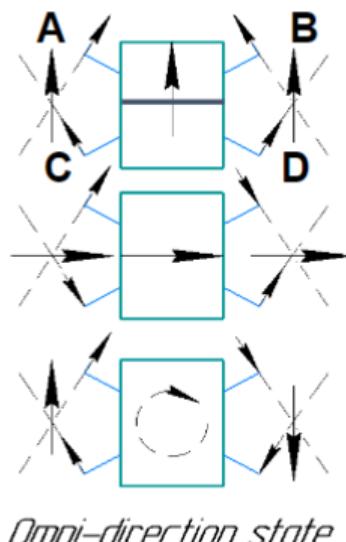
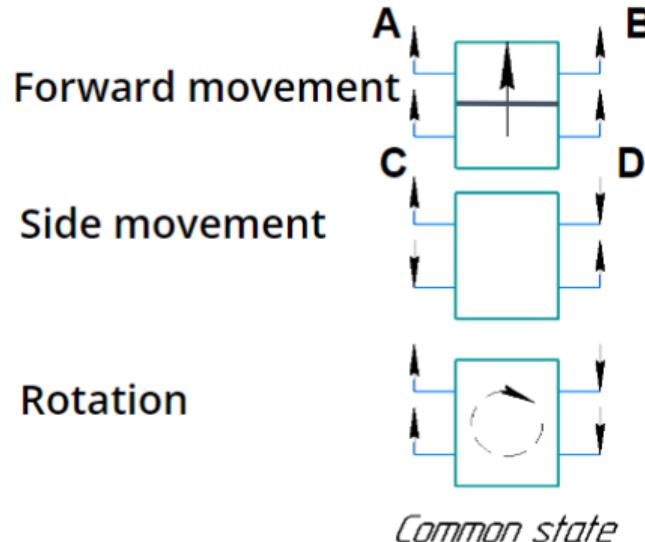
Video





# Robot design

*Proposed solution*

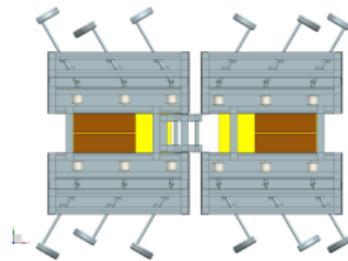
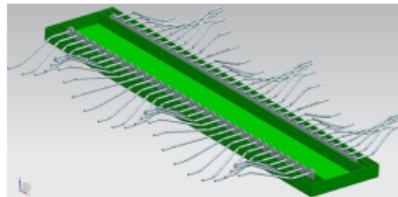


*Vector representation of forces in the conventional and omni-direction states*



# Robot design

## StriRus prototypes (1)

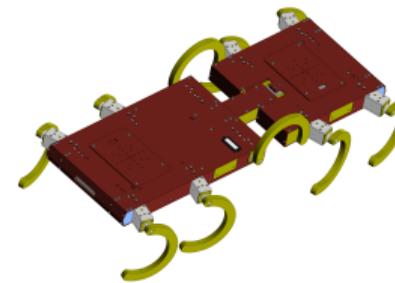
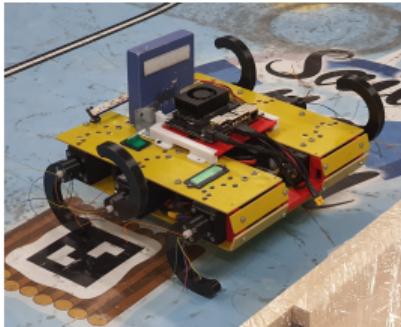


<b>Generation</b>	1	2	3
<b>No. legs</b>	54	12	12
<b>No. segments</b>	1	2	2
<b>Connect. joint</b>	—	Pitch	Pitch, Yaw
<b>Rel. body-leg angle, deg</b>	0	0-45	0, 15, 30, 45
<b>Leg height, mm</b>	54	60	60
<b>Features</b>	Wave propelled robot	Continuous mechanism	2 DoF connection joint mech.
<b>Withdraws</b>	<ul style="list-style-type: none"><li>- Cannot install force sensors</li><li>- high friction</li></ul>	<ul style="list-style-type: none"><li>- Too complicated changing relative angle mechanism</li></ul>	<ul style="list-style-type: none"><li>- Small legs</li><li>- Useless 2nd DoF in connection joint</li></ul>



# Robot design

## *StriRus prototypes (2)*



<b>Generation</b>	3+	4
<b>No. legs</b>	6	10
<b>No. segments</b>	1	2
<b>Connect. joint</b>	—	Pitch
<b>Rel. body-leg angle, deg</b>	0	0, 15
<b>Leg height, mm</b>	90	180
<b>Features</b>	Larger legs	Largest legs
<b>Withdraws</b>	- Still small legs - 1 segment	-



# Force transducer design

## Question

How to receive a reaction force from the ground?



# Force transducer design

## Question

How to receive a reaction force from the ground?

## Answer

- Current/voltage measurements from motors
- Installing torque sensor on a shaft
- Installing a force sensor on each leg



# Force transducer design

## Force sensor types

- **F/T sensors:** too massive and expensive for small robots.
- **Optical:** too thick.
- **Magnetic:** too thick.
- **Capacitive:** expensive, but the best in terms of requirements.
- **Piezoresistive sensors based on conductive inks or polymers:** inexpensive and robust, but has problems with hysteresis
- **Stain gages:** challenge for wiring between continuously rotating legs and the robot body.



# Force transducer design

## Force sensor types

- **F/T sensors:** too massive and expensive for small robots.
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- **Magnetic:** too thick.
- **Capacitive:** expensive, but the best in terms of requirements.
- **Piezoresistive sensors based on Velostat:** inexpensive and robust, but has problems with hysteresis
- **Stain gages:** challenge for wiring between continuously rotating legs and the robot body.

# Force transducer design

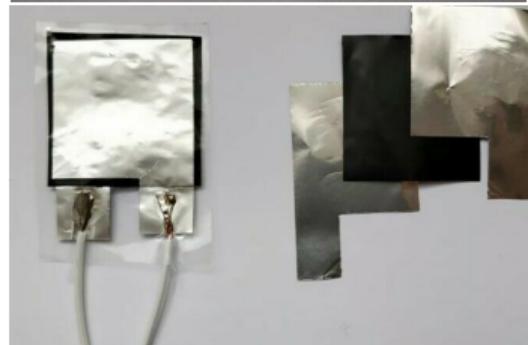
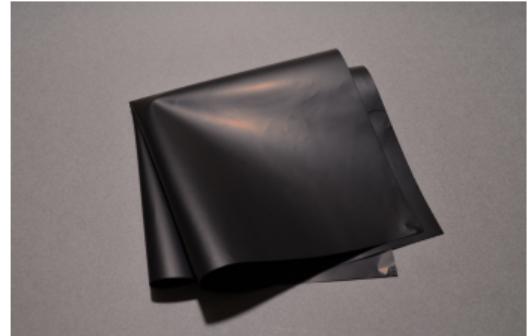
Velostat

## Definition

The Velostat is a polymer material filled with carbon black.

### Expected effects:

- Quantum tunnelling (Precipitation) – Diod is using such effect;
- Piezoresistive – electrical resistivity of semiconductor is changed by mechanical strain;
- Viscoelasticity – material can damp vibrations.



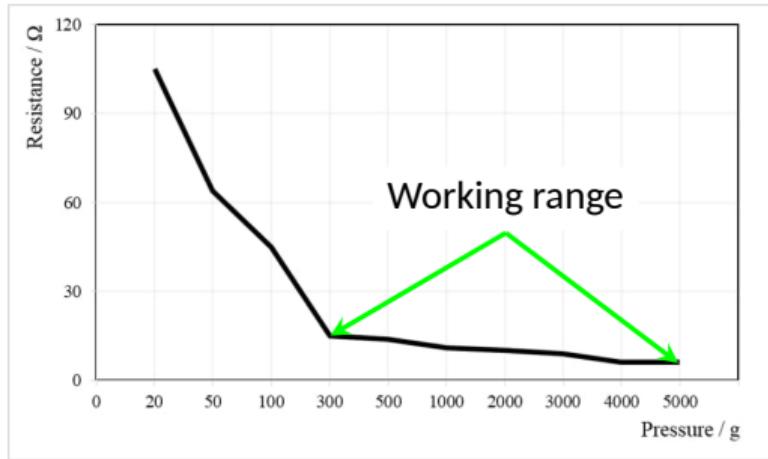
Simplest force transducer



# Force transducer design

*Velostat: Faced problems*

- Hysteresis
- Nonlinearity
- Different values with the same pressure if the square of load is less than the sensor



## Scientific Problem Statement

To characterize Velostat material for cases when point load is less than sensor size and propose solutions for avoiding such issues.



# Force transducer design

*Experimental setup requirements*

- Force control.
- Position and force repeatability.
- To have an ability to apply force only on a particular part of a sensor.



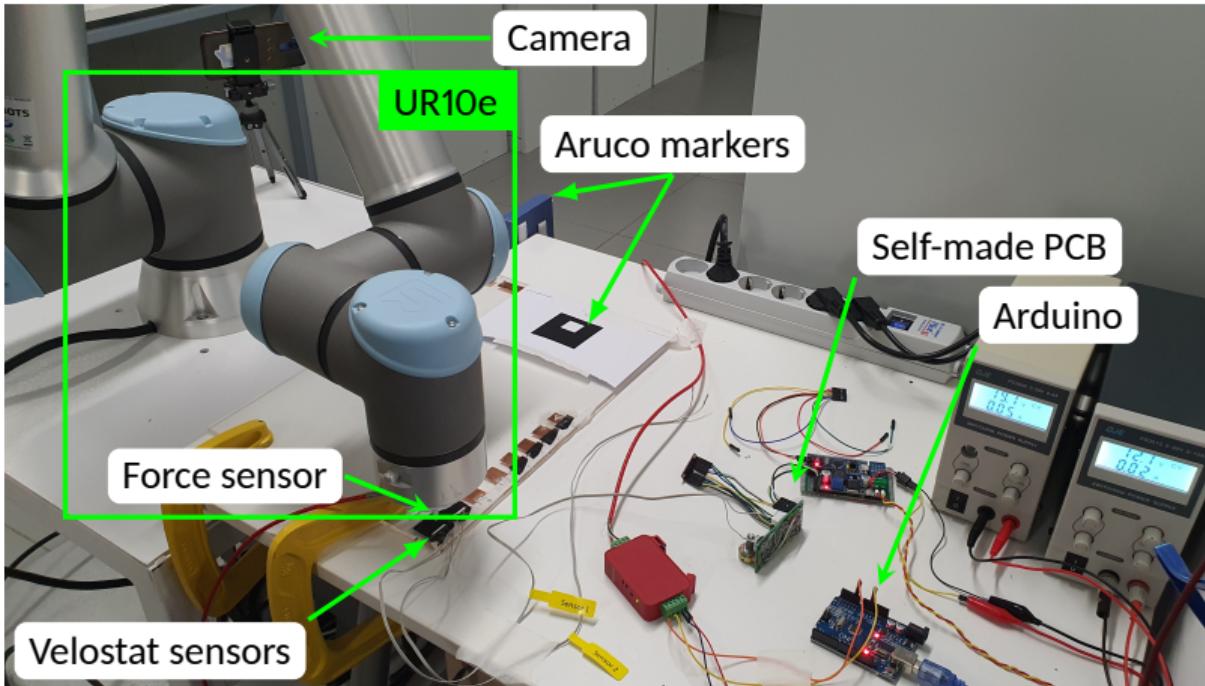
# Force transducer design

*Experimental setup requirements*

- Force control.  
*Solved by Impedance Control*
- Position and force repeatability.  
*Solved by adding a manipulator and a camera*
- To have an ability to apply force only on a particular part of a sensor.  
*Solved by adding several end-effectors*  
  
*All setup requirements are fulfilled*

# Force transducer design

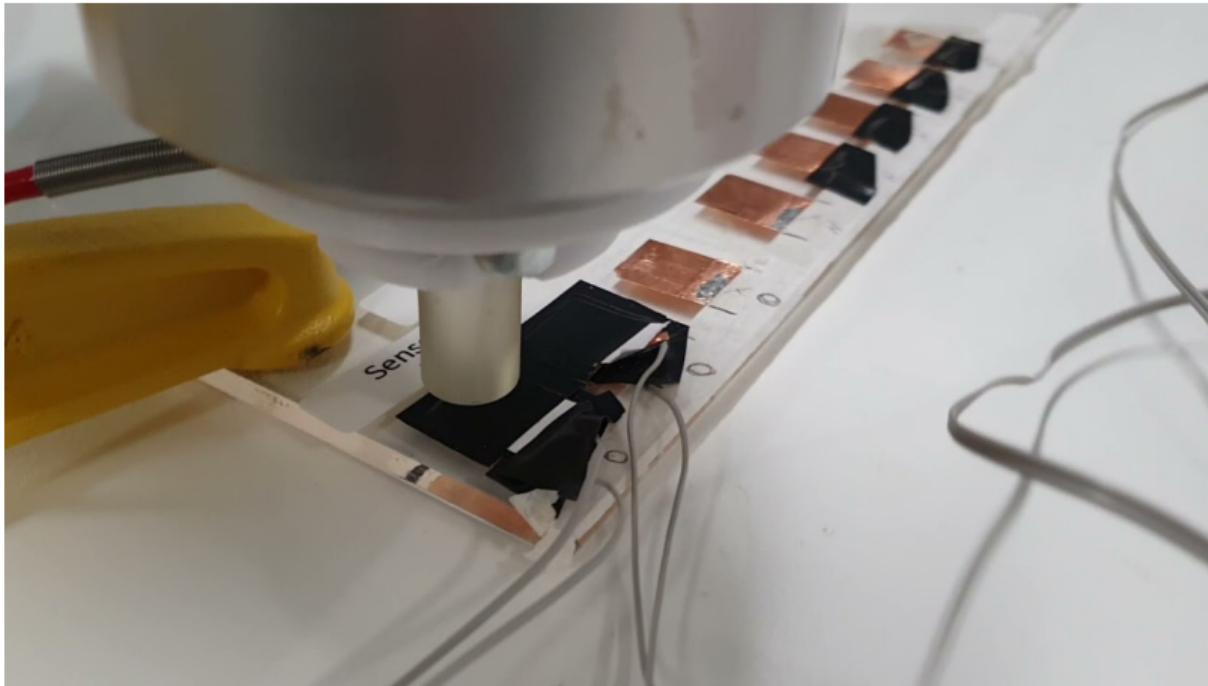
*Experimental Setup: Hardware overall*





# Experiment Design

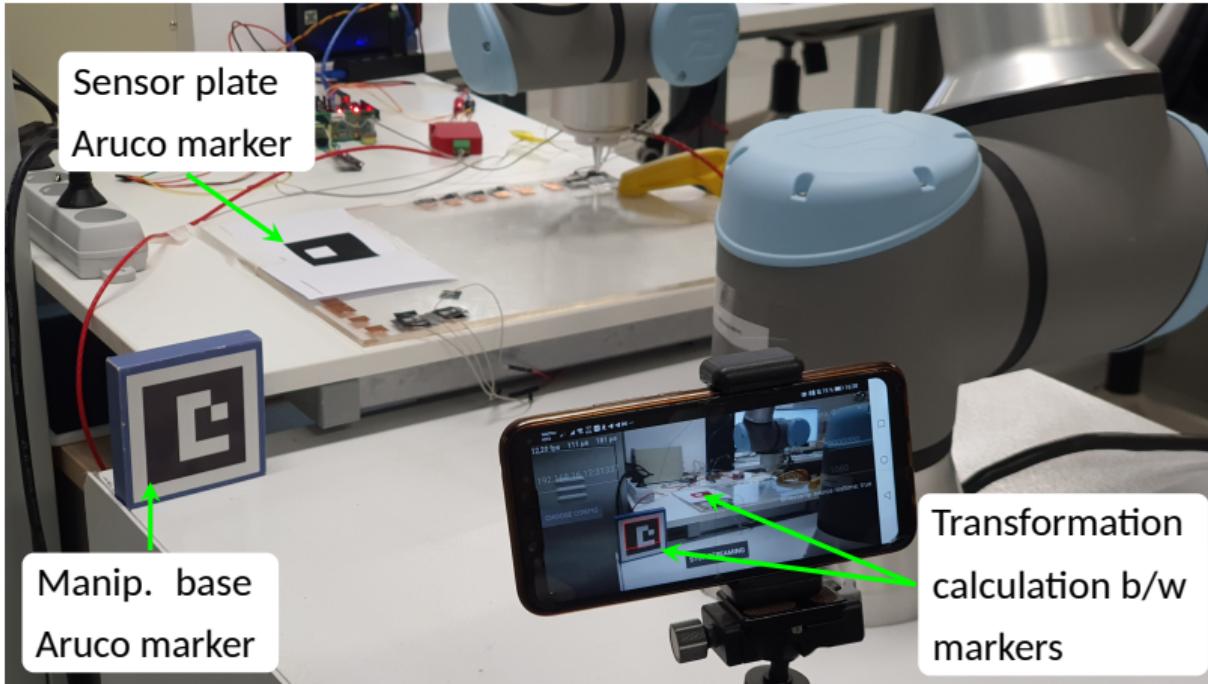
*Experimental Setup: Hardware, video*





# Force transducer design

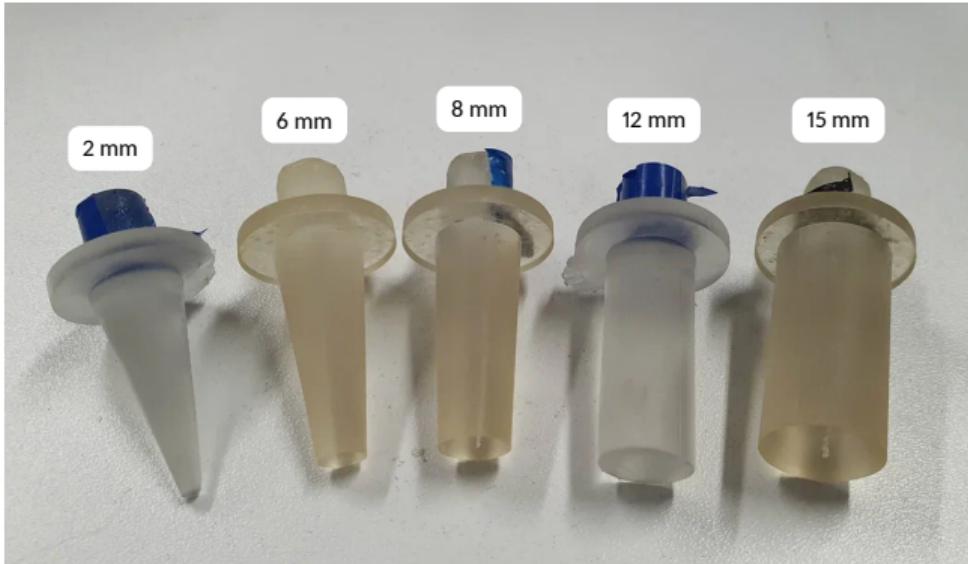
*Experimental Setup: Hardware, Aruco markers*





# Force transducer design

*Experimental Setup: Hardware, end-effector*



*All end-effectors*



*Ground Truth  
Force sensor*

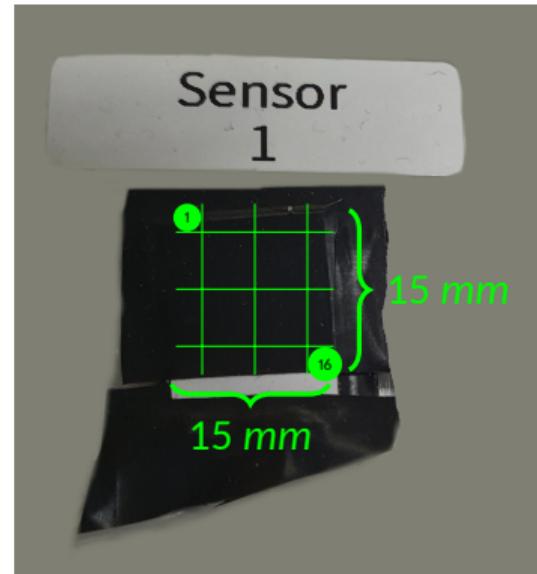


*End-effector in  
assembly*

# Force transducer design

## Experiments

1. **Static experiment.** The goal is to identify the coefficients for the transducer model.
2. **Dynamic experiment.**
  - We are representing a transducer as a  $4 \times 4$  grid. We touch with the same pressure using five different end-effectors (area starting from 2mm till 15mm).
  - We are using 2 mm and 15 mm end-effectors. We touch with different forces (5, 10, 20, 30, 40 H).



*Sensor representation  
as  $4 \times 4$  grid*



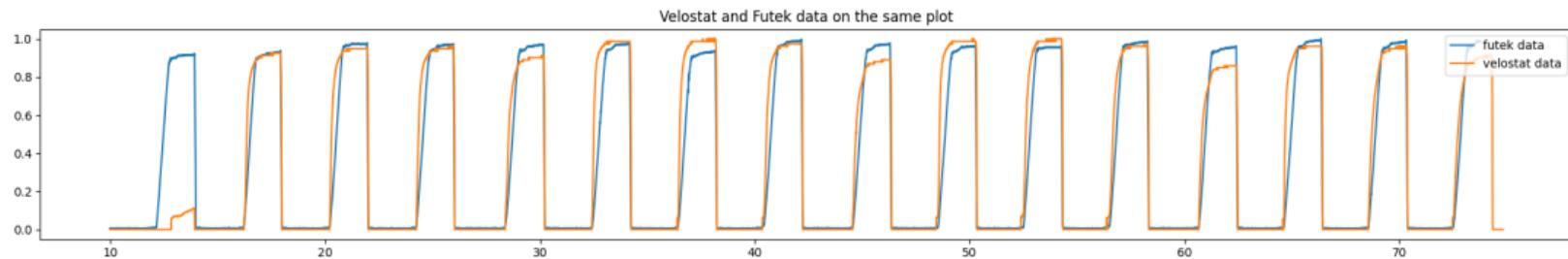
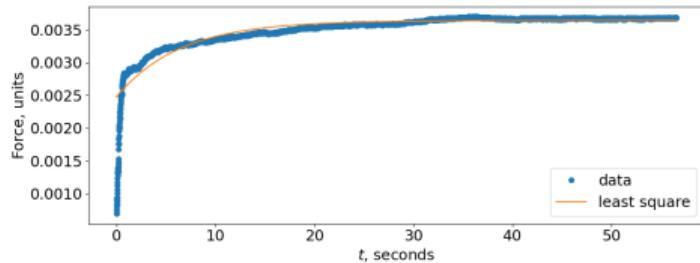
# Force transducer design

*Result: static experiment*

$$V_{out} = V_0 + p[k_p + k_e(1 - e^{\frac{-(t-t_0)}{\tau_{res}}})](1 - e^{-\frac{A}{p}})$$

$$k_p = A_1 e^{-A_2 p}; \tau_{res} = B_0 + B_1 e^{-\frac{p}{B_2}}$$

Where  $V_0$  - initial voltage,  $p$ ,  $A_i$ ,  $B_i$ ,  $\tau_{res}$ ,  $k_i$  are constants,  $t$  - current time,  $t_0$  - the time when the pressure appeared.

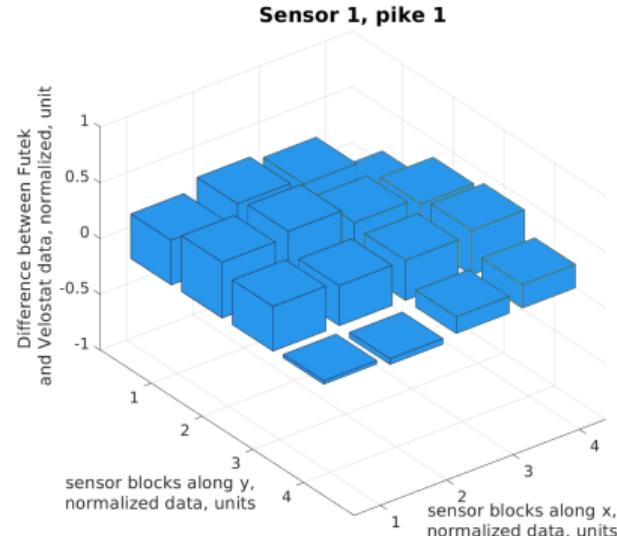


Normalized force data from sensor and transducer

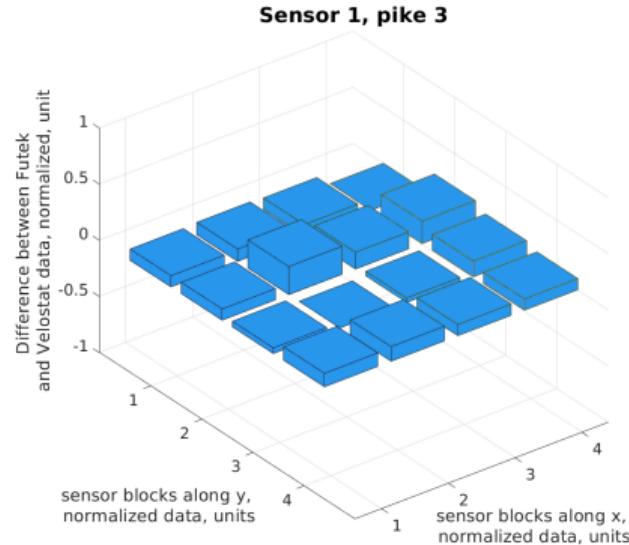


# Force transducer design

*Result: dynamic experiment*



*2mm end-effector diam*



*8mm end-effector diam*



# Force transducer design

## Summary

1. Static experiment: transducers coefficients  $p$ ,  $A_i$ ,  $B_i$ ,  $\tau_{res}$ ,  $k_i$  were identified.
2. Dynamic experiment: a transducer can be represented as a single body, when pressure area is higher than 50% of the sensor area.



# Terrain classification

## Question

How to define the terrain type during the movement on such terrain?



# Terrain classification

## Question

How to define the terrain type during the movement on such terrain?

## Answer

Solving terrain classification problem using Machine learning



# Terrain classification

*Experimental setup requirements*

- To have a possibility to install new surfaces and change it quickly.
- Infinite movement.
- Movement part should be the same as on the StriRus.



# Terrain classification

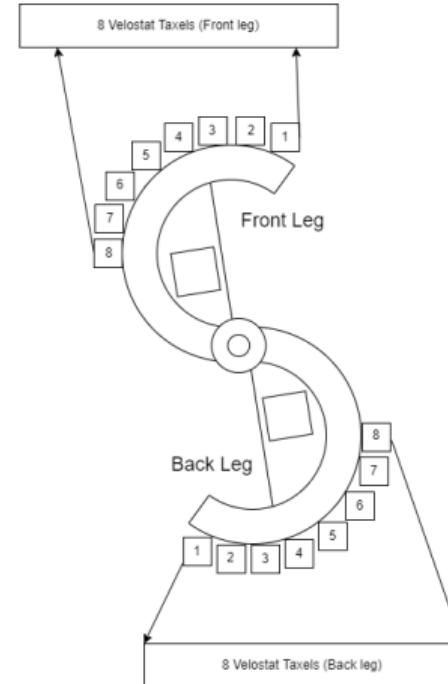
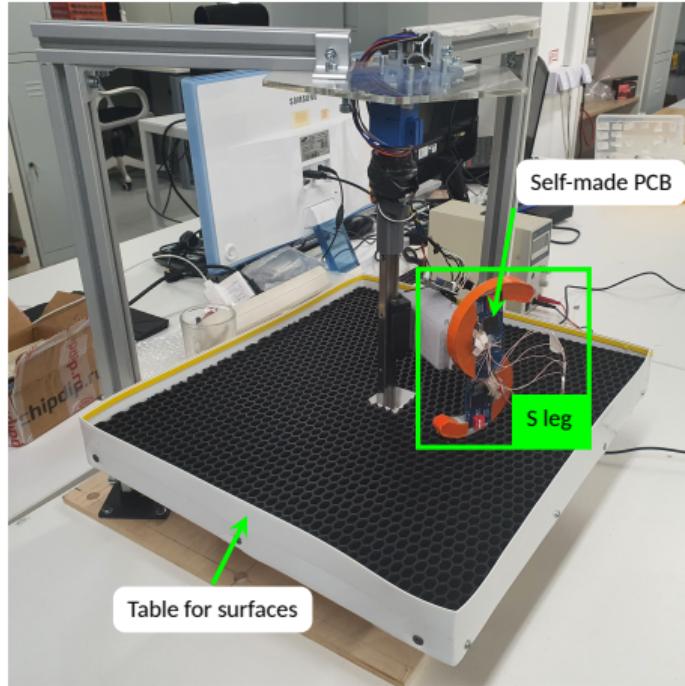
## *Experimental setup requirements*

- To have a possibility to install new surfaces and change it quickly.  
*Solved by quick detachable table*
- Infinite movement.  
*Solved by creating the 2 DoF mechanism and a S-shape leg*
- Movement part should be the same as on the StriRus.  
*Solved by creating a mount for a leg assembly*  
  
*All setup requirements are fulfilled*



# Terrain classification

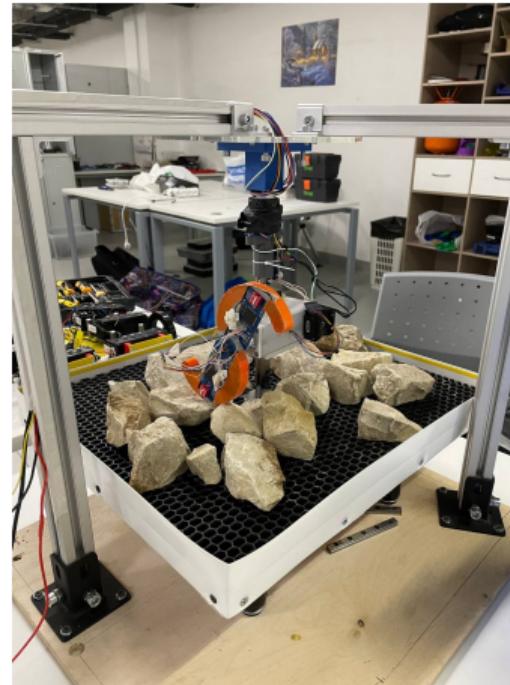
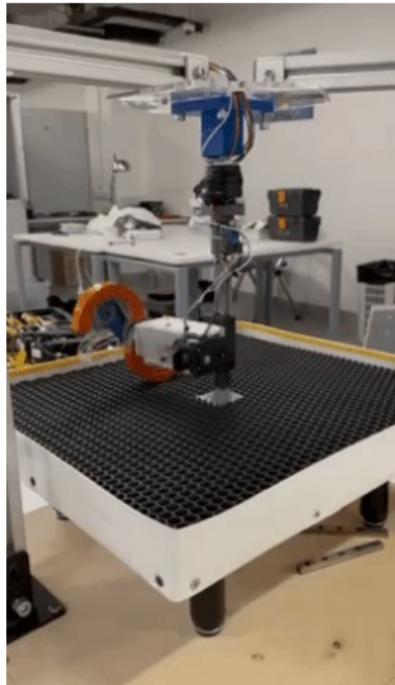
## Experimental Setup





# Terrain classification

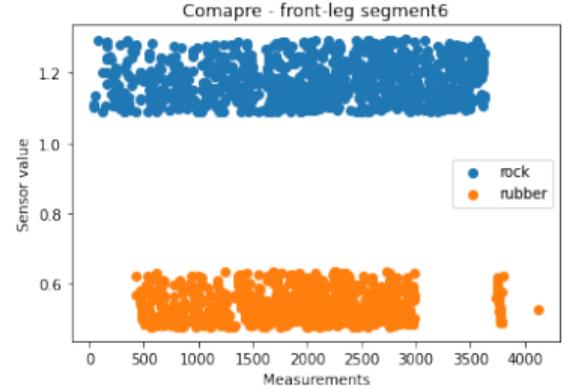
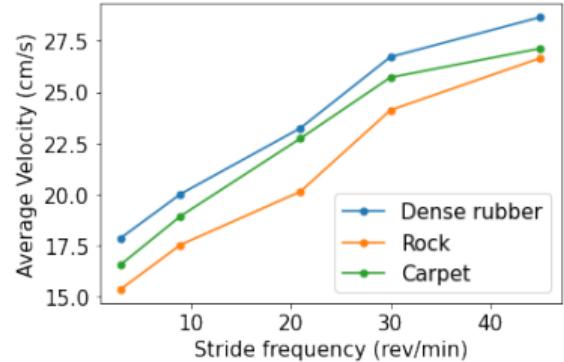
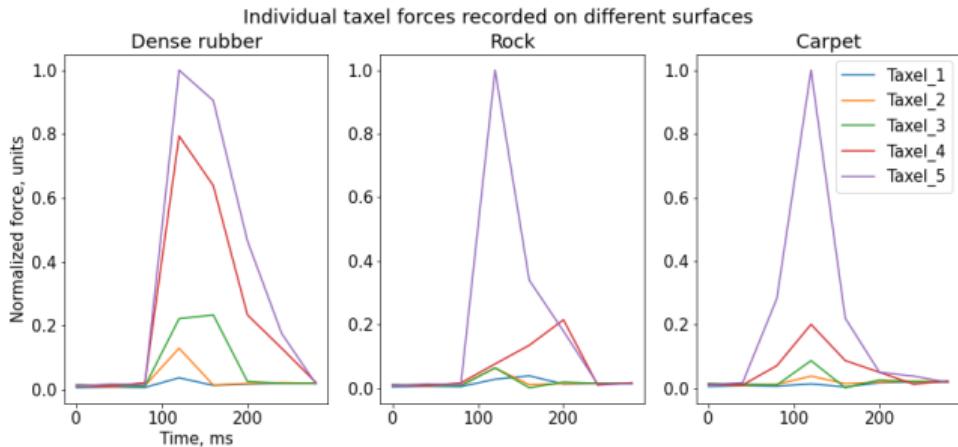
*Experimental setup: surface types, video*





# Terrain classification

Obtained data from one experiment





# Terrain classification

## *Summary*

- We can distinct rubber and rock surfaces
- Terrain classification parameters was chosen:
  - RPM
  - Motor Torque
  - Acceleration from IMU
  - Force data which are represented as Sensor valuesegment, Peak amplitude, Average amplitude
- The force transducer has been proven to work



# Map creation based on tactile data

## Question

How to create a dense point cloud, using sparse data from legs?



# Map creation based on tactile data

## Question

How to create a dense point cloud, using sparse data from legs?

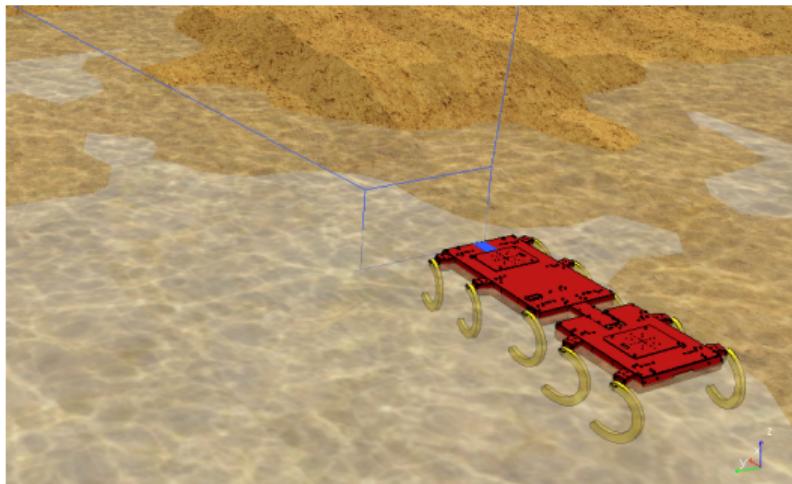
## Answer

*Create a mesh, using concave hull Delaunay triangulation using sparse data,  
sampling it and return to navigation*

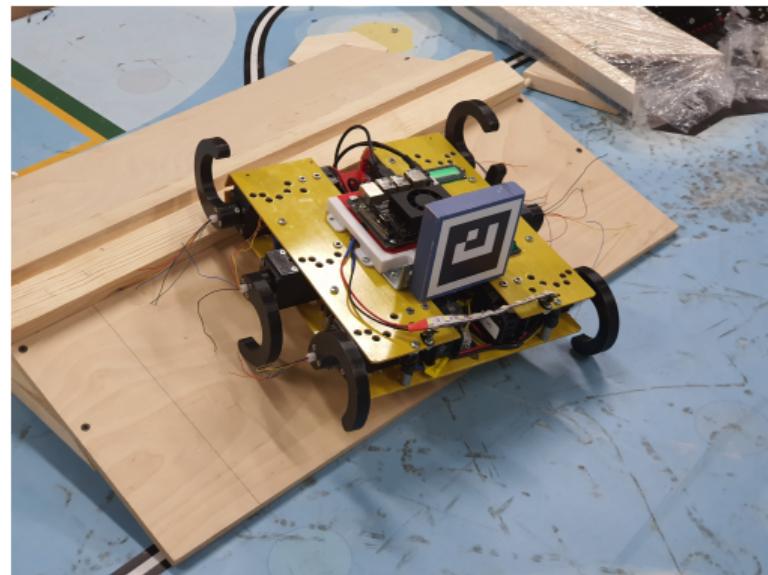


# Map creation based on tactile data

*Experimental setup*



*CoppeliaSim simulator, 4th gen StriRus*



*IRL, 3th+ gen StriRus*



# Map creation based on tactile data

## *Assumptions*

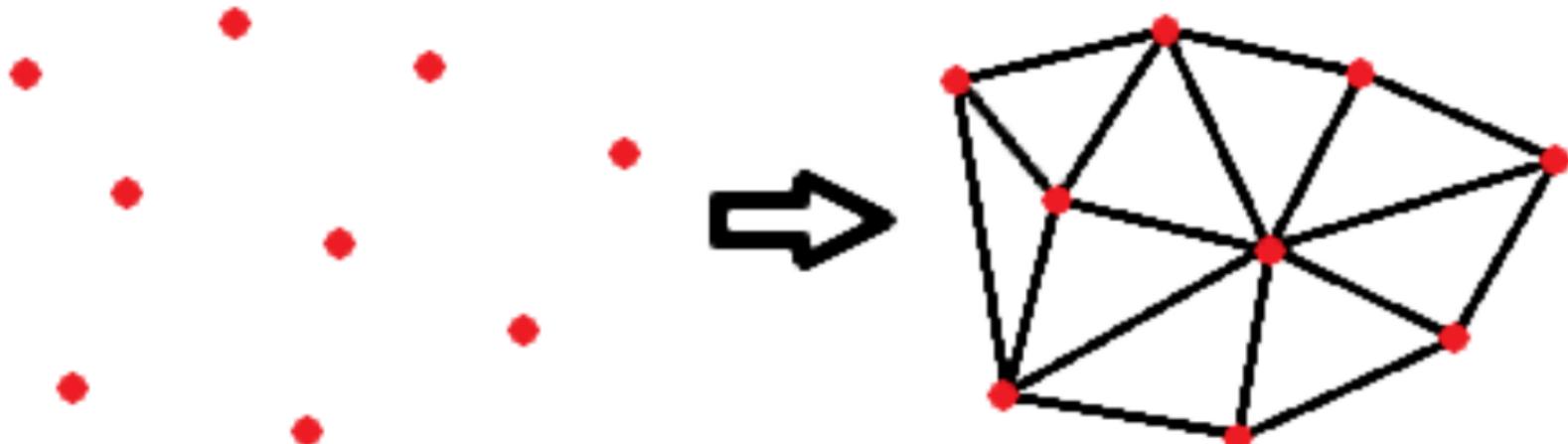
Current solution considering such assumptions:

- Our terrain can be represented  $z = f(x, y)$ . We can use 2D Delaunay triangulation (projected points on a plane)
- All simulation data are preprocessed by white noise



# Map creation based on tactile data

*Delaunay triangulation*

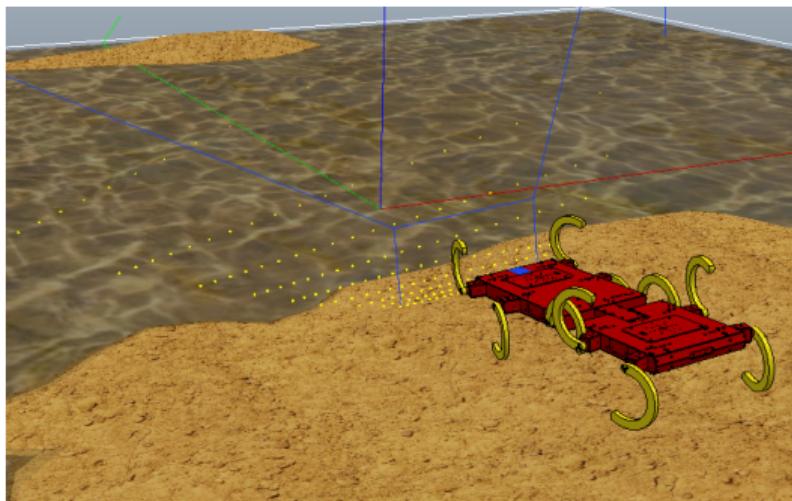


*Common 2D Delaunay triangulation (Convex Hull)  
From Point Cloud to mesh*

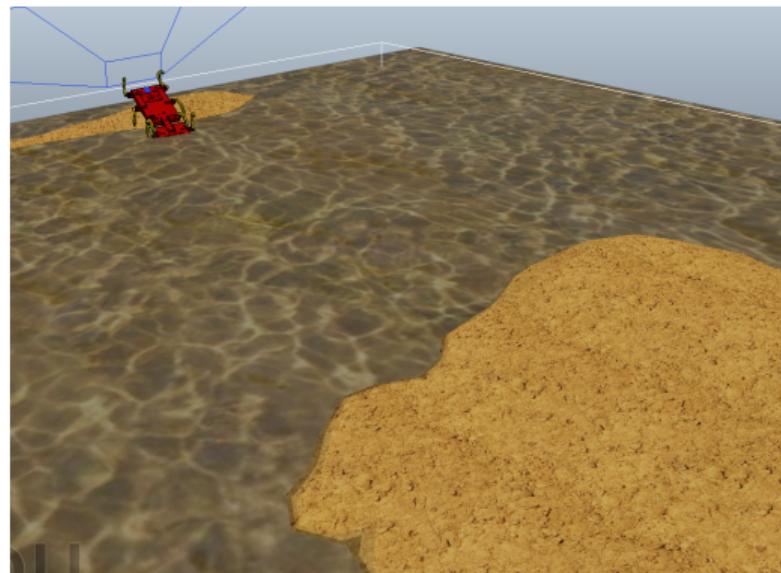


# Map creation based on tactile data

*Result: simulator*



*Start point*

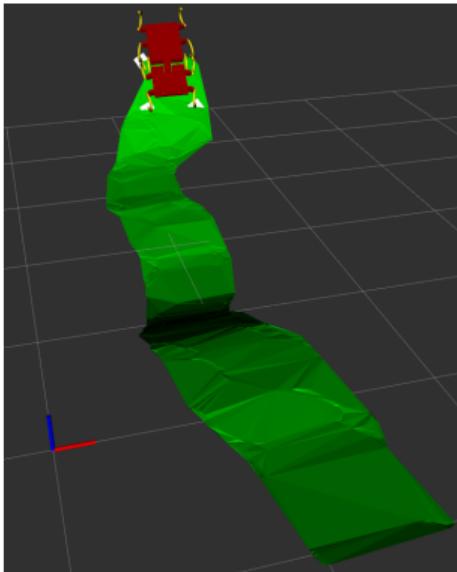


*End point*

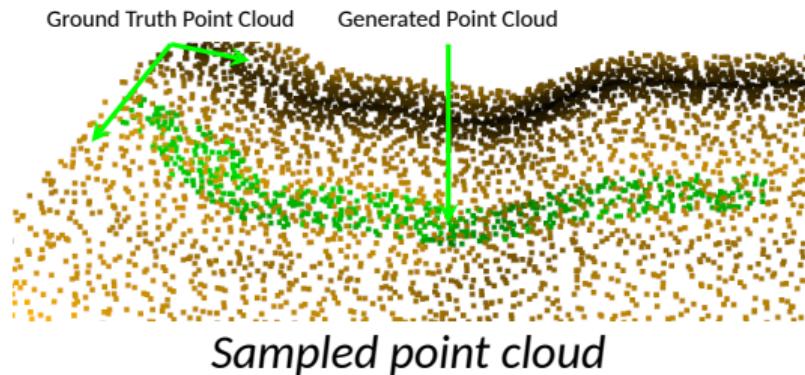


# Map creation based on tactile data

*Result: Mesh*



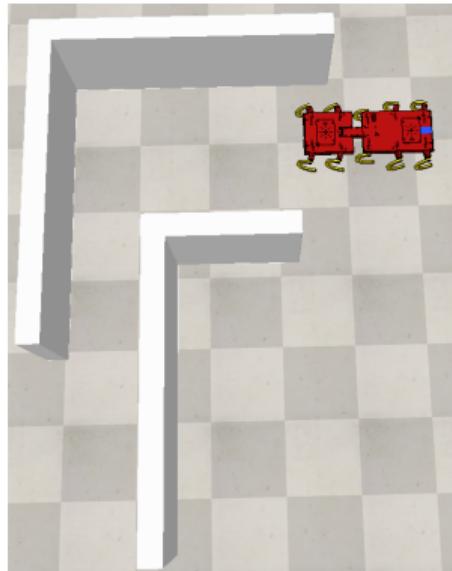
*Mesh created using concave hull 2D  
Delaunay triangulation*



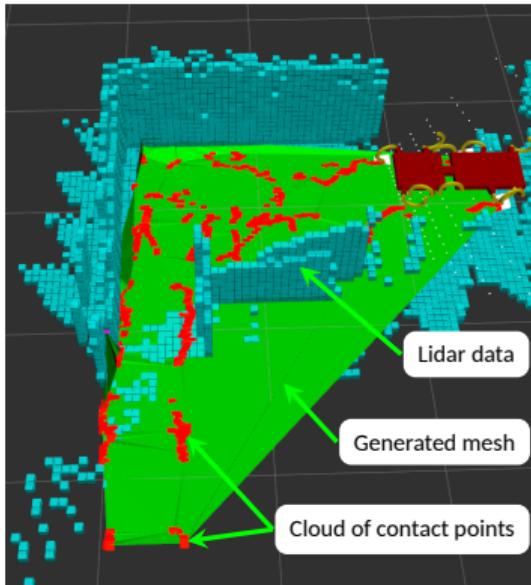


# Map creation based on tactile data

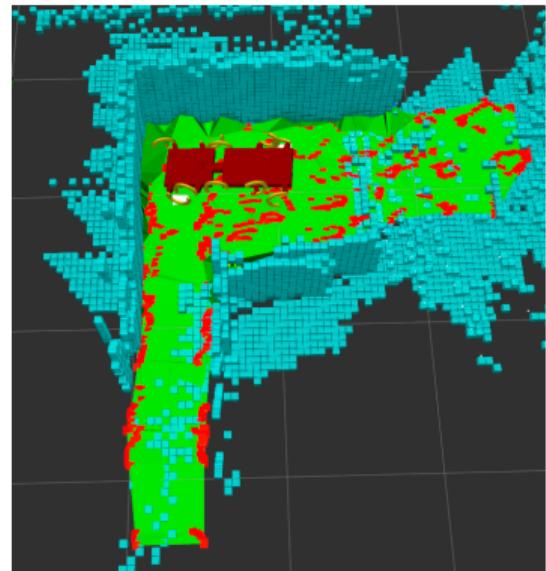
*Why do we need a concave hull solution*



Case study



Convex Hull

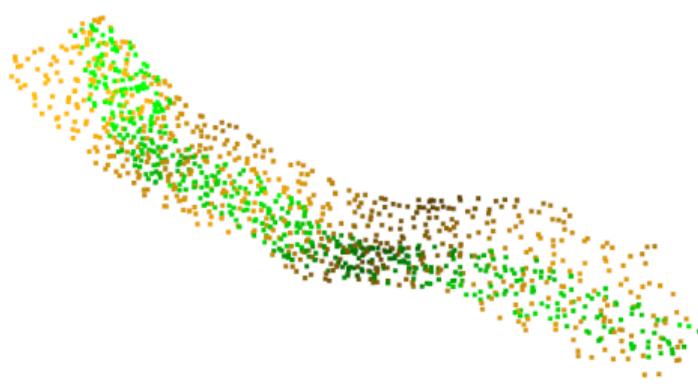


Concave Hull

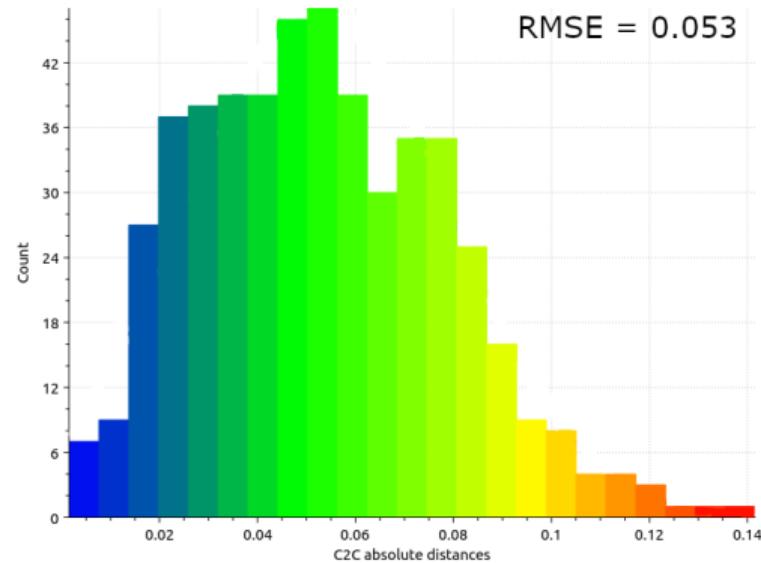


# Map creation based on tactile data

*Metric C2C: reference and ground truth point cloud comparison*



*Overlaid Point Clouds*

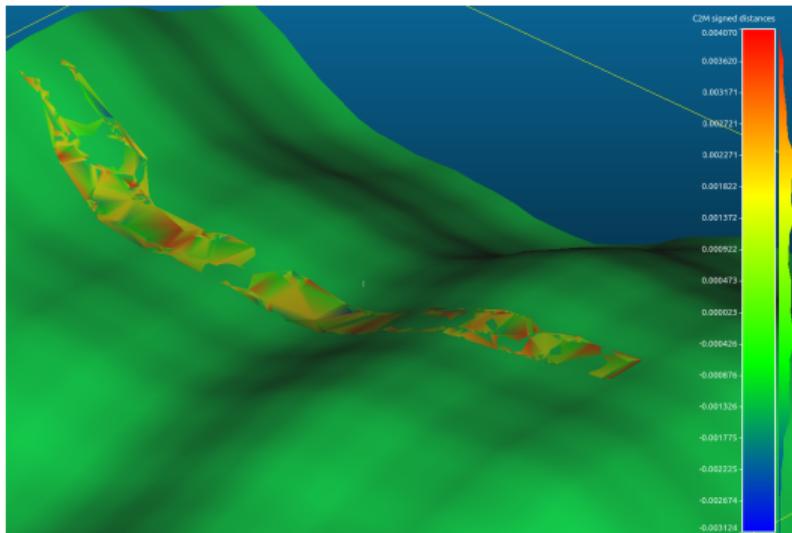


*Error histogram (distance from a point to closest ground truth point)*

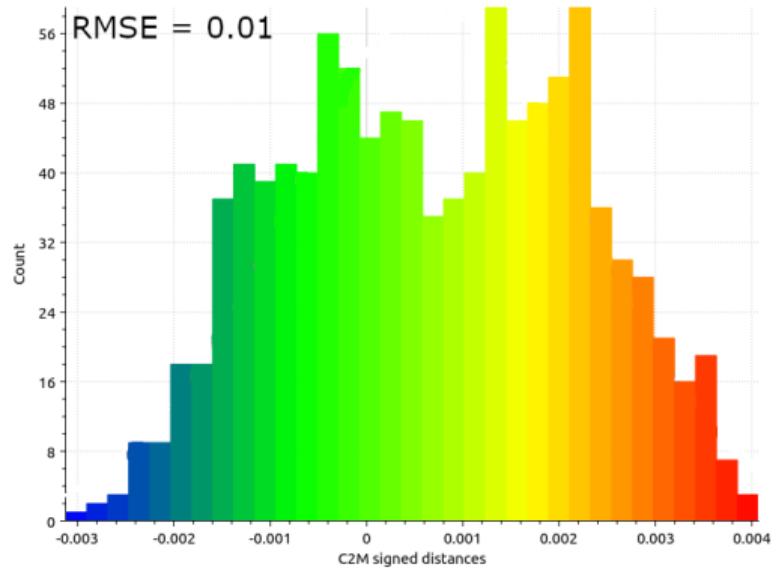


# Map creation based on tactile data

Metric C2M: reference and ground truth mesh comparison



Overlaid Meshes

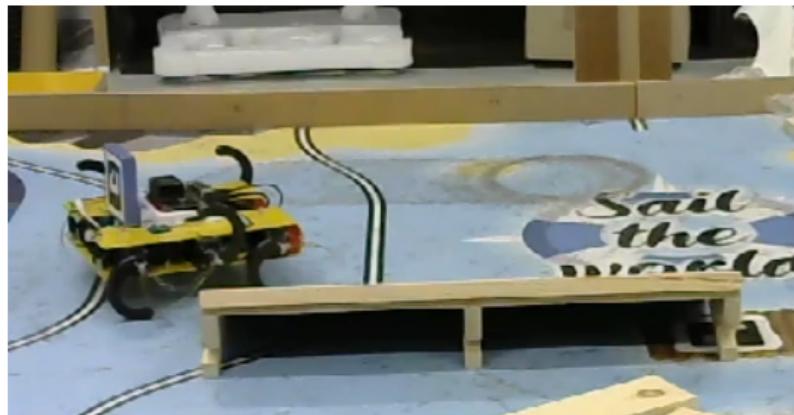


Error histogram (distance from a point to closest ground truth point)

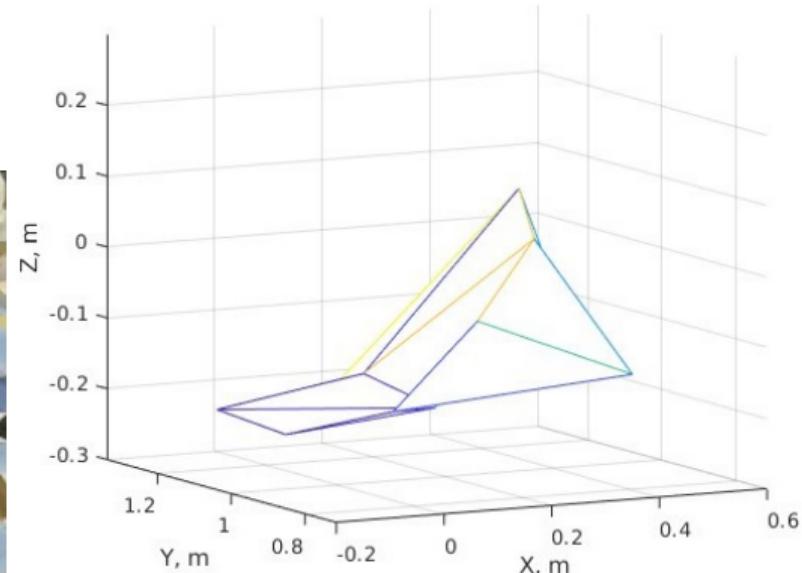


# Map creation based on tactile data

*Result: real world experiment, video*



*Robot is passing the obstacle*



*Mesh, obtained from legs*

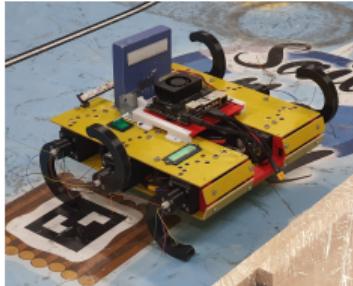


# Map creation based on tactile data

## *Summary*

- Map can be built using concave hull 2D Delaunay triangulation.  
A Sparse point cloud obtained from force sensors, installed on legs.
  - *Simulator:*
    - Avg. Point cloud comparison RMSE is about 5 cm.
    - Avg. Mesh comparison RMSE is about 1 cm.
  - *Real world experiment:*
    - Avg. Point cloud comparison RMSE is about 8 cm.
- It is appropriate accuracy for such task.

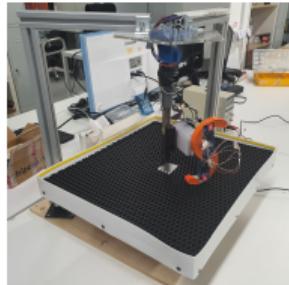
# Summary



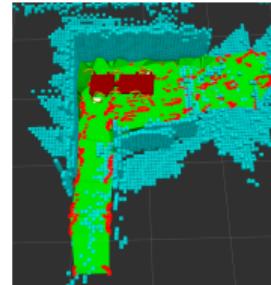
1. *StriRus was designed and assembled*



2. *Force transducer based on Velostat was created and was investigated*



3. *Robot distinct rubber and rock surfaces*



4. *Build a map using tactile sensors*