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MECS 4510: EVOLUTIONARY COMPUTATION AND DESIGN AUTOMATION

Professor Hod Lipson

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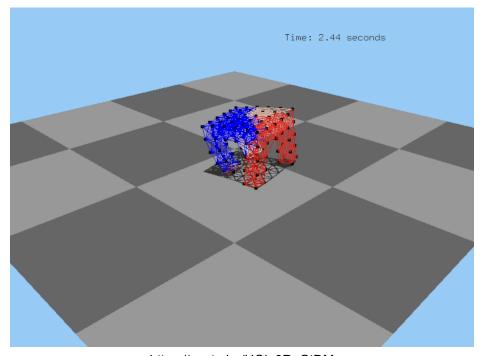
Grace Hours Used: 45

Grace Hours Gained: 1

Grace Hours Remaining: 51

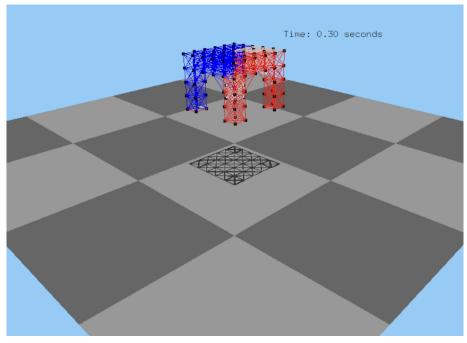
SUMMARY

VIDEO OF FASTEST ROBOT WALKING (4 CYCLES)



https://youtu.be/UOjn6RgGtDM

VIDEO OF FASTEST ROBOT BOUNCING ON THE GROUND



https://youtu.be/b0Hx9RR7q7I

As you can see from the above videos, we have made sure to have nice renderings of the **innovative robot design** that we came up with, we call it *table bot*

We are rendering each spring with a different color, representing each of the material types (white is bone, light orange is soft tissue, and red & blue are opposite muscles)

METHODS

DESCRIPTION OF ALL ROBOT PARAMETERS

Our robot's basic building blocks are Masses and Springs, which are defined with the following parameters:

MASS:

For each of the masses in our robot we keep track of the following information:

- 'm' the mass in kg
- 'p' the [x, y, z] position in meters
- 'v' the [x, y, z] velocity in meters/second
- 'a' the [x, y, z] acceleration in meters/second^2
- 'springs' an array of all of the spring objects connected to that mass

SPRING:

For each of the springs in our robot we keep track of the following information:

- 'm1' the first mass object the spring is connected to
- 'm2' the second mass object the spring is connected to
- 'L0' the rest length of the spring
- 'tissue_type' an indicator of the tissue type (an int between 1-4, used to render the it correctly)
- 'k' the spring constant, dependent on the 'tissue type'
- 'b' the amplitude of oscillation for muscle 'tissue types' that provides actuation
- 'c' the phase shift in oscillation for muscle'tissue types'
- 'center' the [x, y, z] position of the center point of the spring

CUSTOM BODY:

Given the above definitions, we describe our robot in the following way:

- 'masses' an array of all of the mass objects contained in that body
- 'springs' an array of all of the spring objects connecting all of the masses (without repetition)
- 'tissue_dict' a dictionary with the (k, b, c) values characteristic to each tissue_type (namely, bone has a high k & no b nor c, soft tissue has a low k & no b nor c, muscles have a medium k & values for b and/or c)
- 'genome' we initialize springs to be a certain 'tissue_type' is by picking 8 random mass points in our body and, for all of the springs, calculating which is closest (by using the 'center' position of

each spring), we then assign the 'tissue_type' from the 'tissue_dict' as well as the 'k', 'b' and 'c' values to those springs - the genome is simply an array of the 3D point and the tissue type we are initializing fo our Custom body

- 'COM' - the Center-of-Mass value for our robot, used to evaluate the fitness in locomotion

DESCRIPTION OF ALL EVOLUTIONARY PARAMETERS

Currently only doing random search due to the runtime limitation of our simulation - we are simulating a structure of ~750 springs, the equivalent to ~28 cubes, with python - therefore, for this assignment we have only been able to apply a random search as we can not realistically complete more than 10s of evaluations even in 8 hours of running our code

Therefore, for now we are simply generating our fixed 'Custom Body' robots with the random 'genome' (ie. the positions where different tissue types are initialized for our robot)

For the next assignment, once we have properly sped up our runtime with Numba, we will apply evolutionary parameters like mutation (changing the location where tissues are initalized to an adjacent mass location) and cross-over (splitting the genome at two locations and crossing over the description)

A thought about linkage - we will be reorganizing the genome to place points that are closer to each other closer in the genome, therefore when crossing over, genes that work well together are more likely to be passed on

The fitness function currently is the absolute distance the robot moves in the x-y plane from the initial position to the final position. We are planning to use co-evolution whereby we will be optimizing the muscle weights (values 'k', 'b' and 'c') as we optimize the morphology of the robot

DESCRIPTION OF WHAT WORKS AND DID NOT WORK

The main issue we are encountering is that our simulation is running very slow, therefore optimizing on such a large robot is extremely difficult and attempting fancier optimization techniques is simply unfeasible

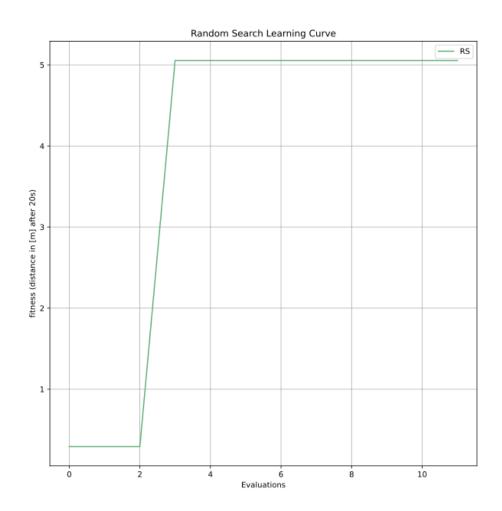
All of our functionality works and we are excited to implement Numba to be able to have our evolutionary algorithms shine

NUMBER OF SPRING EVALS/SECOND

~140,000 (on my personal laptop pre-NUMBA, running python)

RESULTS

LEARNING CURVE



BIBLIOGRAPHY

ChatGPT

PyOpenGL - https://pyopengl.sourceforge.net/index.html

Stack Overflow

APPENDIX

```
from Libraries import *
#응응
class Mass:
class Custom body 1:
0], 0.0), Genome size = 8, prev genome=None, only bounce=False):
```

```
4: (5000, -.125, 0)}
```

```
if name == " main ":
```

```
genome = np.array([[[0.2, 0.4, 0.300000000000004], [1000.0, 0.0, 0.0]], [[0.0, 0.2,
0.3000000000000004], [20000.0, 0.0, 0.0]], [[0.1, 0.300000000000004, 0.4], [1000.0,
0.0, 0.0]], [[0.1, 0.0, 0.0], [20000.0, 0.0, 0.0]], [[0.4, 0.4, 0.2], [5000.0, -0.125,
0.0]], [[0.4, 0.5, 0.2], [20000.0, 0.0, 0.0]], [[0.4, 0.0, 0.0], [1000.0, 0.0, 0.0]],
[[0.1, 0.2, 0.4], [5000.0, -0.125, 0.0]]])
body = Custom_body_1(prev_genome = genome)
#응응
from Libraries import *
from Datastructures import Custom body 1 as custom
from MAIN import *
#응응
class Simulate:
```

```
pygame.display.set mode(display, DOUBLEBUF|OPENGL)
for event in pygame.event.get():
            if event.type == pygame.QUIT:
                pygame.quit()
pygame.display.flip()
   pygame.init()
    pygame.display.set mode(display, DOUBLEBUF|OPENGL)
```

```
if event.type == pygame.QUIT:
                       pygame.quit()
                   pygame.display.flip()
in sim")
```

```
if name == " main ":
#응응
from Libraries cloud import *
from Datastructures cloud import *
from Simulation cloud import *
#응응
class EvolvingGait:
```

```
for i in tqdm(range(self.pop_size), desc='Evaluating:'):
Simulate(body=self.population[i]).run_simulation(Plot=False, max_T=2)
if name == " main ":
```