

Winter 2026 Robot Mapping Workshop



[https://existentialrobotics.org/
RobotProvingGrounds/](https://existentialrobotics.org/RobotProvingGrounds/)



**CONTEXTUAL
ROBOTICS**
INSTITUTE

UC San Diego

JACOBS SCHOOL OF ENGINEERING
Electrical and Computer Engineering

Agenda

5:05 - 5:20: Introduction to ERL

5:20 - 6:35: Mapping Algorithm Presentation and Activities

6:35 - 6:50: Pybullet Simulation (optional)

6:50 - 7:00: Live Robot Demo and Food

Who are we?

Faculty Mentors



**Nikolay
Atanasov**



**Shatha
Jawad**

PHD Student Mentors



**Nikola
Raičević**



**Mani
Amani**



**Yin Zhuang
Yi**

Undergraduate and K12 Students



**Hyungjun
Doh**



**Afraz
Hameed**



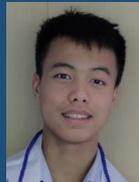
Ryan Teck



**Stephen
Huang**



Allie Dinh



Lek Man



**Steven
Hsiao**



**Tarun
Jaikumar**

Past Contributors

- Eric Zheng
- Ali Hussain
- Muhammad Fadli Alim Arsani
- Shuyan Tan
- Andrew Nemeth
- Darren Ng
- Zeyu (Jeffrey) Chen
- Shuyan Tan
- Tan Muhammad Arsani
- Risab Sankar
- Chengkai Yao
- Aniket Bhosale
- Minghan (Travis) Wu
- Chakshan Kothakota
- Fayyad Hassan
- Kibum Kim
- Ali Hussain
- Yaobang Deng
- Weifan Ou
- Trung Tran
- Aditi Krishnakumar
- Anthony Hiraes
- Ahuatzin
- Peter Stratton
- Shreyas Arora
- Hannah Hui
- Farnia Nafarifard
- Ryan Goh
- Aaron Yu
- Rohan Bosworth
- Ke Ou

What We Do

- Teach basics of various robotics algorithms
- Ex: Mapping, Localization, Planning, and Control



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What is Mapping?

Mapping Problem



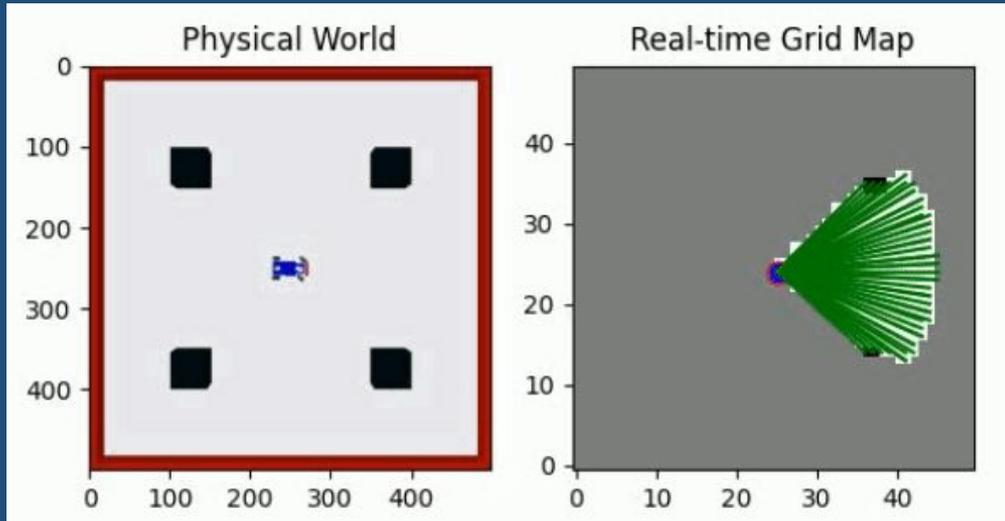
Roomba



Waymo

- For full autonomy, robots need to know what is in their environment
- This allows for many possibilities (exploring dangerous environments, providing services, etc.)

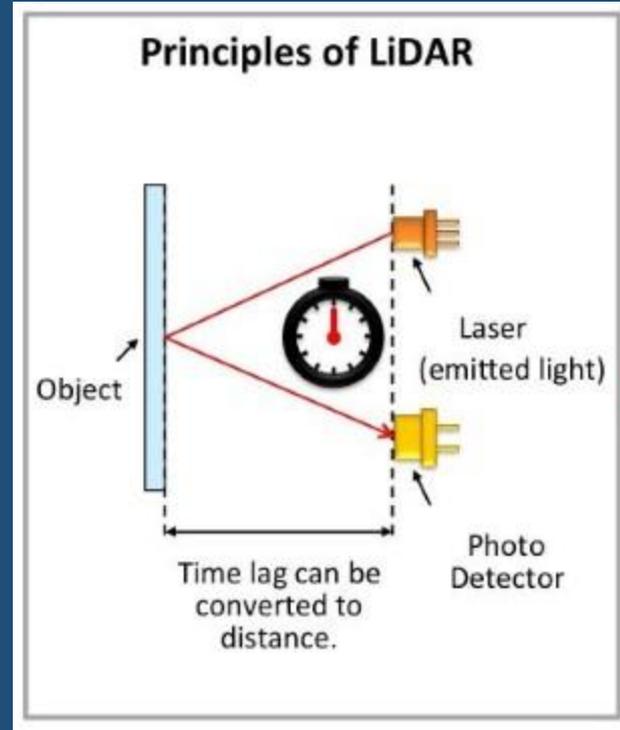
OGM Mapping



- Occupancy Grid Mapping (OGM)
- Uses sensors to create a map of the world
- Map tells if a space is filled, empty, or unknown

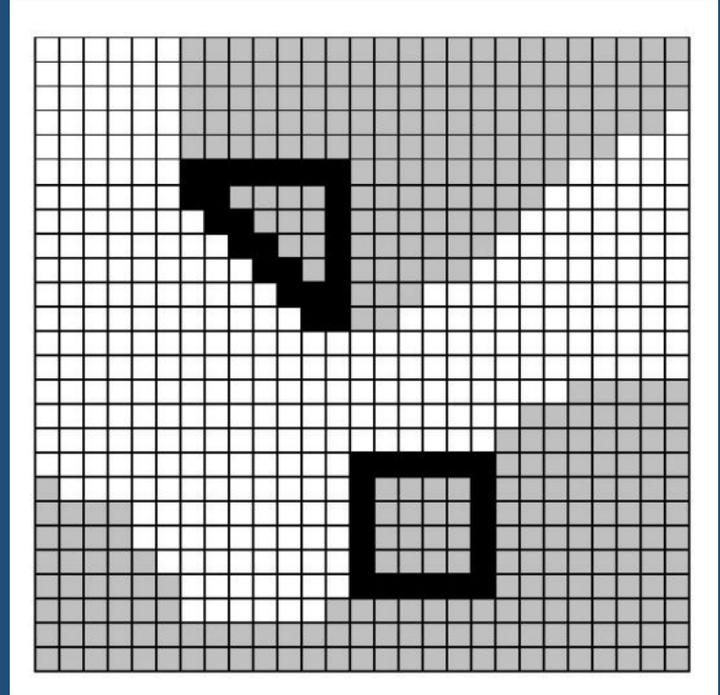
Light Detection and Ranging (LiDAR)

- Emits lasers and measures return time off reflections.
- LiDAR does have a max distance
- LiDAR spins while emitting lasers at high speeds “see” surroundings
- Measurements:
Angle & Distance



Map Updates

- Stores map in a matrix
- Uses sensor data to mark filled cells
- Space between robot and sensor hit is marked as empty
- Draws empty spaces with line drawing algorithm



Line Drawing (Bresenham)

Line Drawing

Basic Line Drawing (Python)

```
1 def draw_line(x0, y0, x1, y1):
2     m = (y1 - y0) / (x1 - x0)
3     dx = abs(x1-x0)
4
5     for i in range (dx):
6         y = round(m * i + y0)
7         print(dx + i, y)
```

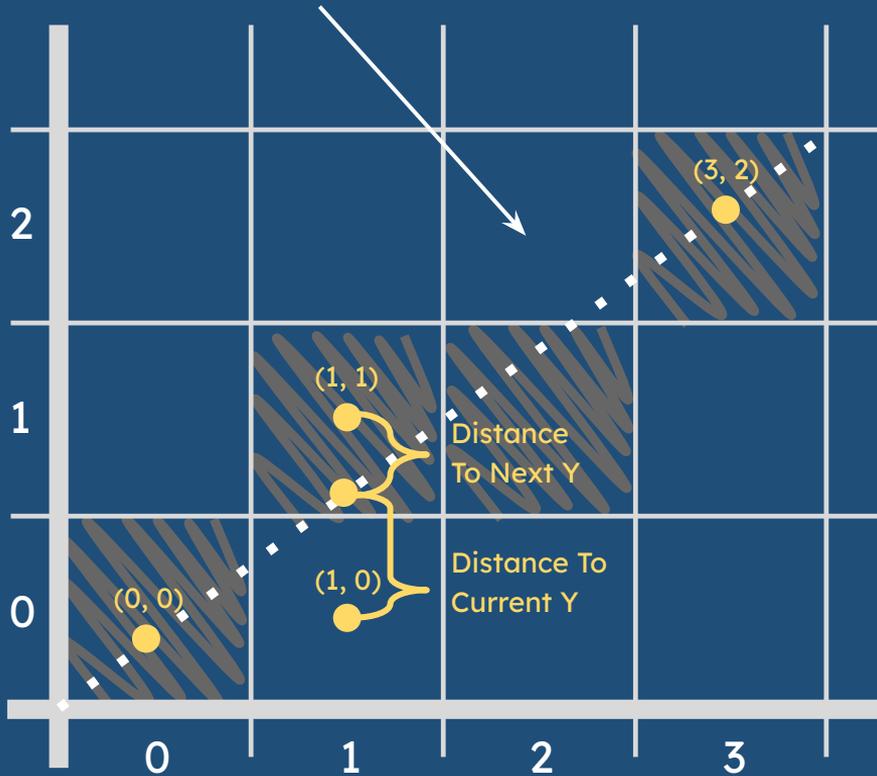
Based off: “ $y = mx + b$ ”

← Finds Slope

← Uses “ $y=mx + b$ ” to find “ y ”
Round decides which whole integer “ y ” should be

Line Drawing

Grid not filled because we're taking approximations



- Using decimals can slow down code
- Variation of Bresenham Algorithm Solves This!
- When To Increase Y:
 - When the real y-value is closer to the next integer
- We'll keep track of this choice with a **decision variable** called "D"

Line Drawing

y_0 = starting y-value
 y = current y-value
 i = x-distance moved

$f(i+1) = m(i+1) + y_0$
↵ function to get y-coord
based on x-distance moved

1	$D = \text{Distance To Current} - \text{Distance To Next}$
2	$= [f(i + 1) - y] - [y + 1 - f(i + 1)]$
3	$= [m(i + 1) + y_0 - y] - [y + 1 - (m(i + 1) + y_0)]$
4	$= 2 m (i + 1) + 2 y_0 - 2 y - 1$
5	$= 2 (\Delta y / \Delta x) * (i + 1) + 2 y_0 - 2 y - 1$

Interpreting D's results

When D is positive

- We're going to be closer to the next y-value and thus increment

When D is zero

- We're going exactly to the next y-value and thus should increment

When D is negative

- We're going to be closer to the current y-value and thus should stay

Line Drawing

$$1 \quad D = 2 (\Delta y / \Delta x)^* (i + 1) + 2 y_0 - 2 y - 1$$

$$2 \quad (\Delta x) D = 2 (\Delta y) (i + 1) + 2 (\Delta x) y_0 - 2 (\Delta x) y - (\Delta x)$$

$$3 \quad D = 2 (\Delta y) (i + 1) + 2 (\Delta x) y_0 - 2 (\Delta x) y - (\Delta x)$$

$$4 \quad D = 2 (\Delta y) i + 2(\Delta y) + 2(\Delta x) y_0 - 2 (\Delta x) y - (\Delta x)$$

Since we only care about sign of D, we can forgo the Δx on the left side

Line Drawing

Pseudocode Recap!

- Intake x_0 , x_1 , y_0 , and y_1
 - Calculate dx & dy
 - Set y to y_0
 - Loop through x -axis
 - Mark current coords
 - Calculate Decision Var
 - Check if D is non-negative
 - If yes, increment y

Line Drawing

What if we calculate D earlier?

- Intake x_0 , x_1 , y_0 , and y_1
 - Calculate dx & dy
 - Set y to y_0
 - Calculate Decision Var
 - Loop through x-axis
 - Mark current coords
 - Check if D is non-negative
 - If yes, inc y
 - Update Decision Var

Line Drawing

What if we calculate D earlier?

- Intake x_0 , x_1 , y_0 , and y_1
 - Calculate dx & dy
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 - Mark current coords
 - Check if D is non-negative
 - If yes, inc y
 - Update Decision Var

$$\begin{aligned} D &= 2(\Delta y)i + 2(\Delta y) + 2(\Delta x)y_0 - 2(\Delta x)y - \Delta x \\ &= 2(\Delta y)(0) + 2(\Delta y) + 2(\Delta x)y_0 - 2(\Delta x)y_0 - \Delta x \\ &= 2(\Delta y) - \Delta x \end{aligned}$$

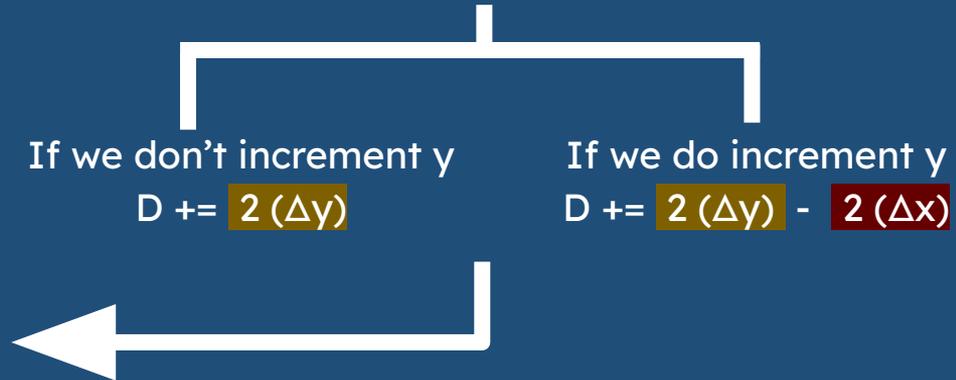
Line Drawing

What if we calculate D earlier?

- Intake x_0 , x_1 , y_0 , and y_1
 - Calculate dx & dy
 - Set y to y_0
 - Calculate Decision Var
 - Loop through x-axis
 - Mark current coords
 - Check if D is non-negative
 - If yes, inc y
 - Update Decision Var

$$D = 2(\Delta y)i + 2(\Delta y) + 2(\Delta x)y_0 - 2(\Delta x)y - \Delta x$$

SO...



Line Drawing

Potential Problems

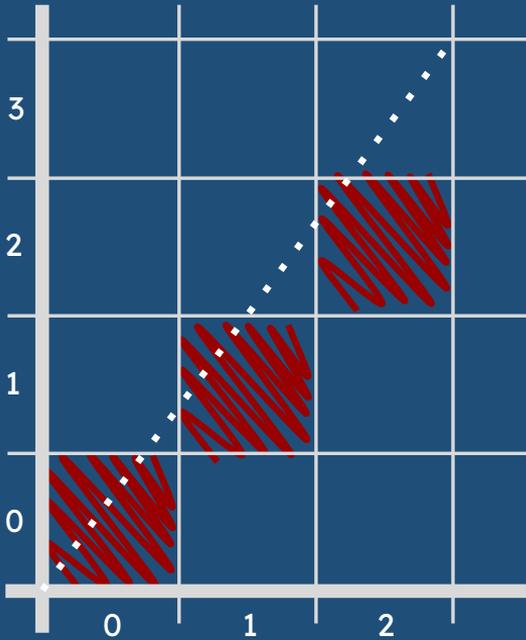
- 1 Slope being negative
- 2 Direction of Line
- 3 Slope greater than 1



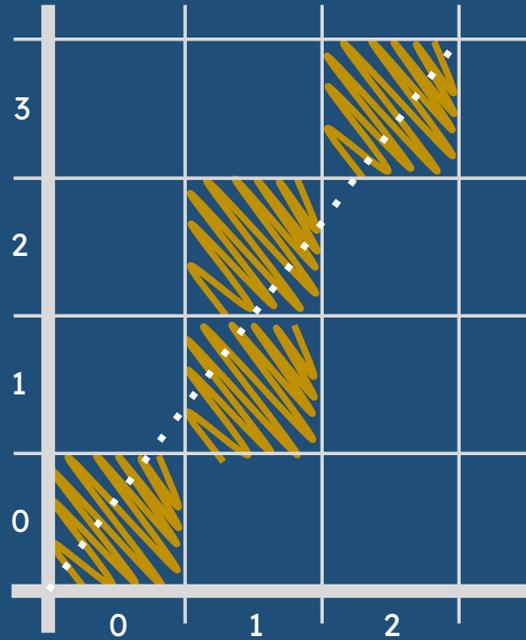
- Keep track of whether you're increasing or decreasing the x and y axis

Line Drawing

Looping through x-axis



Looping through y-axis



Dominant Variable

If dy is greater than dx :

- Swap all references of x and y

Bresenham Activity

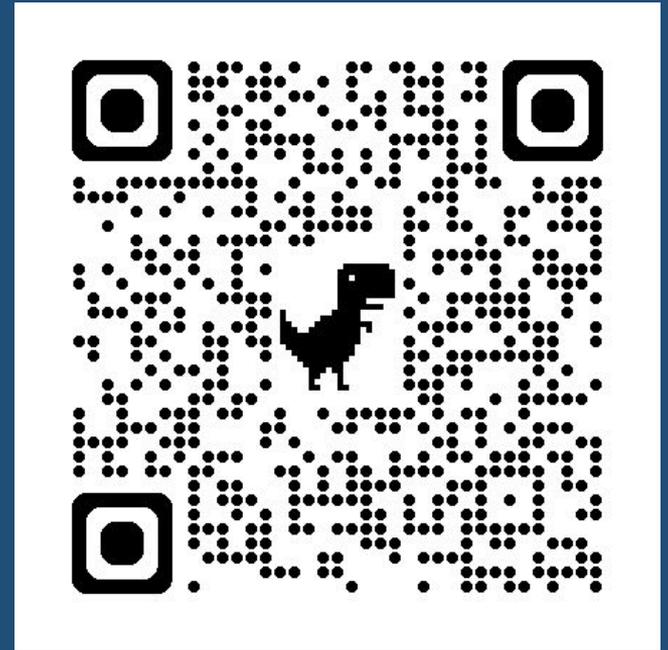
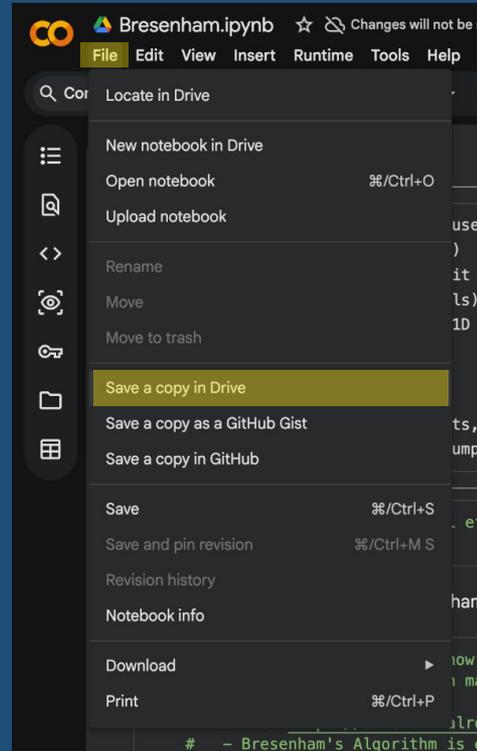
- Go to website
- Navigate to “Workshop” tab
- Scroll down to “Notebooks (During Workshop)”
- Open “**Bresenham**” notebook
- Code! (Instructions in notebook)



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Bresenham Activity

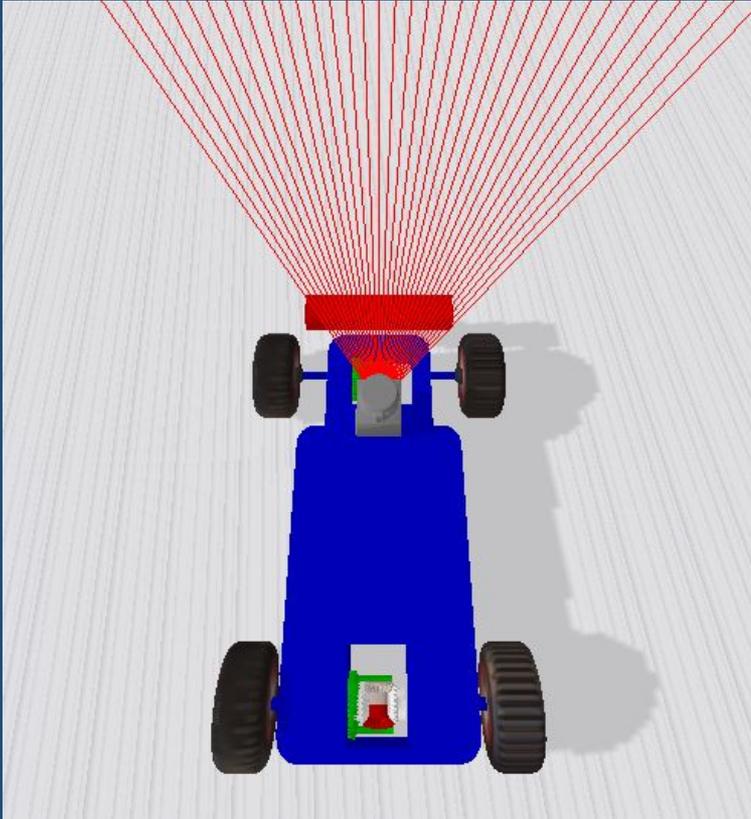
File > Save a copy in Drive



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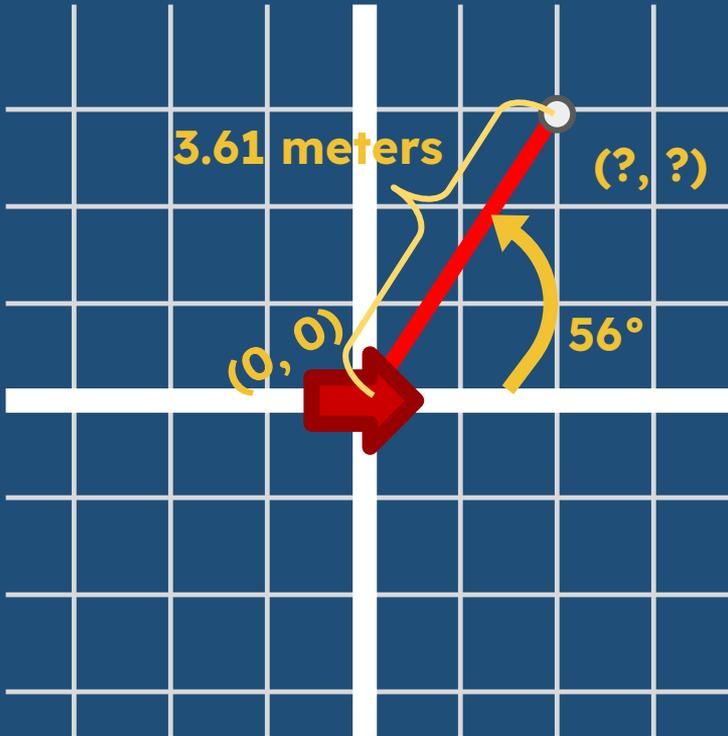
LiDAR

Angle Separation



- Known Specifications of LiDAR
 - Start Angle
 - End Angle
 - Number of rays
 - Max Ray Range (length)
- Angle Separation: Angle between adjacent rays
 - $\text{Angle Separation} = \text{FOV} / \# \text{ of Rays}$
- Can determine angle of each ray from LiDAR's perspective

LiDAR Endpoints

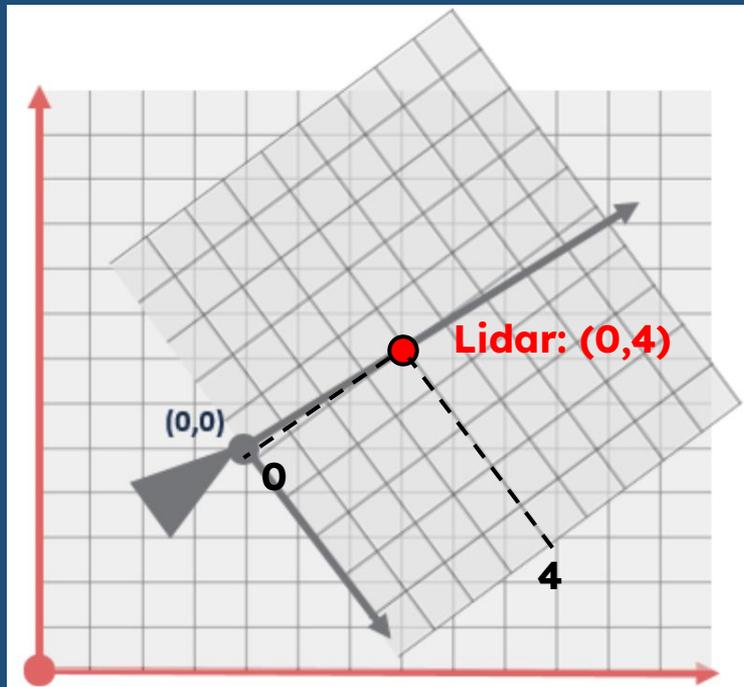


*Assume LiDAR faces 0°

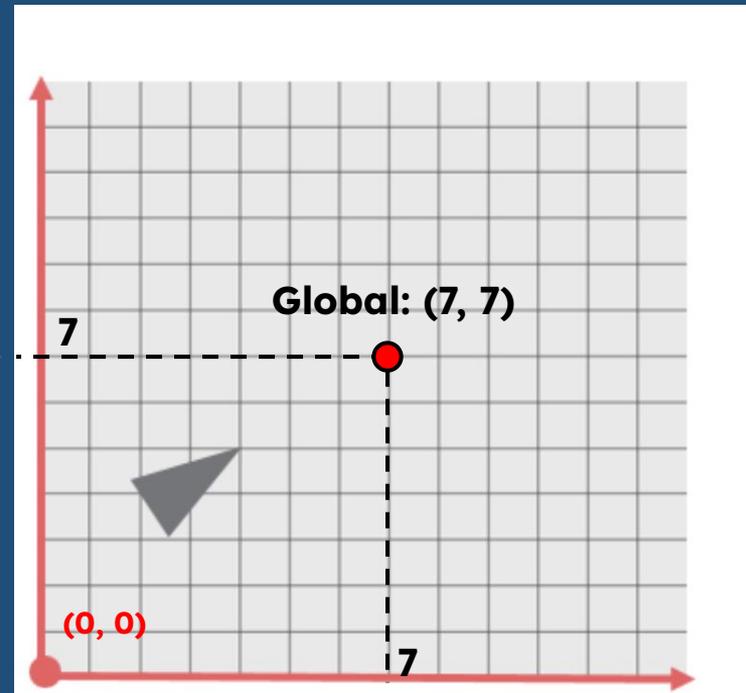
What We Know

- Ray Angle (θ_L): 56° counterclockwise
- Ray Distance (d): 3.61 meters
- How to map onto grid?
 - Need (x, y)
- Use Trigonometry:
 - $x = d * \cos(\theta_L) = 3.61 \text{ m} * \cos(56^\circ) = 2 \text{ m}$
 - $y = d * \sin(\theta_L) = 3.61 \text{ m} * \sin(56^\circ) = 3 \text{ m}$

Frame Transformations



Robot Frame



Global Frame

Frame Transformations

Finding the position of a point in frame A, knowing:

- The point in frame B
- The relative position of B to A

$$\begin{bmatrix} x_A \\ y_A \end{bmatrix} = \begin{bmatrix} x_{B/A} \\ y_{B/A} \end{bmatrix} + \begin{bmatrix} \cos(\phi) & -\sin(\phi) \\ \sin(\phi) & \cos(\phi) \end{bmatrix} \begin{bmatrix} x_B \\ y_B \end{bmatrix}$$



$$x_A = x_{B/A} + [x_B \cos(\phi) - y_B \sin(\phi)]$$

$$y_A = y_{B/A} + [x_B \sin(\phi) + y_B \cos(\phi)]$$

x_A & y_A = Point in Frame A

$x_{B/A}$ & $y_{B/A}$ = Position of Frame B in Frame A

ϕ = Angle of Frame B in Frame A

x_B & y_B = Point in Frame B

LiDAR Frame Transformations

$$\begin{bmatrix} x_{\text{world}} \\ y_{\text{world}} \end{bmatrix} = \begin{bmatrix} x_0 \\ y_0 \end{bmatrix} + \begin{bmatrix} \cos(\phi) & -\sin(\phi) \\ \sin(\phi) & \cos(\phi) \end{bmatrix} \begin{bmatrix} d\cos(\theta) \\ d\sin(\theta) \end{bmatrix}$$



$$x_{\text{world}} = x_0 + d * \cos(\phi + \theta)$$

$$y_{\text{world}} = y_0 + d * \sin(\phi + \theta)$$

x_{world} & y_{world} = Global Frame

x_0 & y_0 = LiDAR Position

ϕ = LiDAR Angle

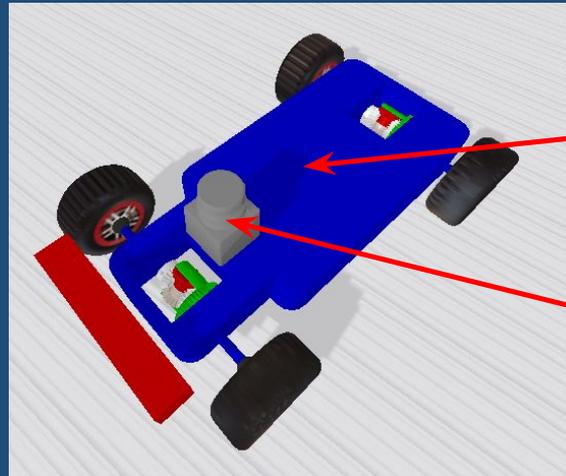
d = Ray Hit Distance

θ = Ray Angle

LiDAR Frame Transformations

What if.....

Robot



Center

LiDAR

Frame Transformations

- 1) LiDAR Frame \Rightarrow Robot Frame \Rightarrow Global Frame
 1. Point in LiDAR Frame \Rightarrow Robot Frame
 - Obtain point in Robot Frame
 2. Point in Robot Frame \Rightarrow Global Frame
 - Obtain point in Global Frame

- 2) LiDAR Frame \Rightarrow Robot Frame (only 1 transformation needed)

Occupancy Grid Mapping (OGM)

Binary VS Probabilistic

Binary OGM

Uses latest
sensor data
only to
identify new
hits

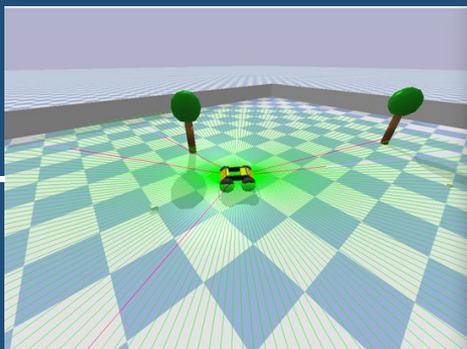
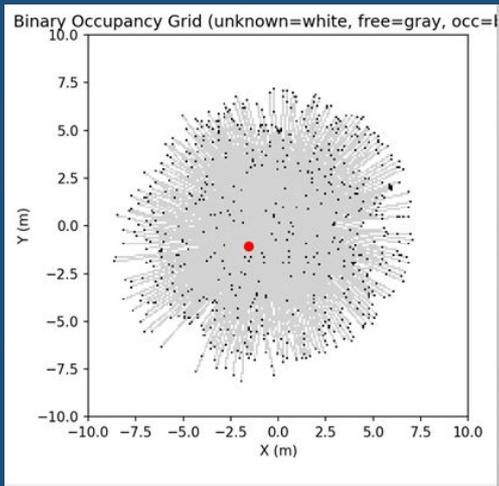
Probabilistic OGM

Combines latest and
past sensor data to
build map

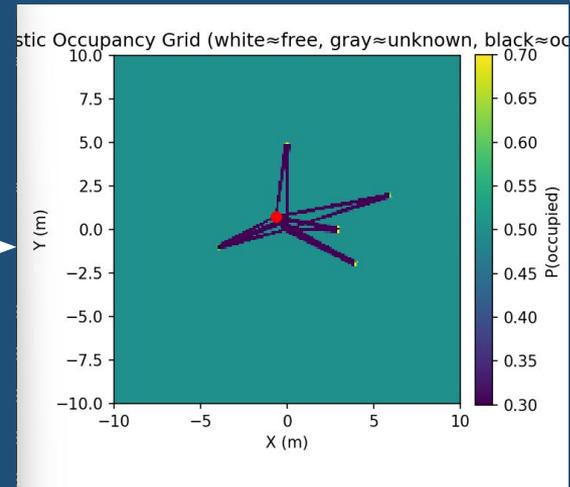
Can Help Partially
with Unpredictability
(ex: Faulty Readings,
Moving Objects/People)

Binary vs Probabilistic Occupancy Grid Mapping (with Sensor Noise)

Binary OGM



Probabilistic OGM



* Noise produces permanent false obstacles *

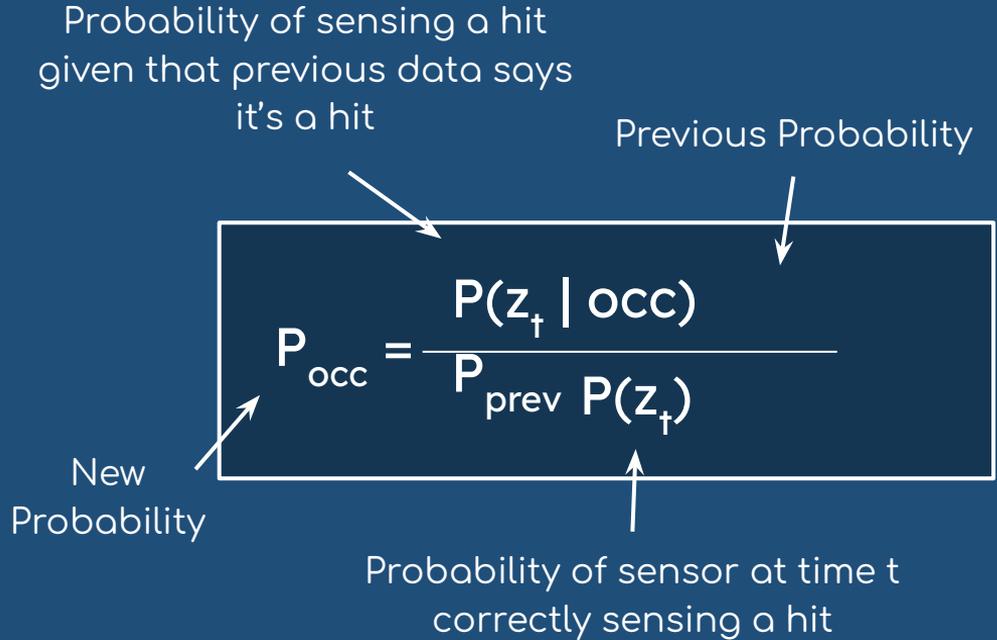
* Noise is averaged out through evidence accumulation *

Probabilistic

Bayes' Theorem

- Probability of event A occurring given that event B has occurred

Log-Odds formulation is more numerically manageable



Odds and Log-Odds

Odds

- Odds is the ratio between our confidence in success over confidence in failure.

$$\text{odds} = \lambda / (1 - \lambda)$$

Confidence in success

Confidence in failure

$$\ln(\text{odds}) = \ln(\lambda / (1 - \lambda))$$

$$\text{odds}_{t+1} = \prod_{i=1}^t (\text{odds}_i) * \text{update}$$

$$\ln(\text{odds}_{t+1}) = \sum_{i=1}^t \ln(\text{odds}_i) + \ln(\text{update})$$

OGM Algorithm

Initialization

Fill every cell with
an initial
probability

0 in terms of
log-odds

Sensor

Likelihood

Find odds of the sensor
data based on confidence
(λ)

$$\text{odds} = \lambda / (1 - \lambda)$$

Compute its log-odds

+ $\ln(\text{odds})$ for hit
- $\ln(\text{odds})$ for miss

Map Update

Find new log-odds

$$L_{\text{new}} = L_{\text{old}} + L_{\text{update}}$$

Convertible back to
probability via

$$P = 1 / (1 + e^{-L})$$

Apply probability
to grid

LiDAR and OGM Activity

- Go to website
- Navigate to “Workshop” tab
- Scroll down to “Notebooks (During Workshop)”
- Complete **“LiDAR and Frame Transformations”** & **“Occupancy Grid Mapping (OGM)”** notebooks

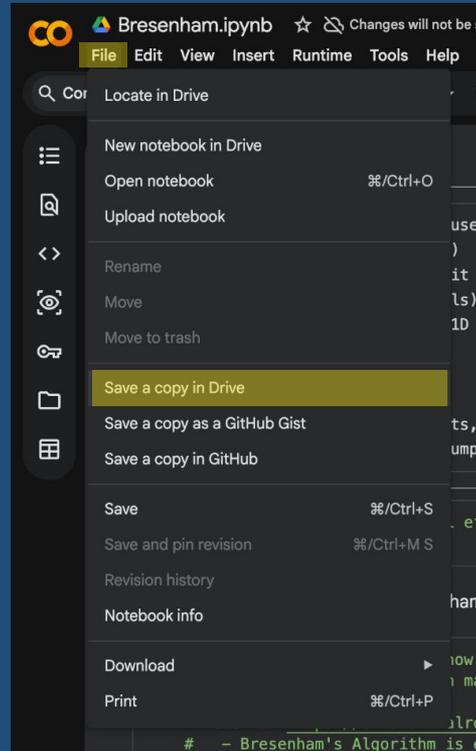
Note: Follow the instructions in the text cells carefully



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LiDAR and OGM Activity

File > Save a copy in Drive



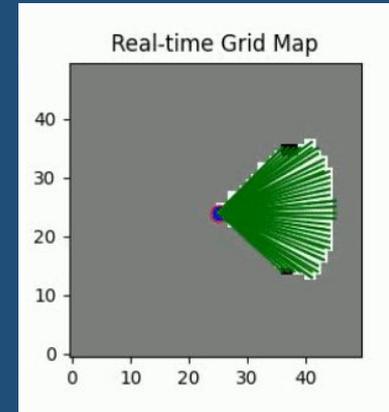
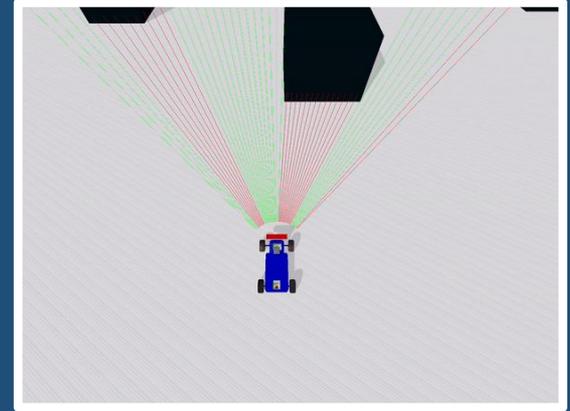
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Simulation

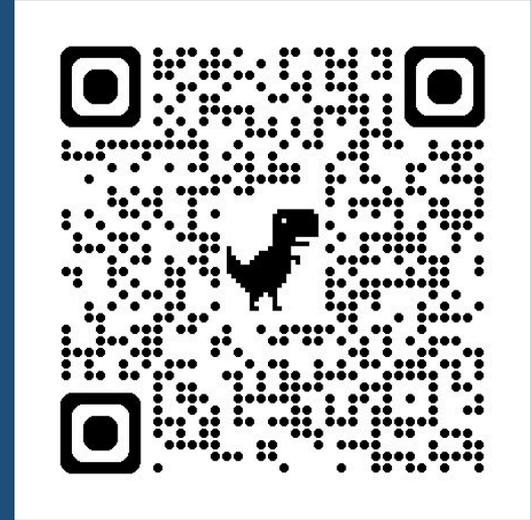
Simulations

- Real life has unpredictable factors
- Simulations allow for reliable code testing
- Ex: Pybullet, IsaacGym, Gazebo, ManiSkill, Mujoco



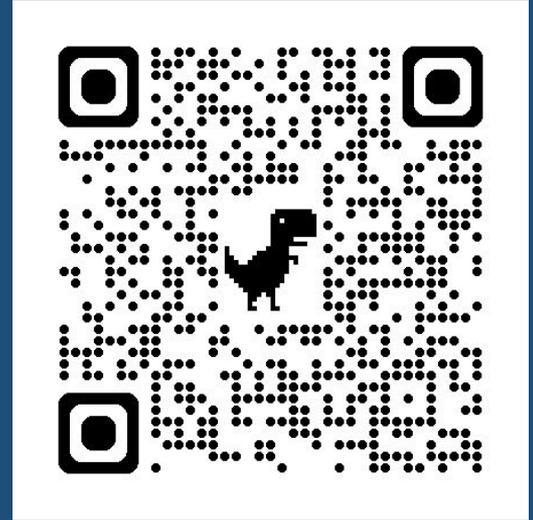
PyBullet Activity

- 1) On your desktop, navigate to the “Workshop” page of the website.
- 2) Download the zip folder corresponding to your OS. Then, unzip the folder.
- 3) Follow the instructions on the README. Total installation time is around 10 minutes.
- 4) Paste these functions (bresenham_line, computer_lidar_hitpoints, prob_to_logodds) you created in the collab into functions.py
- 5) Run the python file pybullet_map.py
- 6) Use the arrow keys to move and observe a separate window with your OGM map



<https://existentialrobotics.org/RobotProvingGrounds/>

PyBullet Activity



<https://existentialrobotics.org/RobotProvingGrounds/>

Post-Attendance Survey



<https://forms.gle/SMgxSiuvW9dSz8YPA>

Website/Google Collabs



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Q&A

Any Questions?