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# Lidar Simulation

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**Aug 14, 2023**



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This repository contains a Lidar simulator that generates Lidar scans in different environments either based on 3D models or mazes in ASCII format.

We use these Lidar scans to test different algorithm for perception, localization and path finding.

## Known Execution Errors

As generator for STL files we will use [Build123d](https://build123d.readthedocs.io/en/latest/). This tool is not installed when generation the jupyter book, hence Square, Race Track and Box will show execution errors in the jupyter notebook

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**OBJECTIVE**

I was looking for a simulation environment to generate Lidar output to test different navigation and localization algorithm. Gazebo does not work for my purposes as it does not run on a Mac and I was not able to find a suitable replacement that meets my needs.

The original idea was to use a bitmap image, convert this into a point cloud and transform the cloud map to a polar coordinate system to generate a Lidar scan.

Turns out that there are always corner cases where a ray through the point cloud that represents a wall will not hit any of the points, the laser “sees through the wall”

To be able to do the calculation required for a simulated laser scan we need a line to represent the wall. A 3D mesh is a collection of triangles and there are many software tools available to generate such a 3d mesh, we will therefore use the stl file format as input.

For now we restrict our studies to two dimensional lidar scans. When dealing with a 3D STL file we will filter all the faces on the ground to have a 2D representation.

It is also straight forward to read a file with a bitmap image like pgm and convert each pixel to two triangles but this will lead to an unnecessary amount of triangles, we will only implement the import of pgm files when we have a specific need for it.

An other input file is using ASCII art as this is used to store maze layouts for different micromouse competition. We will develop a converter for this file format.

As generator for STL files we will use Build123d To run the jupyter notebooks that uses Build123d you need an environment with Build123d installed and that will require the use of conda. However to use STL files we can work with an environment where all dependencies can be installed with pip.





## CREATING 3D MODELS

This section contains a few example of generated environments using different tools.



## SQUARE

We use Build123d to create square with four walls around

```
from jupyter_cadquery import show, open_viewer, set_defaults
import cadquery as cq
from build123d import *
cv = open_viewer("Build123d", cad_width=770, glass=True)
set_defaults(edge_accuracy=0.0001)
```

```
-----
ModuleNotFoundError                                Traceback (most recent call last)
Cell In[1], line 1
----> 1 from jupyter_cadquery import show, open_viewer, set_defaults
      2 import cadquery as cq
      3 from build123d import *

ModuleNotFoundError: No module named 'jupyter_cadquery'
```

```
length = 1000
width = 551

wall_high = 20
wall_thick = 2

with BuildPart() as p:
    Box(length, width, wall_high)
    Box(length - wall_thick, width - wall_thick, wall_high, mode=Mode.SUBTRACT)
```

```
assembly = Compound(children=[p.part])
```

```
assembly
```

```
assembly.export_stl('rectangle.stl')
```

```
True
```



## RACE TRACK

We use Build123d to create a 3D model of a race track. We export the 3D model as a stl file.

We also export a 2D projection as svg file. We use image magic to convert the svg file to a png file and then to convert the png file to a pgm file. Adding some meta data will provide us with a map that we can use in Nav2

```
from jupyter_cadquery import show, open_viewer, set_defaults
import cadquery as cq
from build123d import *
cv = open_viewer("Build123d", cad_width=770, glass=True)
set_defaults(edge_accuracy=0.0001)
```

```
-----
ModuleNotFoundError                                Traceback (most recent call last)
Cell In[1], line 1
----> 1 from jupyter_cadquery import show, open_viewer, set_defaults
      2 import cadquery as cq
      3 from build123d import *

ModuleNotFoundError: No module named 'jupyter_cadquery'
```

```
class SideWall():

    def __init__(self, whichWall):
        ext_r = 150
        int_r = 50
        length = 1000
        width = 551

        len1 = 700
        arc1 = 180
        len2 = 150
        arc2 = -90
        len3 = 50
        arc3 = 360 - 168.2
        len3 = 50
        len4 = 58.78
        arc4 = -122.2
        len5 = 103.05
        arc5 = 360 - 159.6

        wall_height = 20
        wall_thick = 2
```

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```

    if whichWall == 'ext':
        r1 = ext_r
        r2 = int_r
    else:
        r1 = int_r
        r2 = ext_r

    with BuildPart() as self.p:
        with BuildLine(mode=Mode.PRIVATE) as l:
            base_line = Line((0,0),(len1,0))
            righth_curve = JernArc(start=base_line @ 1, tangent=base_line % 1,
↳radius=r1, arc_size=arc1)
            line2 = PolarLine(righth_curve @ 1, len2, direction=righth_curve % 1)
            righth_upper_curve = JernArc(start=line2 @ 1, tangent=line2 % 1,
↳radius=r2, arc_size=arc2)
            line3 = PolarLine(righth_upper_curve @ 1, len3, direction=righth_upper_
↳curve%1)
            upper_curve = JernArc(start=line3 @ 1, tangent=line3 % 1, radius=r1,
↳arc_size=arc3)
            line4 = PolarLine(upper_curve @ 1, len4, direction=upper_curve%1)
            righth_upper_curve = JernArc(start=line4 @ 1, tangent=line4%1,
↳radius=r2, arc_size=arc4)
            line5 = PolarLine(righth_upper_curve @ 1, len5, direction=righth_upper_
↳curve%1)
            left_curve = JernArc(start=line5 @ 1, tangent=line5%1, radius=r1, arc_
↳size=arc5)
            with BuildSketch(Plane.YZ) as s:
                Rectangle(wall_thick, wall_high)
                sweep(path=l.wires()[0])

```

```

p_int = SideWall('int')
p_ext = SideWall('ext')

```

```

p_int.p.part.label = "internal wall"
p_int.p.part.location = Location((0, 100, 0))
p_ext.p.part.label = "external wall"

racetrack_assembly = Compound(label="racetrack", children=[p_int.p.part, p_ext.p.
↳part])

```

```
racetrack_assembly
```

```
100% :-----: (2/2) 0.06s
```

```
racetrack_assembly.export_stl('racetrack.stl')
```

```
True
```

```

racetrack_assembly.export_svg('racetrack.svg', (0, 0, 1000000), (0, 1, 0),
    svg_opts={"pixel_scale": 1, "margin_left": 0, "margin_top":
↳0, "show_axes": False, "show_hidden": False})

```

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```
import xml.etree.ElementTree as ET
tree = ET.parse('racetrack.svg')
root = tree.getroot()
dim = root.attrib
```

```
!convert racetrack.svg racetrack.png
```

```
!convert racetrack.png -flatten racetrack.pgm
```

```
metadata = {
    "image": "racetrack.pgm",
    "resolution": 0.01,
    "origin": [0.0, 0.0, 0.0],
    "occupied_thresh": 0.65,
    "free_thresh": 0.196,
    "negate": 0
}
```

```
metadata
```

```
{'image': 'racetrack.pgm',
 'resolution': 0.01,
 'origin': [0.0, 0.0, 0.0],
 'occupied_thresh': 0.65,
 'free_thresh': 0.196,
 'negate': 0}
```

```
import yaml
with open("racetrack.yaml", "w") as fh:
    yaml.dump(metadata, fh)
```





We build a simple box that we can use as obstacle

```
from jupyter_cadquery import show, open_viewer, set_defaults
import cadquery as cq
from build123d import *
cv = open_viewer("Build123d", cad_width=770, glass=True)
set_defaults(edge_accuracy=0.0001)
```

```
-----
ModuleNotFoundError                                Traceback (most recent call last)
Cell In[1], line 1
----> 1 from jupyter_cadquery import show, open_viewer, set_defaults
      2 import cadquery as cq
      3 from build123d import *

ModuleNotFoundError: No module named 'jupyter_cadquery'
```

```
length = 30
width = 10
hight = 20

with BuildPart() as p:
    Box(length, width, hight)
```

```
assembly = Compound(children=[p.part])
```

```
assembly
```

```
assembly.export_stl('box.stl')
```

```
True
```



## MAZE GENERATION

We will convert ASCII files as used in mazefiles into a suitable format

```
%matplotlib inline
from stl import mesh
import numpy as np
import matplotlib.pyplot as plt

def read_maze(file_name):
    WH_BIT = 1
    WV_BIT = 2

    with open(file_name, 'r') as fh:
        txt = fh.read()

    txt = txt.splitlines()
    post_char = txt[0][0]

    wall_horizontal = txt[::2]
    wall_horizontal = [
        [
            1 if wall == '---' else 0
            for wall in row.lstrip(post_char).split(post_char)
        ]
        for row in wall_horizontal
    ]
    wall_horizontal = np.array(wall_horizontal).astype('uint8')

    wall_vertical = txt[1::2]
    wall_vertical = [
        [1 if wall == '|' else 0 for wall in column[::4]]
        for column in wall_vertical
    ]
    wall_vertical.append([0] * 17)
    wall_vertical = np.array(wall_vertical).astype('uint8')

    wall_horizontal *= WH_BIT
    wall_vertical *= WV_BIT

    return wall_horizontal + wall_vertical
```

```
m = read_maze('maze_in/maze.txt')
```



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```

for line in maze:
    is_wall_x = False
    index_x = 0
    for l in line:
        if l & 1:
            if not is_wall_x:
                wall_x_start = [index_x, index_y]
                is_wall_x = True
            else:
                if is_wall_x:
                    wall_x_end = [index_x, index_y]
                    walls.append([wall_x_start, wall_x_end])
                    is_wall_x = False
        if l & 2:
            if not is_wall_y[index_x]:
                wall_y_start[index_x] = [index_x, index_y]
                is_wall_y[index_x] = True
            else:
                if is_wall_y[index_x]:
                    wall_y_end = [index_x, index_y]
                    walls.append([wall_y_start[index_x], wall_y_end])
                    is_wall_y[index_x] = False
        index_x += 1
    index_y += 1
return np.array(walls)

```

```
walls = find_walls(m)
```

```

data = np.zeros(2*walls.shape[0], dtype=mesh.Mesh.dtype)
index = 0
for wall in walls:
    y_dim = 16
    space = 50
    if wall[0, 0] == wall[1, 0]:
        p0 = [space * wall[0, 0], space * (y_dim - wall[0, 1]), 0]
        p1 = [space * wall[1, 0], space * (y_dim - wall[1, 1]), 0]
        p2 = [space * wall[0, 0], space * (y_dim - wall[0, 1]) + 2, 0]
        p3 = [space * wall[1, 0], space * (y_dim - wall[1, 1]) + 2, 0]
    else:
        p0 = [space * wall[0, 0], space * (y_dim - wall[0, 1]), 0]
        p1 = [space * wall[1, 0], space * (y_dim - wall[1, 1]), 0]
        p2 = [space * wall[0, 0] + 2, space * (y_dim - wall[0, 1]), 0]
        p3 = [space * wall[1, 0] + 2, space * (y_dim - wall[1, 1]), 0]
    data['vectors'][index] = np.array([p0, p1, p2])
    data['vectors'][index + 1] = np.array([p1, p2, p3])
    index += 2

env = mesh.Mesh(data, remove_empty_areas=False)
env.update_units()

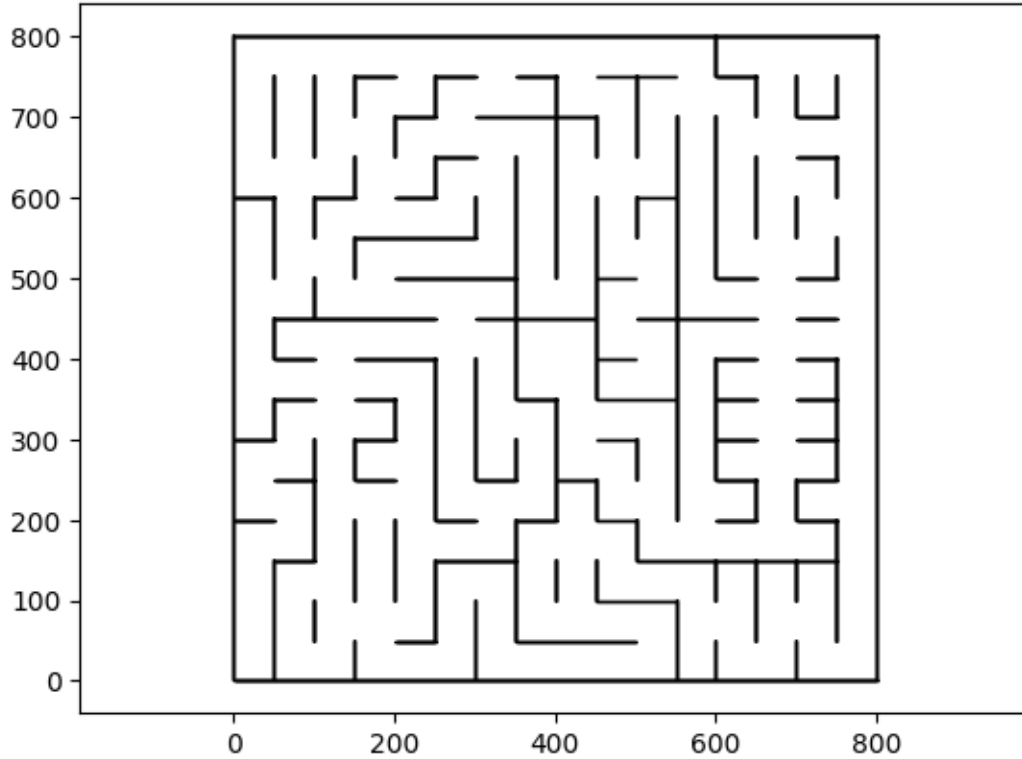
```

```

triangles = []
for t in env.vectors:
    triangles.append(t)

```

```
fig, ax = plt.subplots(1, 1)
ax.axis('equal')
for t in triangles:
    ax.fill(t[:, 0], t[:, 1], fill=False)
```



```
data = np.zeros(12*walls.shape[0], dtype=mesh.Mesh.dtype)
index = 0
for wall in walls:
    y_dim = 16
    space = 50
    if wall[0, 0] == wall[1, 0]:
        p0 = [space * wall[0, 0], space * (y_dim - wall[0, 1]), 0]
        p1 = [space * wall[1, 0], space * (y_dim - wall[1, 1]), 0]
        p2 = [space * wall[0, 0] + 2, space * (y_dim - wall[0, 1]), 0]
        p3 = [space * wall[1, 0] + 2, space * (y_dim - wall[1, 1]), 0]
        p4 = [space * wall[0, 0], space * (y_dim - wall[0, 1]), 20]
        p5 = [space * wall[1, 0], space * (y_dim - wall[1, 1]), 20]
        p6 = [space * wall[0, 0] + 2, space * (y_dim - wall[0, 1]), 20]
        p7 = [space * wall[1, 0] + 2, space * (y_dim - wall[1, 1]), 20]
    else:
        p0 = [space * wall[0, 0], space * (y_dim - wall[0, 1]), 0]
        p1 = [space * wall[1, 0], space * (y_dim - wall[1, 1]), 0]
        p2 = [space * wall[0, 0], space * (y_dim - wall[0, 1]) + 2, 0]
        p3 = [space * wall[1, 0], space * (y_dim - wall[1, 1]) + 2, 0]
        p4 = [space * wall[0, 0], space * (y_dim - wall[0, 1]), 20]
        p5 = [space * wall[1, 0], space * (y_dim - wall[1, 1]), 20]
        p6 = [space * wall[0, 0], space * (y_dim - wall[0, 1]) + 2, 20]
        p7 = [space * wall[1, 0], space * (y_dim - wall[1, 1]) + 2, 20]
```

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```
data['vectors'][index] = np.array([p1, p0, p2])
data['vectors'][index + 1] = np.array([p1, p2, p3])
data['vectors'][index + 2] = np.array([p4, p5, p6])
data['vectors'][index + 3] = np.array([p6, p5, p7])
data['vectors'][index + 4] = np.array([p0, p1, p4])
data['vectors'][index + 5] = np.array([p4, p1, p5])
data['vectors'][index + 6] = np.array([p3, p2, p6])
data['vectors'][index + 7] = np.array([p3, p6, p7])
data['vectors'][index + 8] = np.array([p2, p0, p4])
data['vectors'][index + 9] = np.array([p2, p4, p6])
data['vectors'][index + 10] = np.array([p1, p3, p5])
data['vectors'][index + 11] = np.array([p5, p3, p7])
index += 12
```

```
env = mesh.Mesh(data, remove_empty_areas=False)
env.update_units()
```

```
env.save('maze.stl')
```





## SHOW ALL MODELS

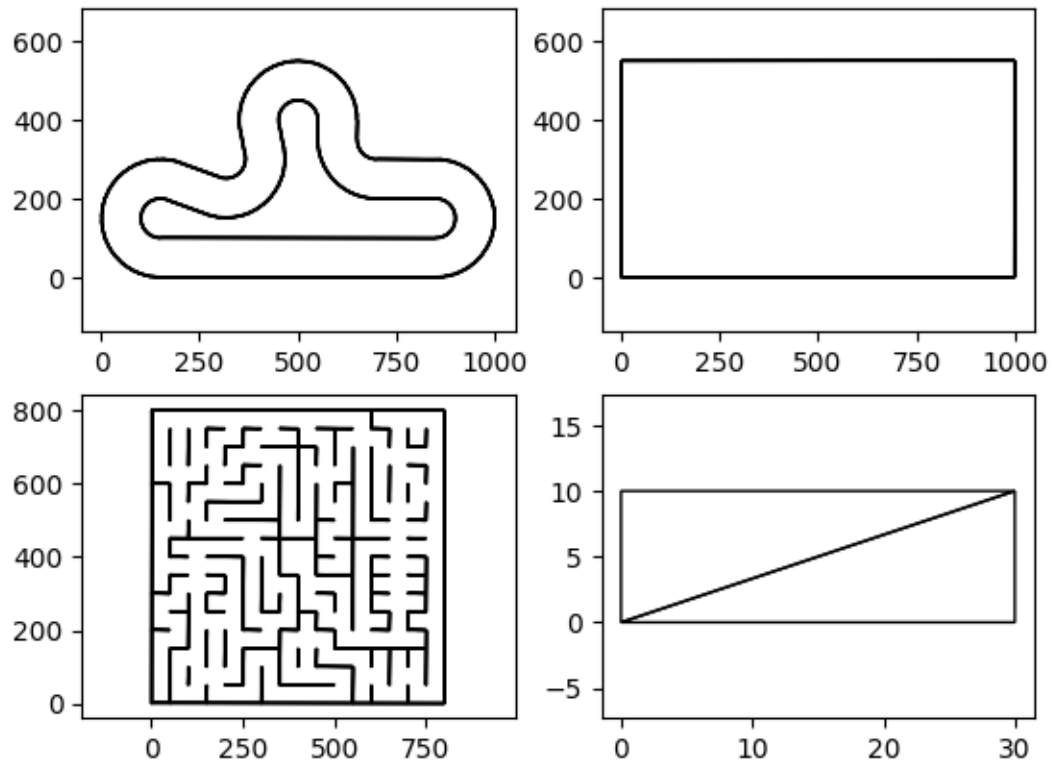
This requires that the python module LidarSim is installed

```
from LidarSim.lidar_sim import LidarSimulator
import matplotlib.pyplot as plt
%matplotlib inline
```

```
import glob
all_models = glob.glob("*.stl")
```

```
columns = 2
rows = int(len(all_models) / columns)
if len(all_models) % columns:
    rows += 1
```

```
fig, axs = plt.subplots(rows, columns)
for i in range(len(all_models)):
    row_index = int(i / columns)
    column_index = i % columns
    lidar = LidarSimulator(all_models[i])
    triangles = lidar.get_map_triangles()
    axs[row_index, column_index].axis('equal')
    for t in triangles:
        axs[row_index, column_index].fill(t[:, 0], t[:, 1], fill=False)
```



## SIMULATING A LIDAR

We read our stl file using numpy-stl.

We select all triangles in the xy plane to get a 2D projection

We then place the robot with the lidar on the map using xy coordinates and yaw as the orientation. We transform the points of the map to the lidar coordinate system and then convert the cartesian coordinate system to a polar coordinate system. We use a simplified robot that consists of only a point

We sweep a full circle around the Lidar in segments which size depends on the Lidar resolution and we retain the closest point to construct our lidar scan.

```
%matplotlib inline
import numpy as np
from stl import mesh
import matplotlib.pyplot as plt
```

```
min_range = 2.0
max_range = 12000.0
resolution = 1 # in degrees
```

```
#simulated_environment = 'racetrack'
simulated_environment = 'square'
#simulated_environment = 'test'
```

```
if simulated_environment == 'test':
    data = np.zeros(3, dtype=mesh.Mesh.dtype)
    data['vectors'][0] = np.array([[100., 500., 0],
                                   [100., 100., 0],
                                   [ 0.,  0., 0]])
    data['vectors'][1] = np.array([[ 900., 100., 0],
                                   [1000.,  0., 0],
                                   [ 0.,  0., 0]])
    data['vectors'][2] = np.array([[1000., 500., 0],
                                   [1000., 200., 0],
                                   [ 100., 500., 0]])
    env = mesh.Mesh(data, remove_empty_areas=False)
    env.update_units()
else:
    env = mesh.Mesh.from_file('%s.stl' % simulated_environment)
```

```

-----
FileNotFoundError                                Traceback (most recent call last)
Cell In[4], line 15
    13     env.update_units()
    14 else:
--> 15     env = mesh.Mesh.from_file('%s.stl' % simulated_environment)

File ~/virtualenvs/lidar_sim/lib/python3.11/site-packages/stl/stl.py:389, in
BaseStl.from_file(cls, filename, calculate_normals, fh, mode, speedups, **kwargs)
    385     name, data = cls.load(
    386         fh, mode=mode, speedups=speedups
    387     )
    388 else:
--> 389     with open(filename, 'rb') as fh:
    390         name, data = cls.load(
    391             fh, mode=mode, speedups=speedups
    392         )
    394 return cls(
    395     data, calculate_normals, name=name,
    396     speedups=speedups, **kwargs
    397 )

FileNotFoundError: [Errno 2] No such file or directory: 'square.stl'

```

## 8.1 Adjust coordinate system

```

env.x = env.x - env.x.min()
env.y = env.y - env.y.min()

```

## 8.2 Select bottom Face

```

subset = env.vectors[(env.normals[:, 0] == 0.0) & (env.normals[:, 1] == 0.0) & (env.
    normals[:, 2] < 0.0)]

```

## 8.3 Set initial Pose

```

pose = [900, 50, np.pi]
pose = [500, 50, np.radians(-45)]
#pose = [500, 50, 0]
#pose = [850, 52, 3.0892327760299634]
#pose = [753, 77, 2.792526803190927]
#pose = [ 166.648113,  213.283612,  0.20943951]
#pose = [ 215.555493,  223.679196,  0.20943951]
#pose = [ 260.735529,  234.518542,  0.2268928 ]
#pose = [107.199835, 189.184783, 1.30899694]
pose = [107, 189, 1.30899694]

```

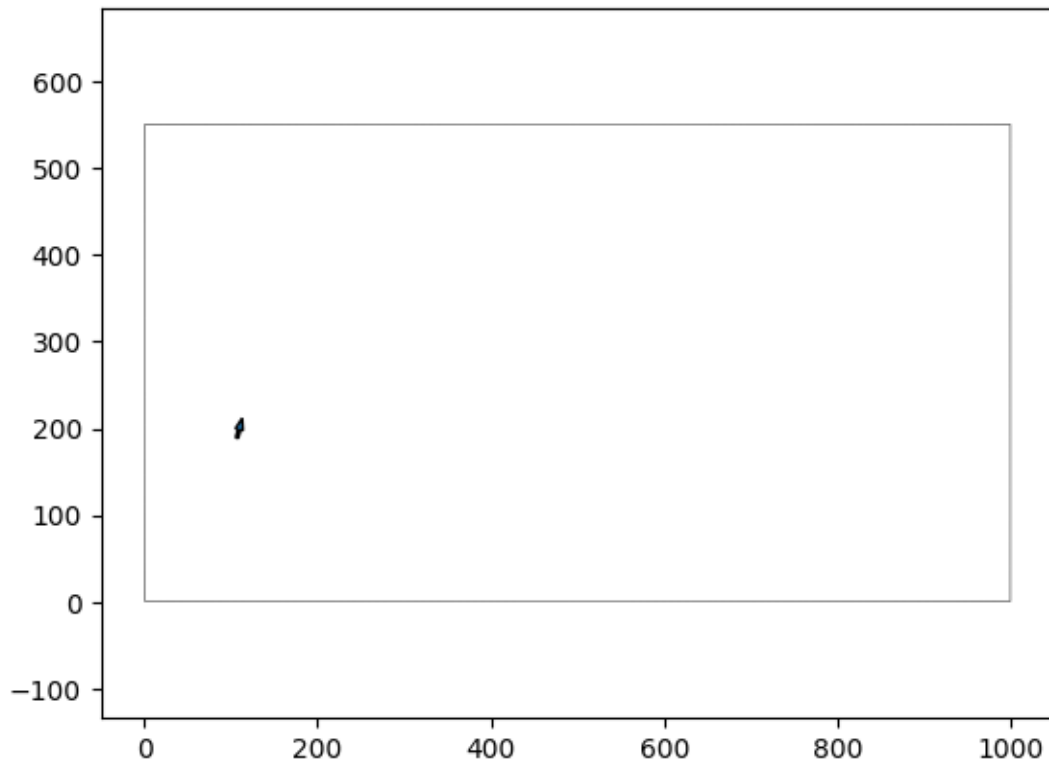
```

from matplotlib.patches import Polygon

fig, ax = plt.subplots()
for y in subset[:, :, 0:2]:
    p = Polygon(y, facecolor = 'k')
    ax.add_patch(p)
ax.axis('equal')
plt.arrow(pose[0], pose[1], 10 * np.cos(pose[2]), 10 * np.sin(pose[2]), width=3.0)

```

```
<matplotlib.patches.FancyArrow at 0x1140a6bd0>
```



## 8.4 Transform to Lidar Frame

```

env.x = env.x - pose[0]
env.y = env.y - pose[1]
env.rotate([0.0, 0.0, 1.0], pose[2])

```

```

subset = env.vectors[(env.normals[:, 0] == 0.0) & (env.normals[:, 1] == 0.0) & (env.
    normals[:, 2] < 0.0)]

```

```

fig, ax = plt.subplots()
for y in subset[:, :, 0:2]:
    p = Polygon(y, facecolor = 'k')
    ax.add_patch(p)

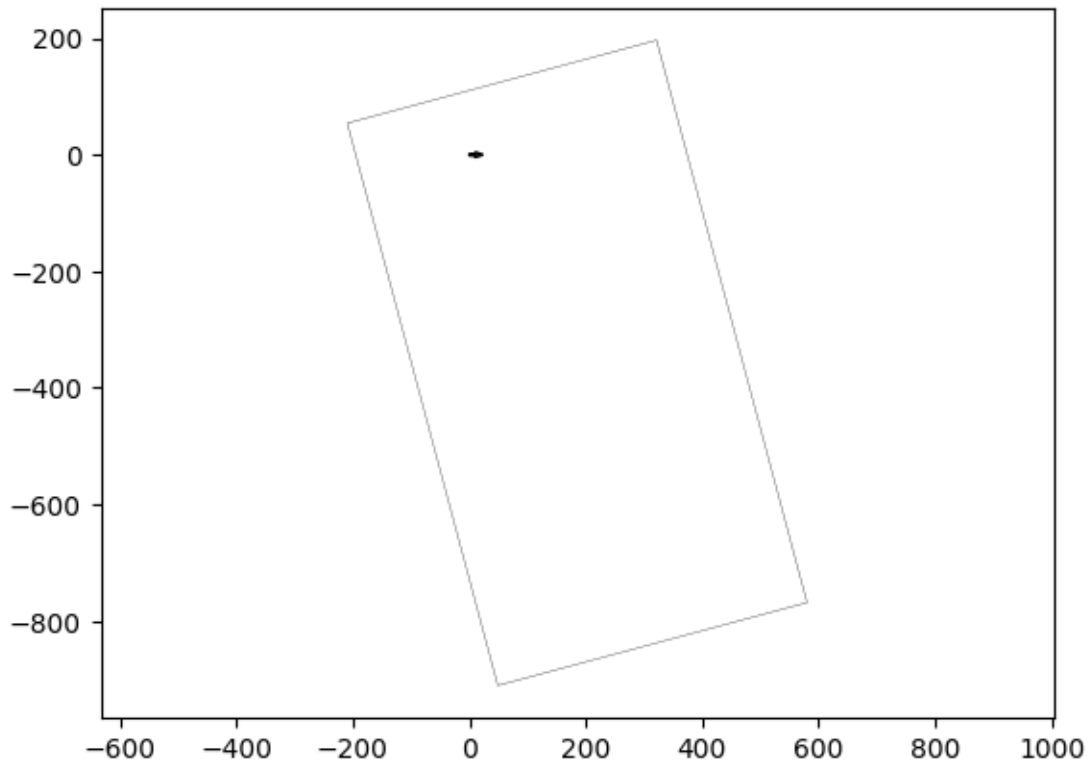
```

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```
ax.axis('equal')
plt.arrow(0, 0, 10, 0, width=3.0)
```

```
<matplotlib.patches.FancyArrow at 0x114323350>
```



## 8.5 Convert to polar coordinate system

```
import cmath
def cart2polC(xyz):
    x, y, z = xyz
    return (cmath.polar(complex(x, y))) # rho, phi
def convert_array(arr):
    theta = []
    r = []
    for x in arr:
        rho, phi = cart2polC(x)
        theta.append(phi)
        r.append(rho)
    return theta, r
```

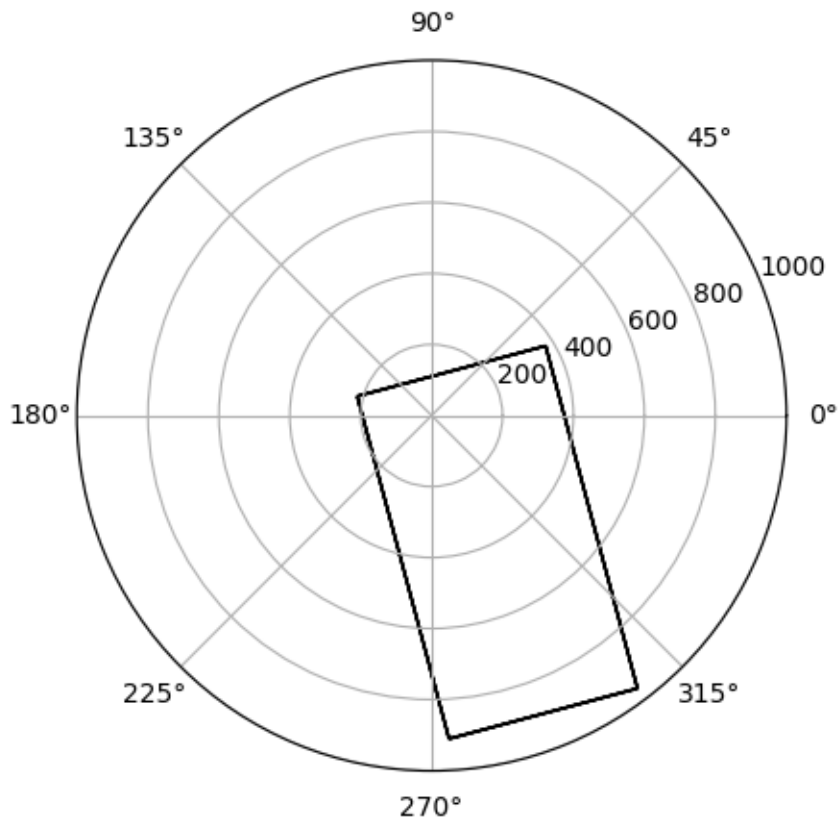
```
triangles = []
for t in subset[:, :, :]:
    a = np.array(convert_array(t)).transpose()
```

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```
triangles.append(a[a[:, 0].argsort()])
triangles = np.array(triangles)
```

```
fig, ax = plt.subplots(subplot_kw={'projection': 'polar'})
for t in triangles:
    ax.fill(t[:, 0], t[:, 1], fill=False)
```



## 8.6 Filter closest points

### 8.6.1 Lines in polar coordinates

A line in cartesian coordinate system is described as

$$y = m * x + b$$

We make the following substitution:

$$x = r * \cos(\theta)$$

$$y = r * \sin(\theta)$$

Given two points in polar coordinates

$$P_1 = (\theta_1, r_1)$$

$$P_2 = (\theta_2, r_2)$$

we can calculate m and b. Finally we can calculate the intersection of the line connecting  $P_1$  and  $P_2$  with the line from the origin and an angle of  $\theta$

```
import sympy as sp
from sympy.abc import theta
```

```
x, y = sp.symbols("x, y")
m, b, r, r1, r2 = sp.symbols("m, b, r, r1, r2", real=True)
theta_1, theta_2 = sp.symbols('theta_1, theta_2')
```

```
expr = y - m*x - b
expr = expr.subs(x, r * sp.cos(theta))
expr = expr.subs(y, r * sp.sin(theta))
expr
```

$$-b - mr \cos(\theta) + r \sin(\theta)$$

```
p1 = expr.subs(r, r1).subs(theta, theta_1)
p2 = expr.subs(r, r2).subs(theta, theta_2)
```

```
p2.subs(b, p1 * sp.sin(theta_1)) - m * r1 * sp.cos(theta_1)
```

$$-mr_1 \cos(\theta_1) - mr_2 \cos(\theta_2) - r_1 \sin(\theta_1) + r_2 \sin(\theta_2)$$

```
def get_distance(p1, p2, theta):
    theta_plus = theta
    if (p1[0] > np.pi) or (p2[0] > np.pi):
        # triangles are transformed in filter_triangles
        if theta < 0:
            theta_plus = theta + 2 * np.pi
    p = np.array([p1[0], p2[0]])
    if not (p.min() <= theta_plus) & (theta_plus <= p.max()):
        return max_range + 1.0
    r_s_1 = np.sin(p1[0]) * p1[1]
    r_c_1 = np.cos(p1[0]) * p1[1]
    r_s_2 = np.sin(p2[0]) * p2[1]
    r_c_2 = np.cos(p2[0]) * p2[1]
    m = (r_s_2 - r_s_1) / (r_c_2 - r_c_1)
    b = r_s_1 - m * r_c_1
    dist = b / (np.sin(theta_plus) - m * np.cos(theta_plus))
    return dist
```



## 8.7 Select Triangles hit by Ray

Must have vertices on both sides of the ray

Handle case where triangle is on both sides of the x axis

Handle case where triangle crosses  $\pm \pi$  line

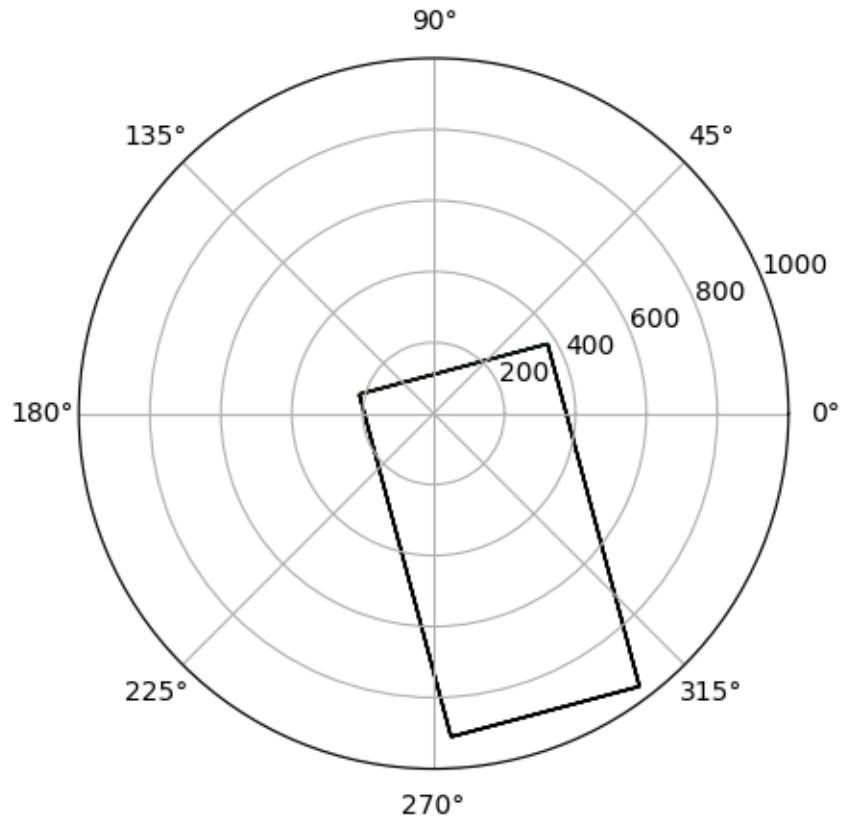
```
def filter_triangles(triangles, theta):
    special_cases = triangles[np.any(triangles >= 0, axis=1)[: ,0] & np.any(triangles
    <-> <0, axis=1)[: , 0]]
    other_cases = triangles[np.invert(np.any(triangles >= 0, axis=1)[: ,0] & np.
    <-> any(triangles <0, axis=1)[: , 0])]
    # vertices on both sides
    triangles_hit = other_cases[np.any(other_cases >= theta, axis=1)[: ,0] & np.
    <-> any(other_cases <= theta, axis=1)[: , 0]]
    # handle special cases:
    sc = []
    for t in special_cases:
        if ((t[: , 0].max() - t[: , 0].min()) < np.pi):
            # not so special after all
            if (theta <= t[: , 0].max()) & (t[: , 0].min() <= theta):
                sc.append(t)
        else:
            for e in t:
                if e[0] < 0:
                    e[0] += 2 * np.pi
            if (theta + 2 * np.pi <= t[: , 0].max()) & (t[: , 0].min() <= theta + 2 * np.
    <-> pi):
                sc.append(t)
            if (theta <= t[: , 0].max()) & (t[: , 0].min() <= theta):
                sc.append(t)

    result = []
    for t in triangles_hit:
        result.append(t)
    for t in sc:
        result.append(t)
    return np.array(result)
```

```
#theta = np.radians(0) # OK
theta = np.radians(90) # OK
#theta = np.radians(-90) # OK
#theta = np.radians(-80) # OK
#theta = np.radians(180) # OK
#theta = np.radians(-135) # OK

triangles_hit = filter_triangles(triangles, theta)
```

```
fig, ax = plt.subplots(subplot_kw={'projection': 'polar'})
for t in triangles:
    ax.fill(t[: , 0], t[: , 1], fill=False)
for t in triangles_hit:
    ax.fill(t[: , 0], t[: , 1], fill=True)
```



```

intersection_points = np.zeros((2, 2))
intersection_points[:, 0] = theta
points = []
dist = get_distance(t[0], t[1], theta)
if dist < max_range:
    points.append(dist)
dist = get_distance(t[0], t[2], theta)
if dist < max_range:
    points.append(dist)
dist = get_distance(t[1], t[2], theta)
if dist < max_range:
    points.append(dist)
if len(points) > 0:
    intersection_points[0, 1] = points[0]
    intersection_points[1, 1] = points[1]
else:
    intersection_points[:, 1] = max_range + 1.0
t = np.zeros((3, 2))

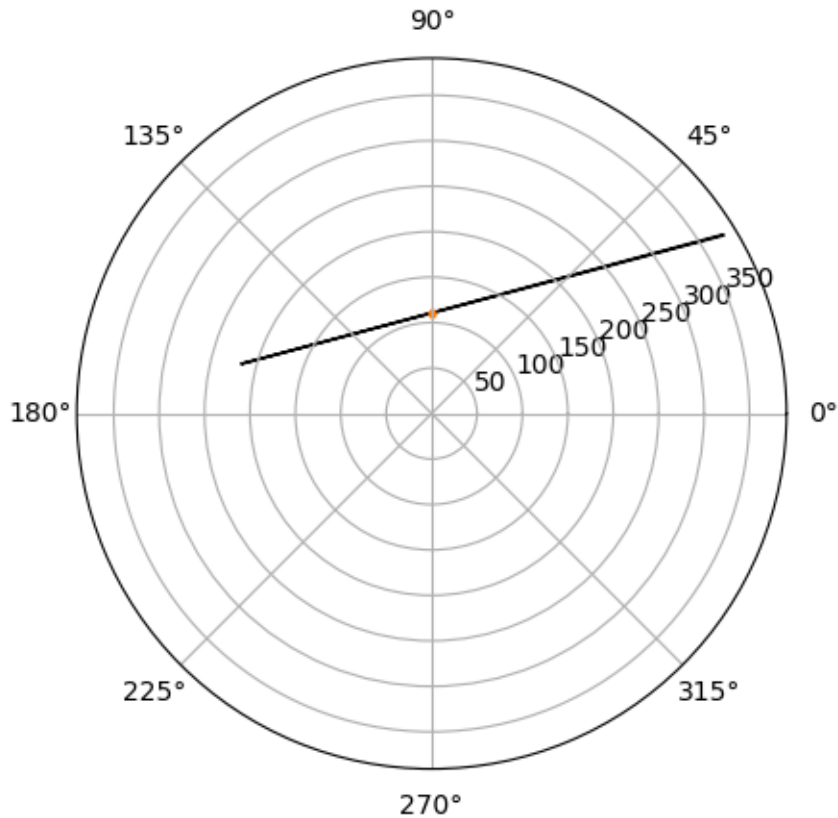
```

```

fig, ax = plt.subplots(subplot_kw={'projection': 'polar'})
ax.fill(t[:, 0], t[:, 1], fill=False)
ax.scatter(intersection_points[:, 0], intersection_points[:, 1], s=4.0)

```

```
<matplotlib.collections.PathCollection at 0x127a1d790>
```



```
def lidar_filter(triangles):
    scan = []
    samples = np.arange(-np.pi, np.pi, np.radians(resolution))
    for sample in samples:
        #start with out of range
        dist = max_range + 1.0
        # select all triangles hit by the ray
        triangles_hit = filter_triangles(triangles, sample)
        for t in triangles_hit:
            dist_t = np.empty(3)
            dist_t[0] = get_distance(t[0], t[1], sample)
            dist_t[1] = get_distance(t[0], t[2], sample)
            dist_t[2] = get_distance(t[1], t[2], sample)
            dist = min(dist_t.min(), dist)
        scan.append(dist)
        if dist > max_range:
            scan[-1] = None
        if dist < min_range:
            scan[-1] = None
    return np.roll(np.array(scan), int(np.pi / np.radians(resolution)))
```

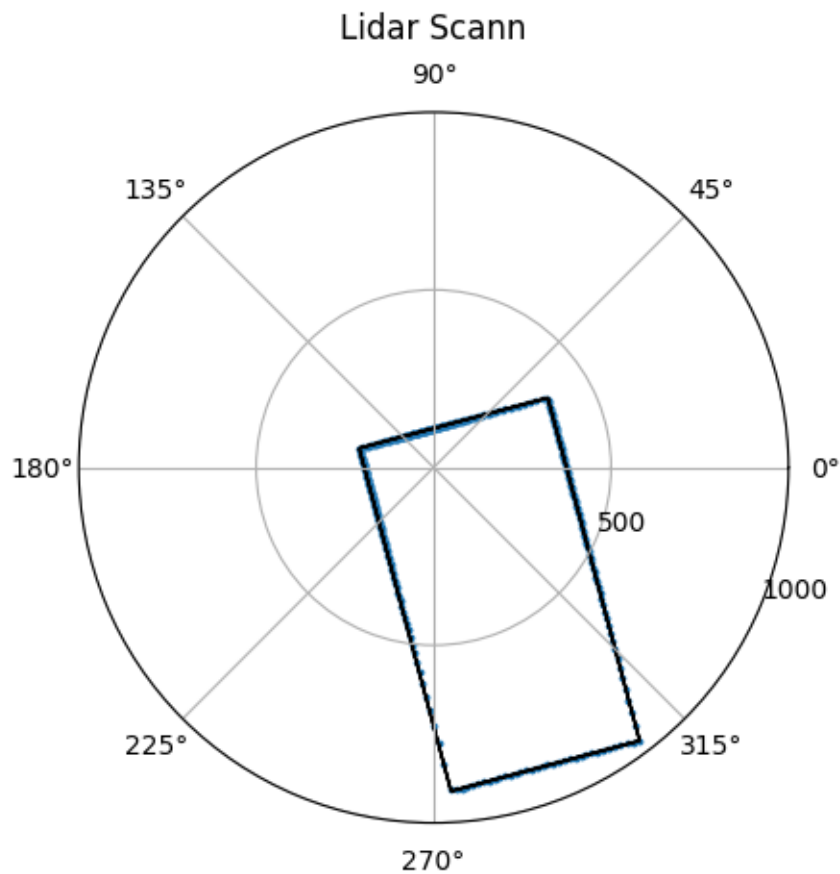
```
lidar_scan = lidar_filter(triangles)
```

```
plot_scan = np.stack((np.arange(0, 2 * np.pi, np.radians(resolution)), lidar_scan),
    ↪axis=1)
plot_scan = plot_scan[plot_scan[:, 1] != np.array(None)]
```

```
fig, ax = plt.subplots(subplot_kw={'projection': 'polar'})
ax.scatter(plot_scan[:, 0], plot_scan[:, 1], s=3.0)
for t in triangles:
    ax.fill(t[:, 0], t[:, 1], fill=False)
ax.set_rmax(1000)
ax.set_rticks([500, 1000]) # Less radial ticks
ax.set_rlabel_position(-22.5) # Move radial labels away from plotted line
ax.grid(True)

ax.set_title("Lidar Scann", va='bottom')
```

```
Text(0.5, 1.0, 'Lidar Scann')
```



```
from LidarSim.lidar_sim import LidarSimulator
```

```
test_lidar = LidarSimulator("racetrack.stl")
point = [107, 189]
yaw = np.radians(45)
plot_scan = test_lidar.get_lidar_points(point[0], point[1], yaw)
triangles = test_lidar.get_env_triangles(point[0], point[1], yaw)
```

```
fig, ax = plt.subplots(subplot_kw={'projection': 'polar'})
ax.scatter(plot_scan[:, 0], plot_scan[:, 1], s=3.0)
```

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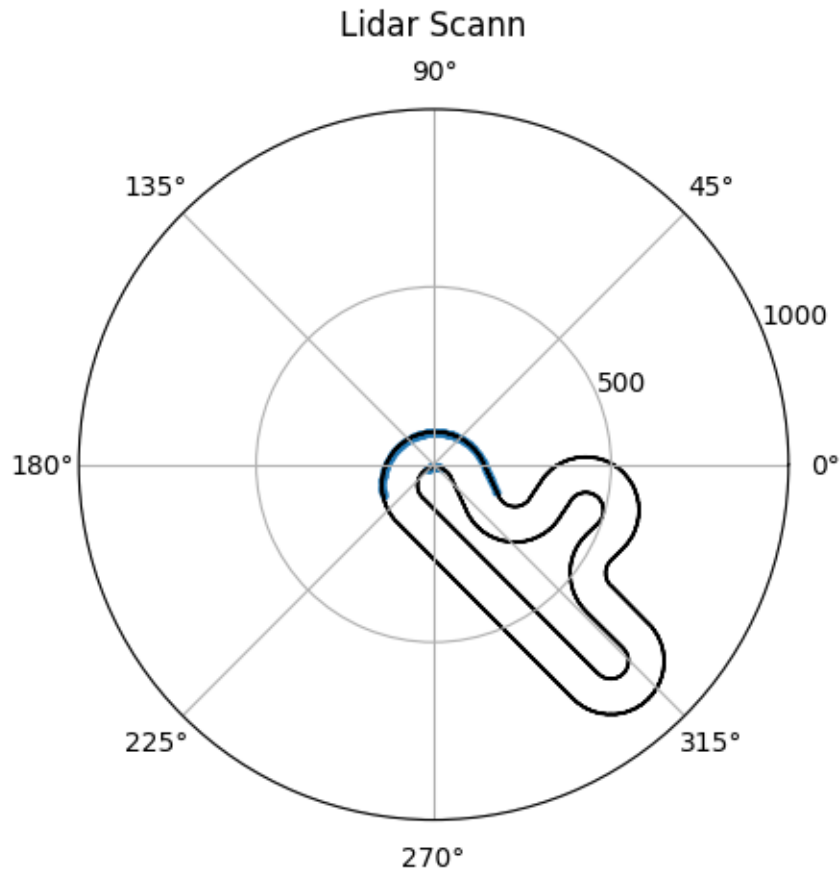
```

for t in triangles:
    ax.fill(t[:, 0], t[:, 1], fill=False)
ax.set_rmax(1000)
ax.set_rticks([500, 1000]) # Less radial ticks
ax.grid(True)

ax.set_title("Lidar Scann", va='bottom')

```

```
Text(0.5, 1.0, 'Lidar Scann')
```



```

test_lidar = LidarSimulator("maze.stl")
point = [25, 25]
yaw = np.radians(90)
plot_scan = test_lidar.get_lidar_points(point[0], point[1], yaw)
triangles = test_lidar.get_env_triangles(point[0], point[1], yaw)

```

```

fig, ax = plt.subplots(subplot_kw={'projection': 'polar'})
ax.scatter(plot_scan[:, 0], plot_scan[:, 1], s=3.0)
for t in triangles:
    ax.fill(t[:, 0], t[:, 1], fill=False)
ax.set_rmax(1000)
ax.set_rticks([500, 1000]) # Less radial ticks
ax.grid(True)

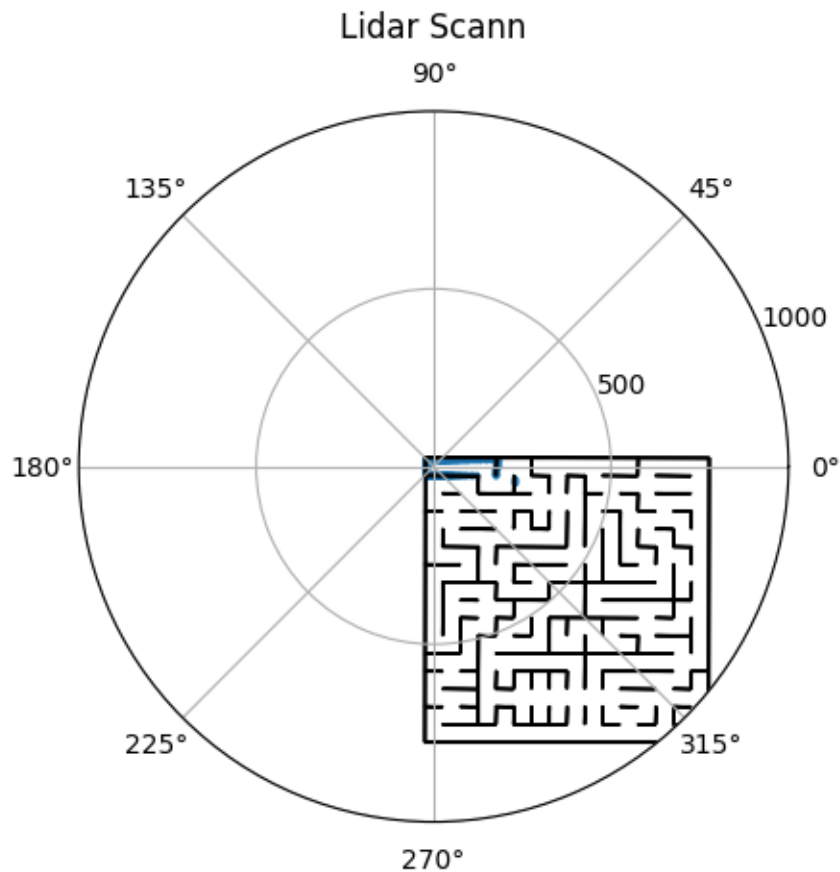
```

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```
ax.set_title("Lidar Scann", va='bottom')
```

```
Text(0.5, 1.0, 'Lidar Scann')
```

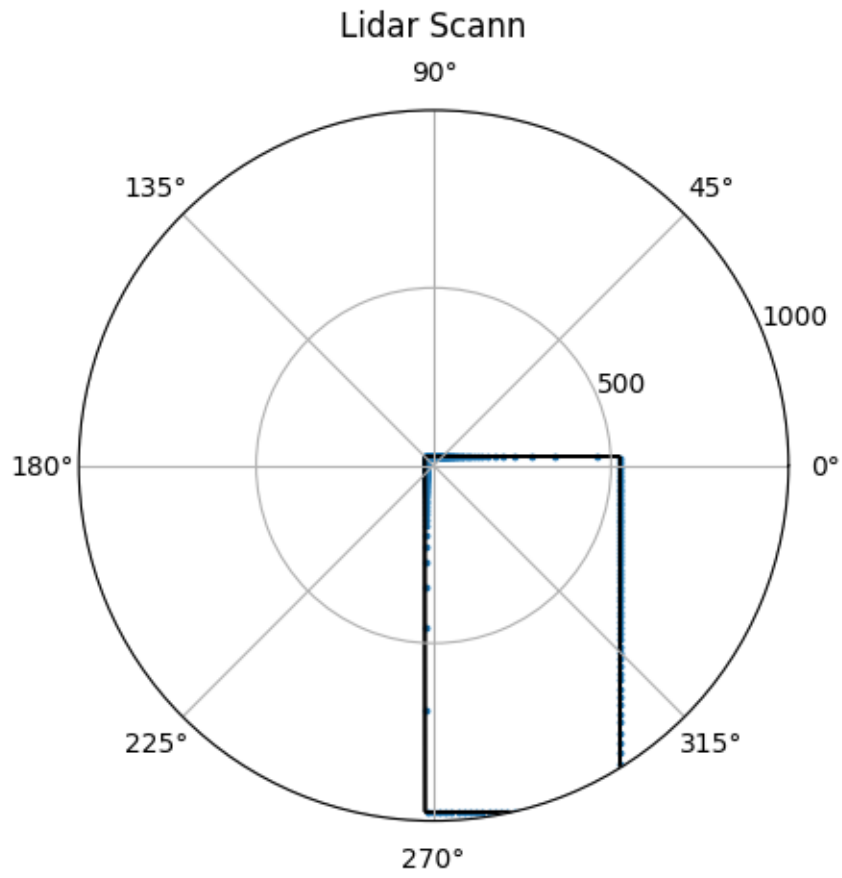


```
test_lidar = LidarSimulator("rectangle.stl")
point = [25, 25]
yaw = np.radians(90)
plot_scan = test_lidar.get_lidar_points(point[0], point[1], yaw)
triangles = test_lidar.get_env_triangles(point[0], point[1], yaw)
```

```
fig, ax = plt.subplots(subplot_kw={'projection': 'polar'})
ax.scatter(plot_scan[:, 0], plot_scan[:, 1], s=3.0)
for t in triangles:
    ax.fill(t[:, 0], t[:, 1], fill=False)
ax.set_rmax(1000)
ax.set_rticks([500, 1000]) # Less radial ticks
ax.grid(True)

ax.set_title("Lidar Scann", va='bottom')
```

```
Text(0.5, 1.0, 'Lidar Scann')
```







## ADDING STATISTICAL ERRORS

```
%matplotlib inline
import matplotlib.pyplot as plt
import numpy as np
```

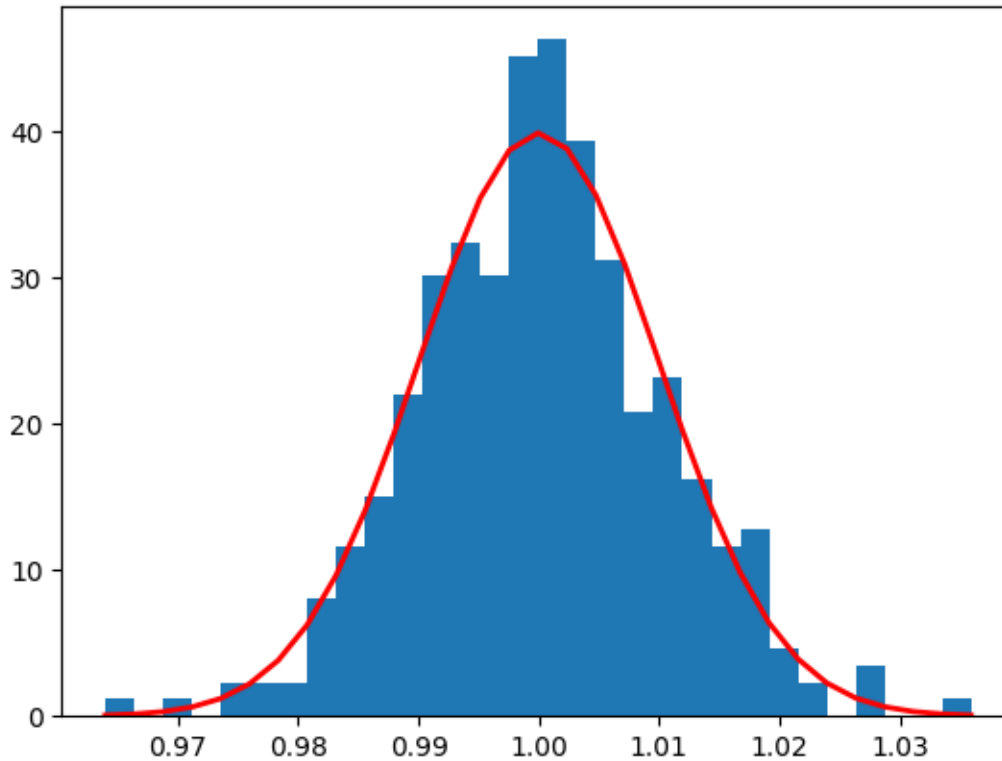
```
from LidarSim.lidar_sim import LidarSimulator
```

```
test_lidar = LidarSimulator("rectangle.stl")
point = [25, 25]
yaw = np.radians(90)
plot_scan = test_lidar.get_lidar_points(point[0], point[1], yaw)
triangles = test_lidar.get_env_triangles(point[0], point[1], yaw)
```

```
mu, sigma = 1.0, 0.01 # mean and standard deviation
s = np.random.normal(mu, sigma, plot_scan.shape[0])
```

```
count, bins, ignored = plt.hist(s, 30, density=True)
plt.plot(bins, 1/(sigma * np.sqrt(2 * np.pi)) *
         np.exp( - (bins - mu)**2 / (2 * sigma**2) ),
         linewidth=2, color='r')
```

```
[<matplotlib.lines.Line2D at 0x109942050>]
```

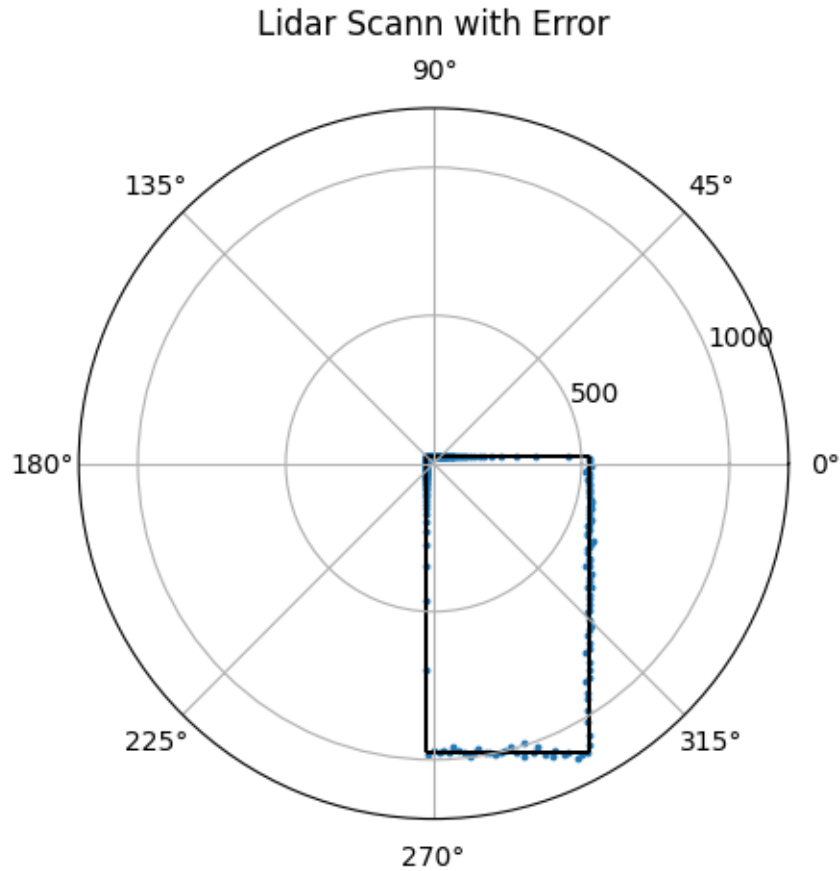


```
plot_scan[:,1] = plot_scan[:,1] * s
```

```
fig, ax = plt.subplots(subplot_kw={'projection': 'polar'})
ax.scatter(plot_scan[:, 0], plot_scan[:, 1], s=3.0)
for t in triangles:
    ax.fill(t[:, 0], t[:, 1], fill=False)
ax.set_rmax(1200)
ax.set_rticks([500, 1000]) # Less radial ticks
ax.grid(True)

ax.set_title("Lidar Scann with Error", va='bottom')
```

```
Text(0.5, 1.0, 'Lidar Scann with Error')
```

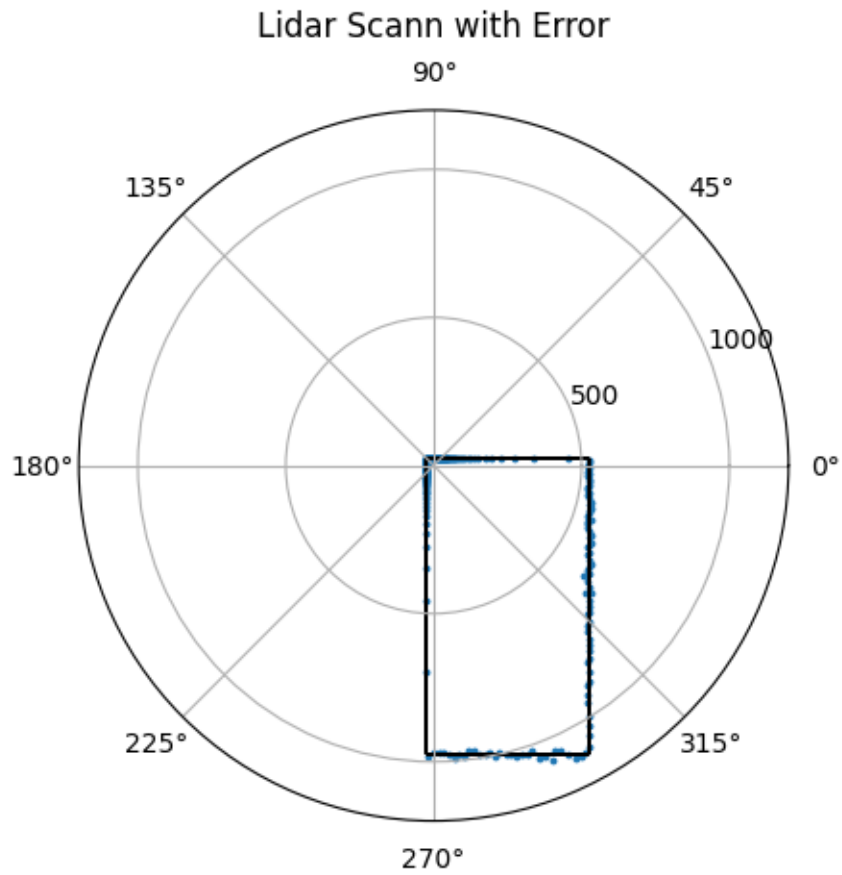


```
test_lidar = LidarSimulator("rectangle.stl", error=0.01)
point = [25, 25]
yaw = np.radians(90)
plot_scan = test_lidar.get_lidar_points(point[0], point[1], yaw)
triangles = test_lidar.get_env_triangles(point[0], point[1], yaw)
```

```
fig, ax = plt.subplots(subplot_kw={'projection': 'polar'})
ax.scatter(plot_scan[:, 0], plot_scan[:, 1], s=3.0)
for t in triangles:
    ax.fill(t[:, 0], t[:, 1], fill=False)
ax.set_rmax(1200)
ax.set_rticks([500, 1000]) # Less radial ticks
ax.grid(True)

ax.set_title("Lidar Scann with Error", va='bottom')
```

```
Text(0.5, 1.0, 'Lidar Scann with Error')
```



## FEATURE EXTRACTION FROM LIDAR

We analyze the data from the Lidar to find simple geometric features like lines or circles.

We must answer questions like:

- how many lines are there
- which point belongs to which line
- what are the characteristics of this line

### 10.1 Field-of-View (FoV)

Our simulated Lidar provides a view of 360 degrees, to extract features we will restrict our field of view to a direction  $\theta$  and a range specified in degrees. Our FoV is then  $\theta$  plus/minus the range

```
%matplotlib inline
import matplotlib.pyplot as plt
import numpy as np
from LidarSim.lidar_sim import LidarSimulator
```

```
lidar = LidarSimulator("rectangle.stl")
point = [500, 300]
yaw = np.radians(0)
plot_scan = lidar.get_lidar_points(point[0], point[1], yaw)
triangles = lidar.get_env_triangles(point[0], point[1], yaw)
```

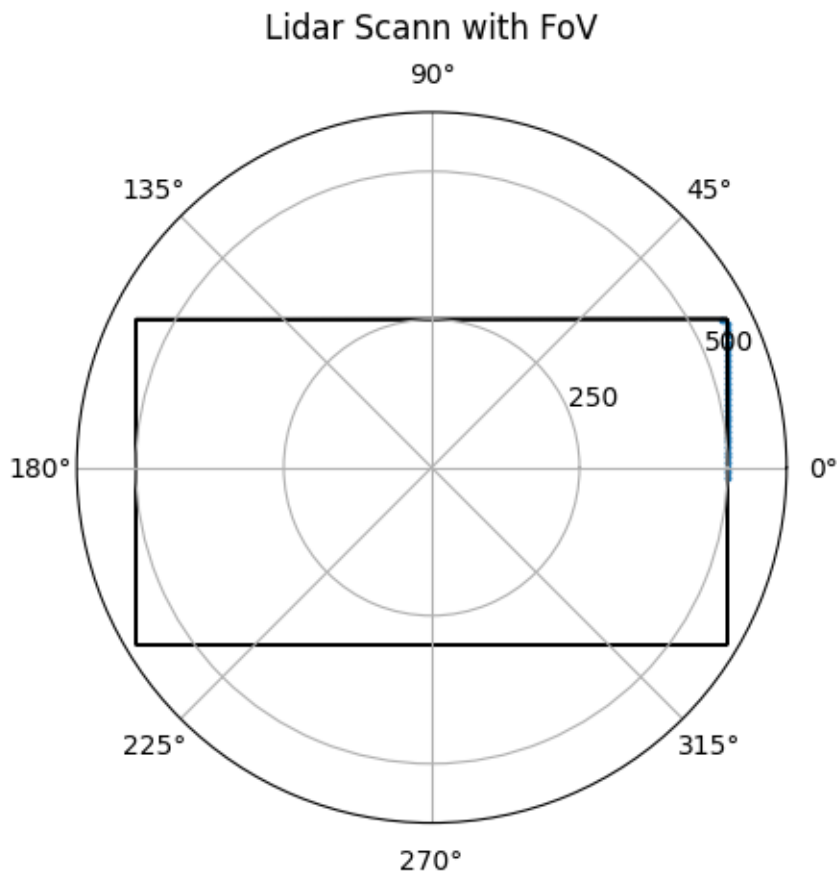
```
def get_fov(theta, view_range, scan, resolution):
    idx = np.searchsorted(scan[:, 0], theta) - 1
    idx_max = int(idx + view_range/resolution)
    idx_min = int(idx - view_range/resolution)
    if idx_max <= scan.shape[0] and idx_min >= 0:
        return scan[idx_min:idx_max]
    if idx_min < 0:
        return np.roll(scan, -idx_min, axis = 0)[:2*int(view_range/resolution)]
    if idx_max > scan.shape[0]:
        return np.roll(scan, -(idx_max - scan.shape[0]), axis = 0)[-2*int(view_range/
    ↪resolution):]
```

```
fov = get_fov(np.radians(14), 15, plot_scan, lidar.resolution)
```

```
fig, ax = plt.subplots(subplot_kw={'projection': 'polar'})
ax.scatter(fov[:, 0], fov[:, 1], s=3.0)
for t in triangles:
    ax.fill(t[:, 0], t[:, 1], fill=False)
ax.set_rmax(600)
ax.set_rticks([250, 500]) # Less radial ticks
ax.grid(True)

ax.set_title("Lidar Scann with FoV", va='bottom')
```

```
Text(0.5, 1.0, 'Lidar Scann with FoV')
```



## 10.2 FoV and LidarSimulator module

This is an example how to set the Fow when calling the LidarSimulator module

```
lidar = LidarSimulator("rectangle.stl")
point = [500, 300]
yaw = np.radians(0)
plot_scan = lidar.get_lidar_points(point[0], point[1], yaw, theta=np.radians(90),
    ↪view_range=30)
triangles = lidar.get_env_triangles(point[0], point[1], yaw)
```

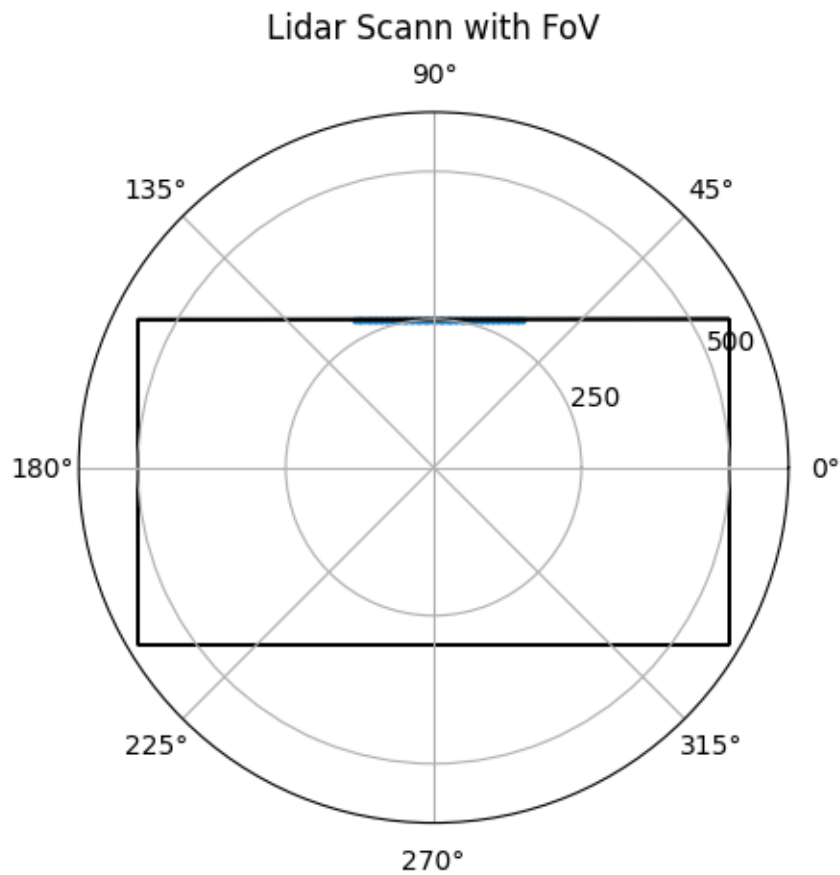
```

fig, ax = plt.subplots(subplot_kw={'projection': 'polar'})
ax.scatter(plot_scan[:, 0], plot_scan[:, 1], s=3.0)
for t in triangles:
    ax.fill(t[:, 0], t[:, 1], fill=False)
ax.set_rmax(600)
ax.set_rticks([250, 500]) # Less radial ticks
ax.grid(True)

ax.set_title("Lidar Scann with FoV", va='bottom')

```

```
Text(0.5, 1.0, 'Lidar Scann with FoV')
```







## PROBLEM STATEMENT

Assuming we have a set of points that all belong to a single line. By choosing two points we can calculate the characteristics of the line and assuming we have no statistical error we can prove that all other points are part of the line

However we will always have measurement errors with Lidar, therefore we need to use statistical methods to fit a line through the points that minimize the error

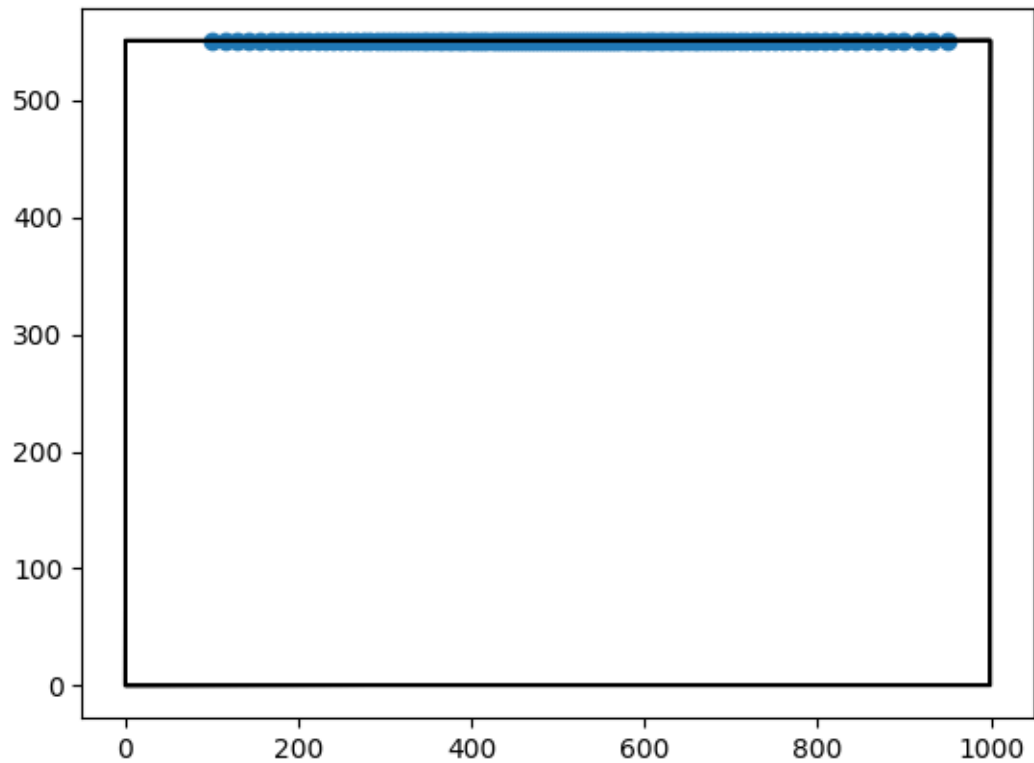
### 11.1 Ideal Situation

```
%matplotlib inline
import matplotlib.pyplot as plt
import numpy as np
from LidarSim.lidar_sim import LidarSimulator
```

```
lidar = LidarSimulator("rectangle.stl")
point = [500, 300]
yaw = np.radians(0)
plot_scan = lidar.get_lidar_points(point[0], point[1], yaw, theta=np.radians(90),
    view_range=60)
triangles = lidar.get_map_triangles()
```

```
# get cartesian coordinates
x = []
y = []
for alpha, r in plot_scan:
    x.append(r * np.cos(alpha) + point[0])
    y.append(r * np.sin(alpha) + point[1])
```

```
fig, ax = plt.subplots()
plt.scatter(x, y)
for t in triangles:
    ax.fill(t[:, 0], t[:, 1], fill=False)
```



```
m = (y[-1] - y[0]) / (x[-1] - x[0])
b = y[-1] - m * x[-1]
y_err = []
for i in range(len(x)):
    y_err.append(y[i] - m * x[i] - b)

y_err = np.array(y_err)
print("mean: %0.2f, median: %0.2f, standard deviation: %0.2f" % (np.mean(y_err), np.
↪median(y_err), np.std(y_err)))
```

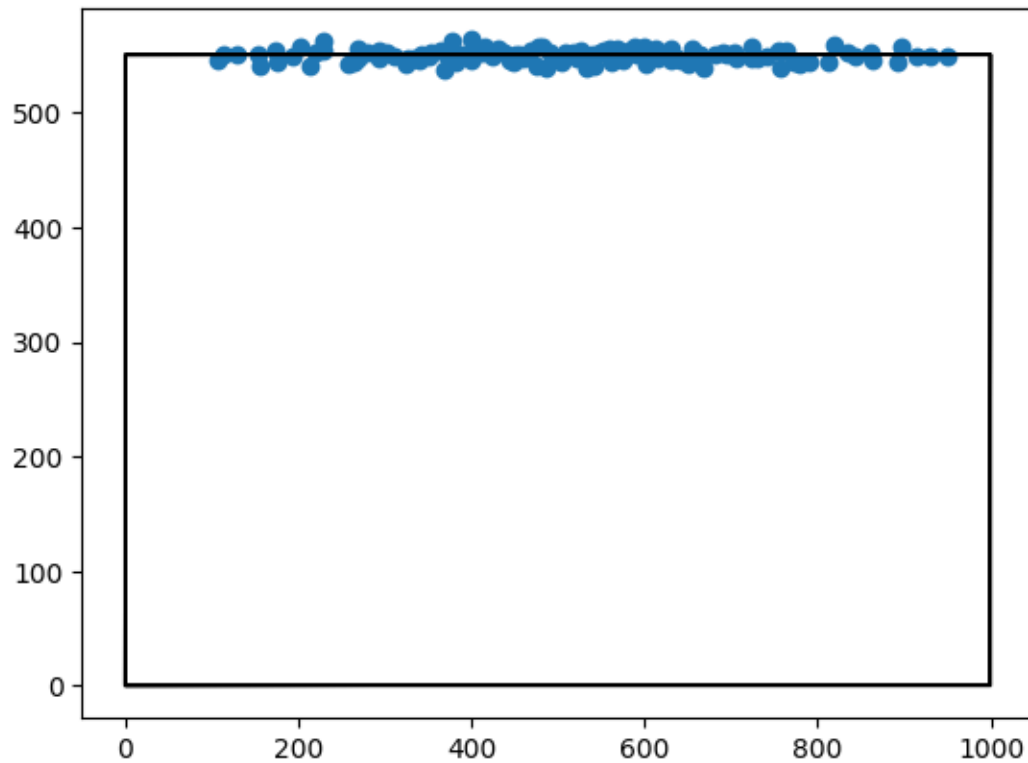
```
mean: 0.00, median: 0.00, standard deviation: 0.00
```

## WITH STATISTICAL ERROR

```
lidar = LidarSimulator("rectangle.stl", error=0.02)
point = [500, 300]
yaw = np.radians(0)
plot_scan = lidar.get_lidar_points(point[0], point[1], yaw, theta=np.radians(90),
    ↪view_range=60)
triangles = lidar.get_map_triangles()
```

```
# get cartesian coordinates
x = []
y = []
for alpha, r in plot_scan:
    x.append(r * np.cos(alpha)+ point[0])
    y.append(r * np.sin(alpha) + point[1])
```

```
fig,ax = plt.subplots()
plt.scatter (x, y)
for t in triangles:
    ax.fill(t[:, 0],t[:, 1],fill=False)
```



```

m = (y[-1] - y[0]) / (x[-1] - x[0])
b = y[-1] - m * x[-1]
y_err = []
for i in range(len(x)):
    y_err.append(y[i] - m * x[i] - b)

y_err = np.array(y_err)
print("mean: %0.2f, median: %0.2f, standard deviation: %0.2f" % (np.mean(y_err), np.
    ↪median(y_err), np.std(y_err)))

```

```

mean: 2.50, median: 2.93, standard deviation: 5.41

```

## 12.1 Statistical Line Fitting

```

model = np.polyfit (x, y, 1)

```

```

y_err = []
for i in range(len(x)):
    y_err.append(y[i] - model[0] * x[i] - model[1])

y_err = np.array(y_err)
print("mean: %0.2f, median: %0.2f, standard deviation: %0.2f" % (np.mean(y_err), np.
    ↪median(y_err), np.std(y_err)))

```

```
mean: 0.00, median: 0.35, standard deviation: 5.32
```

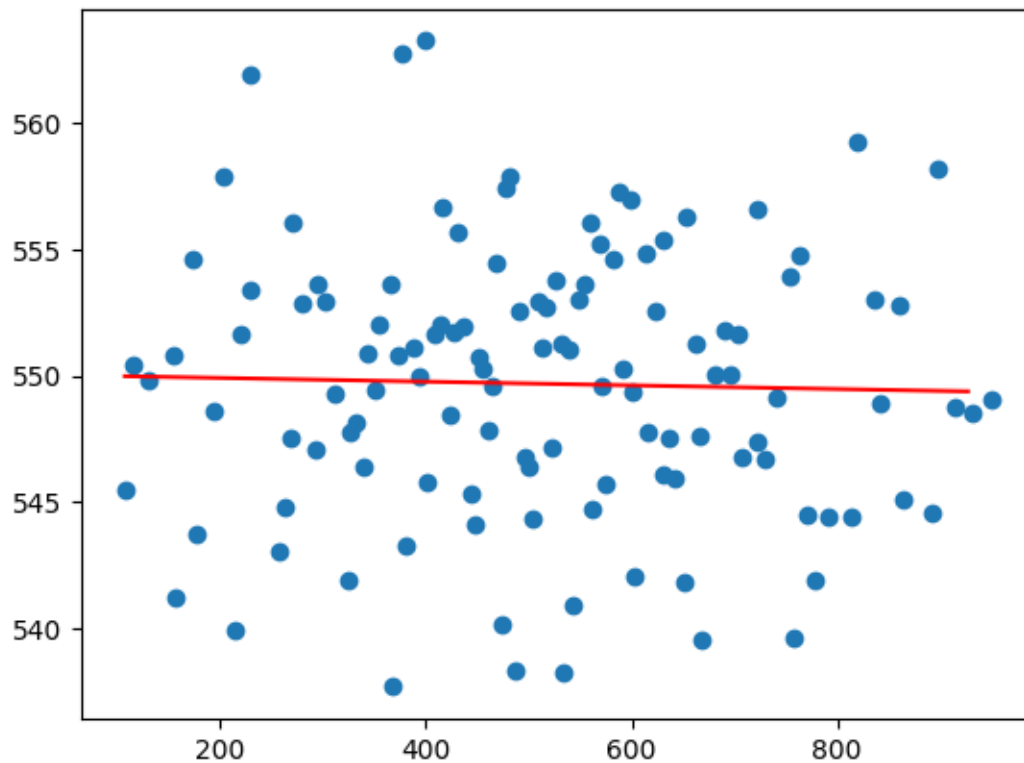
```
x[-1]
```

```
107.13131145805664
```

```
x_lin_reg = np.arange(x[-1], x[1], 10)
predict = np.poly1d(model)
y_lin_reg = predict(x_lin_reg)

fig, ax = plt.subplots()
plt.scatter (x, y)
plt.plot (x_lin_reg, y_lin_reg, c = 'r')
```

```
[<matplotlib.lines.Line2D at 0x110d3ae50>]
```



## 12.2 Fitting a line with Least Square Method

Reference: [https://www.varsitytutors.com/hotmath/hotmath\\_help/topics/line-of-best-fit](https://www.varsitytutors.com/hotmath/hotmath_help/topics/line-of-best-fit)

Step 1: Calculate the mean of the x-values and the mean of the y-values.

$$\bar{X} = \frac{\sum_{i=1}^n x_i}{n}$$

$$\bar{Y} = \frac{\sum_{i=1}^n y_i}{n}$$

Step 2: The following formula gives the slope of the line of best fit:

$$m = \frac{\sum_{i=1}^n (x_i - \bar{X})(y_i - \bar{Y})}{\sum_{i=1}^n (x_i - \bar{X})^2}$$

Step 3: Compute the y-intercept of the line by using the formula:

$$b = \bar{Y} - m\bar{X}$$

Step 4: Use the slope m and the y-intercept b to form the equation of the line.

To construct a perpendicular line we have to solve the equations

$$y = mx + b$$

$$y = -\frac{1}{m}x$$

```
angles = np.arange(0, 85, 5)
values = [0.5197, 0.4404, 0.4850, 0.4222, 0.4132, 0.4371, 0.3912, 0.3949, 0.3919, 0.4276, 0.4075, 0.3956, 0.4053, 0.4752, 0.5032, 0.5273, 0.4879]
```

```
# get cartesian coordinates
x_cart = []
y_cart = []
for i in range(len(values)):
    r = values[i]
    alpha = np.radians(angles[i])
    x_cart.append(r * np.cos(alpha))
    y_cart.append(r * np.sin(alpha))
```

```
# Calculate centroid
X = np.average(x_cart)
Y = np.average(y_cart)
```

```
# calculate line parameters
X_d = (np.array(x_cart) - X)
Y_d = (np.array(y_cart) - Y)
X_d_2 = X_d * X_d
m = (X_d * Y_d).sum() / X_d_2.sum()
b = Y - m * X
```

```
# calculate line
```

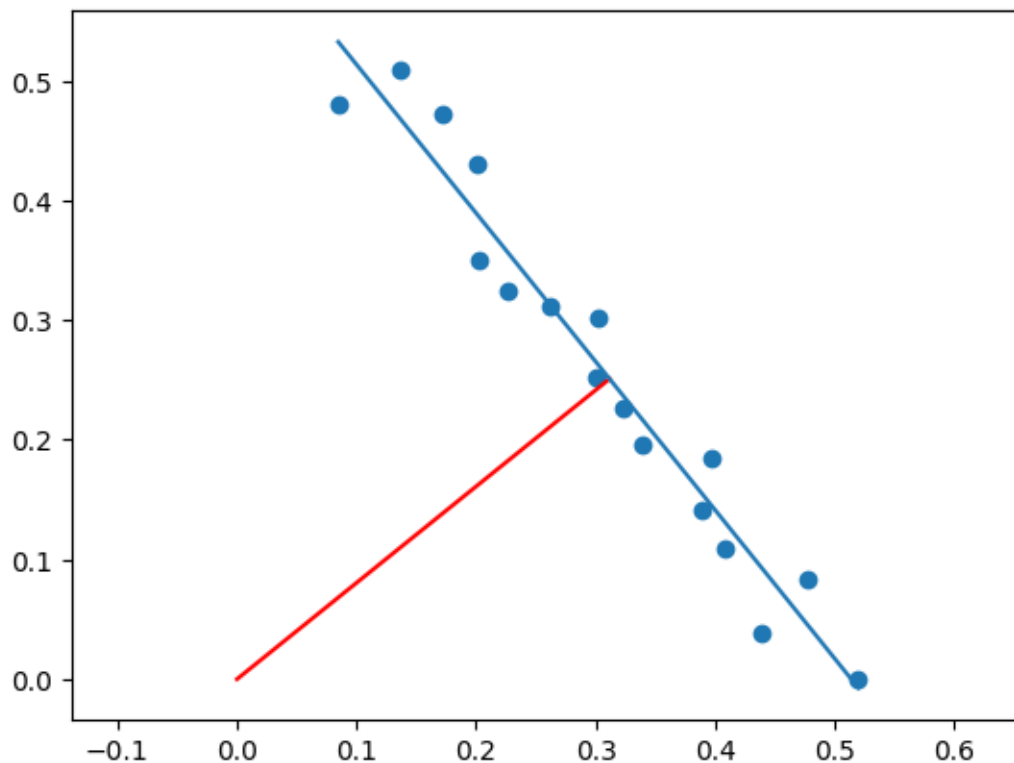
```
line = np.empty((2,2))
line[0,0] = np.array(x_cart).min()
line[1,0] = np.array(x_cart).max()
line[0,1] = m * line[0,0] + b
line[1,1] = m * line[1,0] + b
```

```
# caculate perpendicular line
```

```
x = -b*m/(m*m+1)
y = -x/m
```

```
fig,ax = plt.subplots()
plt.plot([0, x], [0, y], c = 'r')
plt.plot(line[:, 0], line[:, 1])
plt.scatter (x_cart, y_cart)
ax.axis('equal')
```

```
(-0.025985000000000005,
 0.54568500000000001,
 -0.034309997394643715,
 0.5597197002622666)
```



```
alpha = np.arctan2(y, x)
r = np.sqrt(x**2 + y**2)
print("alpha = %.2f r = %.2f" % (np.degrees(alpha), r))
```

```
alpha = 38.85 r = 0.40
```

```
def fitLine(x, y):
    X = np.average(x)
    Y = np.average(y)
    X_d = (np.array(x) - X)
    Y_d = (np.array(y) - Y)
    X_d_2 = X_d * X_d
    X_d_2_sum = X_d_2.sum()
    if np.isclose([X_d_2_sum], [0.0])[0]:
        if X >= 0:
            alpha = 0.0
            r = X
        else:
            alpha = np.pi
            r = -X
        return None, None, alpha, r
    m = (X_d * Y_d).sum() / X_d_2.sum()
    b = Y - m * X
    # calculate perpendicular line
    x_p = -b*m/(m*m+1)
    if np.isclose([m], [0.0])[0]:
        if Y >= 0:
            alpha = np.pi / 2
            r = Y
        else:
            alpha = 3 * np.pi / 2
            r = -Y
        return None, None, alpha, r
    y_p = -x_p/m
    alpha = np.arctan2(y_p, x_p)
    r = np.sqrt(x_p**2 + y_p**2)
    return m, b, alpha, r
```

```
x_cart = -1 * np.array(x_cart)
y_cart = -1 * np.array(y_cart)
```

```
m, b, alpha, r = fitLine(x_cart, y_cart)
print("alpha = %.2f r = %.2f" % (np.degrees(alpha), r))
```

```
alpha = -141.15 r = 0.40
```

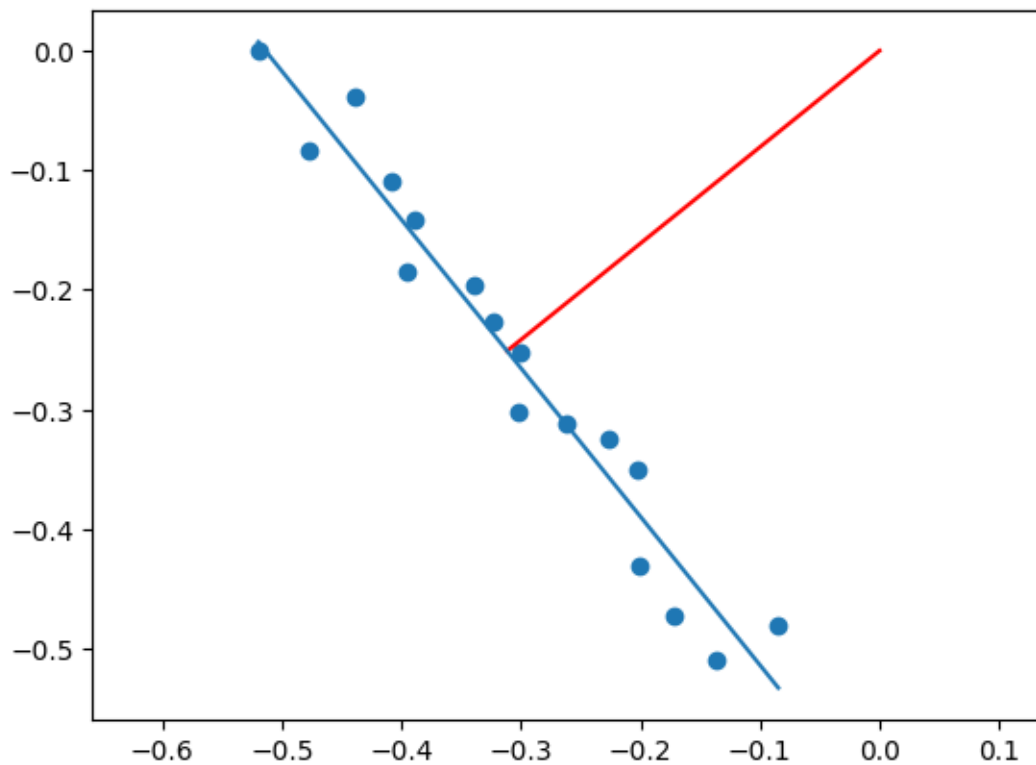
```
# calculate line
line = np.empty((2,2))
line[0,0] = np.array(x_cart).min()
line[1,0] = np.array(x_cart).max()
line[0,1] = m * line[0,0] + b
line[1,1] = m * line[1,0] + b
```

```
# calculate perpendicular line
x = -b*m/(m*m+1)
y = -x/m
```



```
fig, ax = plt.subplots()
plt.plot([0, x], [0, y], c = 'r')
plt.plot(line[:, 0], line[:, 1])
plt.scatter (x_cart, y_cart)
ax.axis('equal')
```

```
(-0.545685000000000001,
 0.0259850000000000005,
 -0.5597197002622666,
 0.034309997394643715)
```



```
cosA = np.cos(alpha);
sinA = np.sin(alpha);

xcosA = x_cart * cosA;
ysinA = y_cart * sinA;
```

```
m, b, alpha, r = fitLine([0, 1], [1, 0])
np.testing.assert_allclose([alpha, r], [np.pi / 4, np.sqrt(0.5)])
```

```
m, b, alpha, r = fitLine([-1, 0], [0, 1])
np.testing.assert_allclose([alpha, r], [3 * np.pi / 4, np.sqrt(0.5)])
```

```
m, b, alpha, r = fitLine([-1, 1], [1, 1])
np.testing.assert_allclose([alpha, r], [np.pi/2, 1.0])
```

```
m, b, alpha, r = fitLine([-1, 1], [-1, -1])
np.testing.assert_allclose([alpha, r], [3 * np.pi / 2, 1.0])
```

```
m, b, alpha, r = fitLine([1, 1], [1, -1])
np.testing.assert_allclose([alpha, r], [0.0, 1.0])
```

```
m, b, alpha, r = fitLine([-1, -1], [1, -1])
np.testing.assert_allclose([alpha, r], [np.pi, 1.0])
```

```
rng = np.random.default_rng(seed=43)
for i in range(30):
    alpha_expected = rng.uniform(low=-np.pi, high=np.pi)
    r_expected = rng.uniform(low=1.0, high=2.0)
    ts = np.arange(-1.0, 1.1, 0.1)
    p = [r_expected * np.cos(alpha_expected), r_expected * np.sin(alpha_expected)]
    v = [-np.sin(alpha_expected), np.cos(alpha_expected)]
    x_cart = p[0] + v[0] * ts
    y_cart = p[1] + v[1] * ts
    m, b, alpha, r = fitLine(x_cart, y_cart)
    np.testing.assert_allclose([alpha, r], [alpha_expected, r_expected])
    error = rng.uniform(low=-0.01, high=0.01, size=(2, len(ts)))
    x_cart += error[0]
    y_cart += error[1]
    m, b, alpha, r = fitLine(x_cart, y_cart)
    np.testing.assert_allclose([alpha, r], [alpha_expected, r_expected], atol=0.01)
```

## 12.3 Fitting a line with Polar Coordinates

$\sigma_i^2$  is the variance that models the uncertainty regarding distance  $\rho_i$  of a particular sensor measurement

$$w_i = \frac{1}{\sigma_i^2}$$

$$\alpha = \frac{1}{2} \operatorname{atan} \left( \frac{\sum w_i \rho_i^2 \sin 2\theta_i - \frac{2}{\sum w_i} \sum \sum w_i w_j \rho_i \rho_j \cos \theta_i \sin \theta_j}{\sum w_i \rho_i^2 \cos 2\theta_i - \frac{1}{\sum w_i} \sum \sum w_i w_j \rho_i \rho_j \cos(\theta_i + \theta_j)} \right)$$

$$r = \frac{\sum w_i \theta_i \cos(\theta_i - \alpha)}{\sum w_i}$$

## 12.4 TODO

```
c = np.cos(angles) c2 = np.cos(2 * angles) s = np.sin(angles) s2 = np.sin(2 * angles) r_square = np.array(values)**2 N =
angles.shape[0] y = r_square * s2 - 2/N * np.array(values) * c * np.array(values) * s x = r_square * c2 - csIJ / N; alpha =
0.5 * (atan2(y, x) + pi); r = rho * cos(theta - ones(size(theta)) * alpha) / N;
```

## ALGORITHM 1: SPLIT-AND-MERGE

**Data:** Set  $S$  consisting of all  $N$  points, a distance threshold  $d > 0$

**Result:**  $L$ , a list of sets of points each resembling a line

```
L ← (S), i ← 1;
while i ≤ len(L) do
    fit a line (r, α) to the set Li;
    detect the point P ∈ Li with the maximum distance D to the line (r, α);
    if D < d then
        i ← i + 1
    else
        split Li at P into S1 and S2;
        Li ← S1; Li+1 ← S2;
    end
end
Merge collinear sets in L;
```

### 13.1 Reference

Roland Siegwart, Illah Nourbakhsh, and Davide Scaramuzza. Introduction to Autonomous Mobile Robots. MIT Press, 2nd edition, 2011.

```
%matplotlib inline
import matplotlib.pyplot as plt
import numpy as np
import pickle
from SplitAndMerge.split_and_merge import SplitAndMerge
```

```
testdata = []
for i in range(6):
    with open('/Users/hdumcke/git/lidar-simulator/jupyternb/data/testLineExtraction%s.
    mat.pickle' % str(i+1), 'rb') as f:
        testdata.append(pickle.load(f))
```

```
fig, axs = plt.subplots(3, 2)
for i in range(6):
    row_index = int(i / 2)
    column_index = i % 2
```

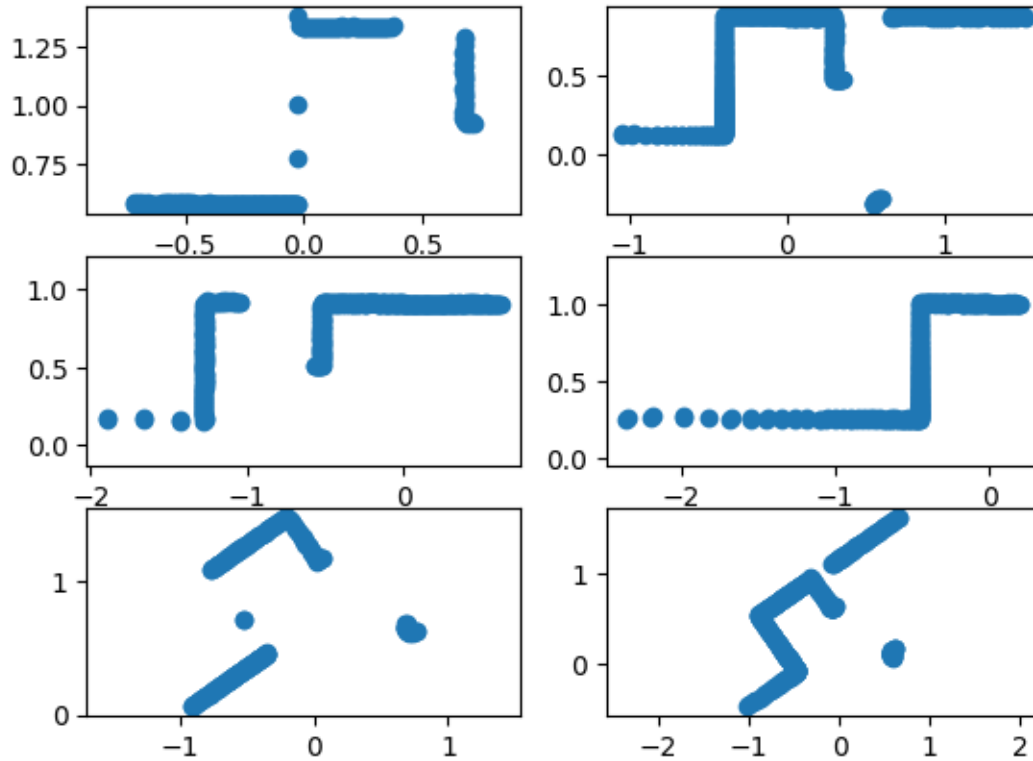
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```

axs[row_index, column_index].axis('equal')
x = testdata[i]['rho'] * np.cos(testdata[i]['theta'])
y = testdata[i]['rho'] * np.sin(testdata[i]['theta'])
axs[row_index, column_index].scatter(x, y)

```

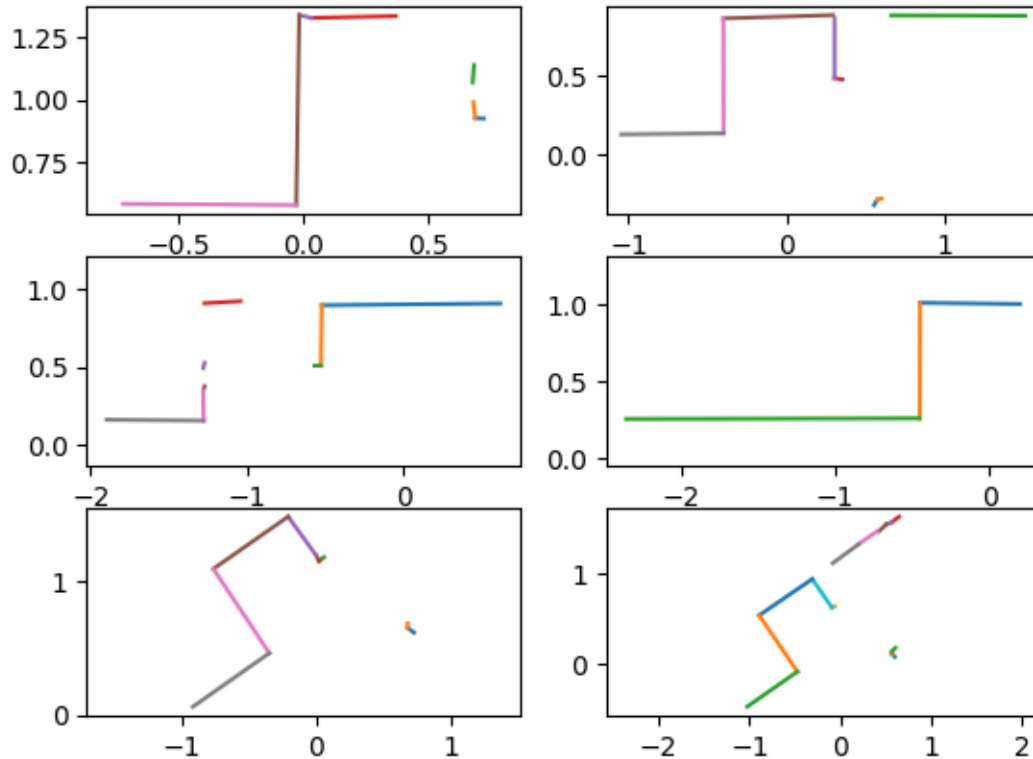


```

sam = SplitAndMerge(line_point_dist_threshold=0.004, min_points_per_segment=4, min_
    ↪seg_length=0.01)

fig, axs = plt.subplots(3, 2)
for i in range(6):
    row_index = int(i / 2)
    column_index = i % 2
    axs[row_index, column_index].axis('equal')
    x = testdata[i]['rho'] * np.cos(testdata[i]['theta'])
    y = testdata[i]['rho'] * np.sin(testdata[i]['theta'])
    alpha_a, r_a, segend, seglen, pointIdx_a = sam.extractLines(x[0], y[0])
    for j in range(segend.shape[0]):
        axs[row_index, column_index].plot([segend[j,0], segend[j,2]], [segend[j,1],
    ↪segend[j,3]])

```



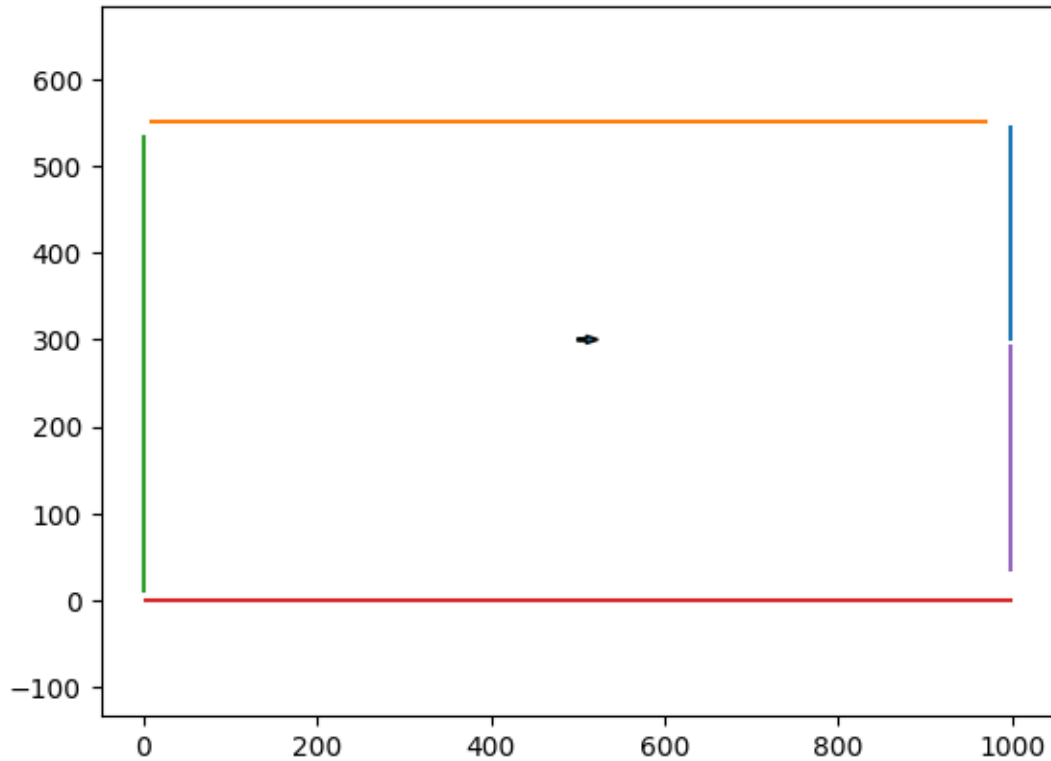
```
from LidarSim.lidar_sim import LidarSimulator
lidar = LidarSimulator("rectangle.stl")
```

```
point = [500, 300]
yaw = np.radians(0)
#plot_scan = lidar.get_lidar_points(point[0], point[1], yaw, theta=0, view_range=25)
plot_scan = lidar.get_lidar_points(point[0], point[1], yaw)
```

```
# get cartesian coordinates
x = []
y = []
for alpha, r in plot_scan:
    x.append(r * np.cos(alpha) + point[0])
    y.append(r * np.sin(alpha) + point[1])
```

```
sam = SplitAndMerge(line_point_dist_threshold=0.005, min_points_per_segment=5)

fig, ax = plt.subplots()
ax.axis('equal')
plt.arrow(point[0], point[1], 10 * np.cos(yaw), 10 * np.sin(yaw), width=3.0)
#plt.scatter (x[211:329], y[211:329])
alpha_a, r_a, segend, seglen, pointIdx_a = sam.extractLines(x, y)
for j in range(segend.shape[0]):
    ax.plot([segend[j,0], segend[j,2]], [segend[j,1], segend[j,3]])
```



```
import Utilities.utilities as utilities
```

```
from LidarSim.lidar_sim import LidarSimulator
lidar = LidarSimulator("maze.stl")
triangles = lidar.get_map_triangles()
point = [25, 25]
yaw = np.radians(90)
```

```
pose = [[25, 25, 90 ],
        [25, 175, 90 ],
        [25, 175, 0 ],
        [75, 175, 0 ],
        [75, 125, 0 ],
        [25, 325, 90 ]]

fig, axs = plt.subplots(3, 2, figsize=(15, 15))
for i in range(6):
    row_index = int(i / 2)
    column_index = i % 2
    axs[row_index, column_index].axis('equal')
    point = pose[i][0:2]
    yaw = np.radians(pose[i][2])
    plot_scan = lidar.get_lidar_points(point[0], point[1], yaw)
    # get cartesian coordinates
    x = []
    y = []
    for alpha, r in plot_scan:
        x.append(r * np.cos(alpha) + point[0])
```

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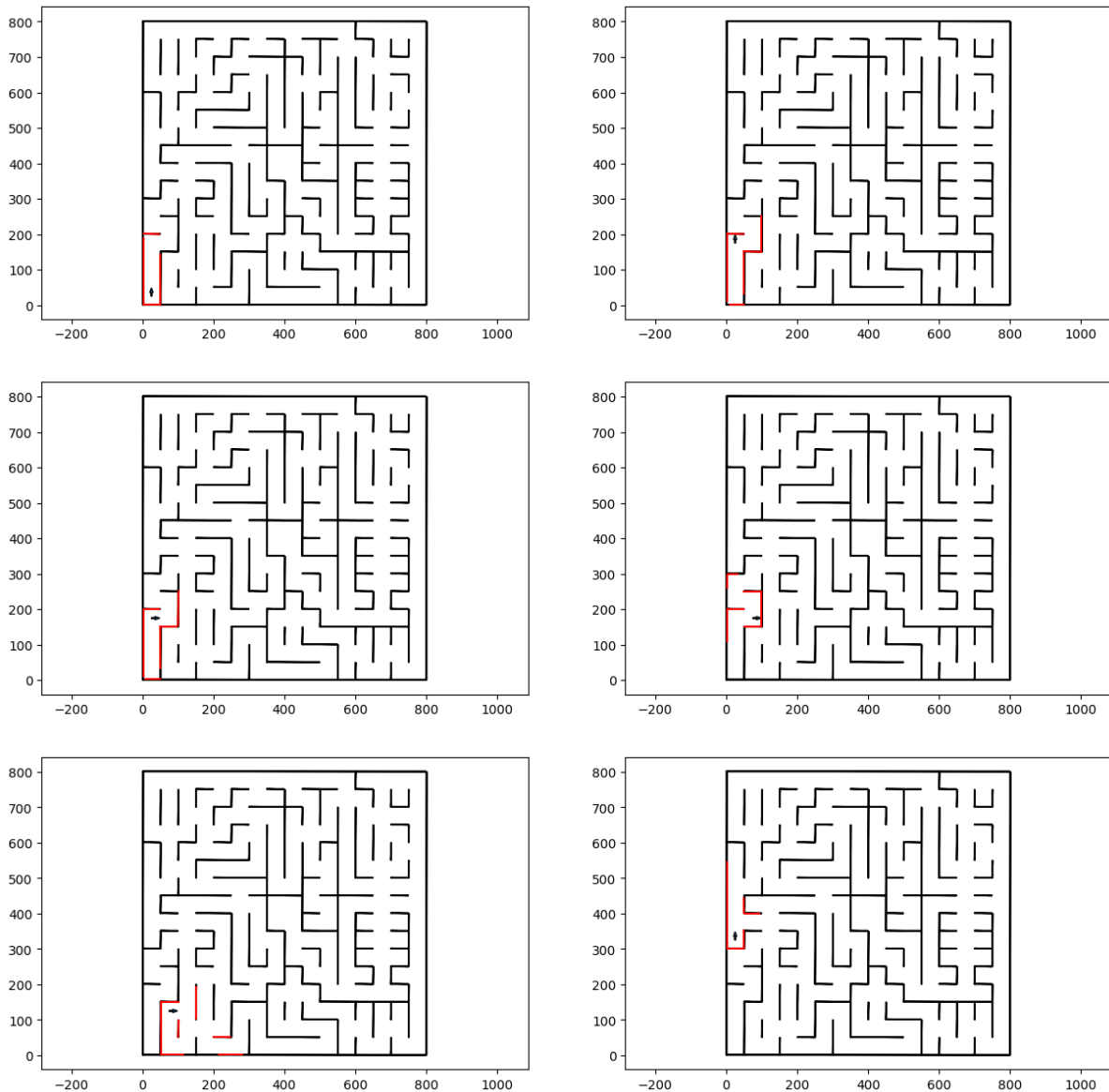
```

y.append(r * np.sin(alpha) + point[1])

sam = SplitAndMerge(line_point_dist_threshold=0.005, min_points_per_segment=5)

axs[row_index, column_index].arrow(point[0], point[1], 10 * np.cos(yaw), 10 * np.
sin(yaw), width=3.0)
alpha_a, r_a, segend, seglen, pointIdx_a = sam.extractLines(x, y)
for j in range(segend.shape[0]):
    x_p, y_p = utilities.rotate_segend(segend[j], point, yaw)
    axs[row_index, column_index].plot(x_p, y_p, 'r')
for t in triangles:
    axs[row_index, column_index].fill(t[:, 0], t[:, 1], fill=False)

```



```

from LidarSim.lidar_sim import LidarSimulator
lidar = LidarSimulator("racetrack.stl")

```

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```
triangles = lidar.get_map_triangles()
yaw = np.radians(180)
```

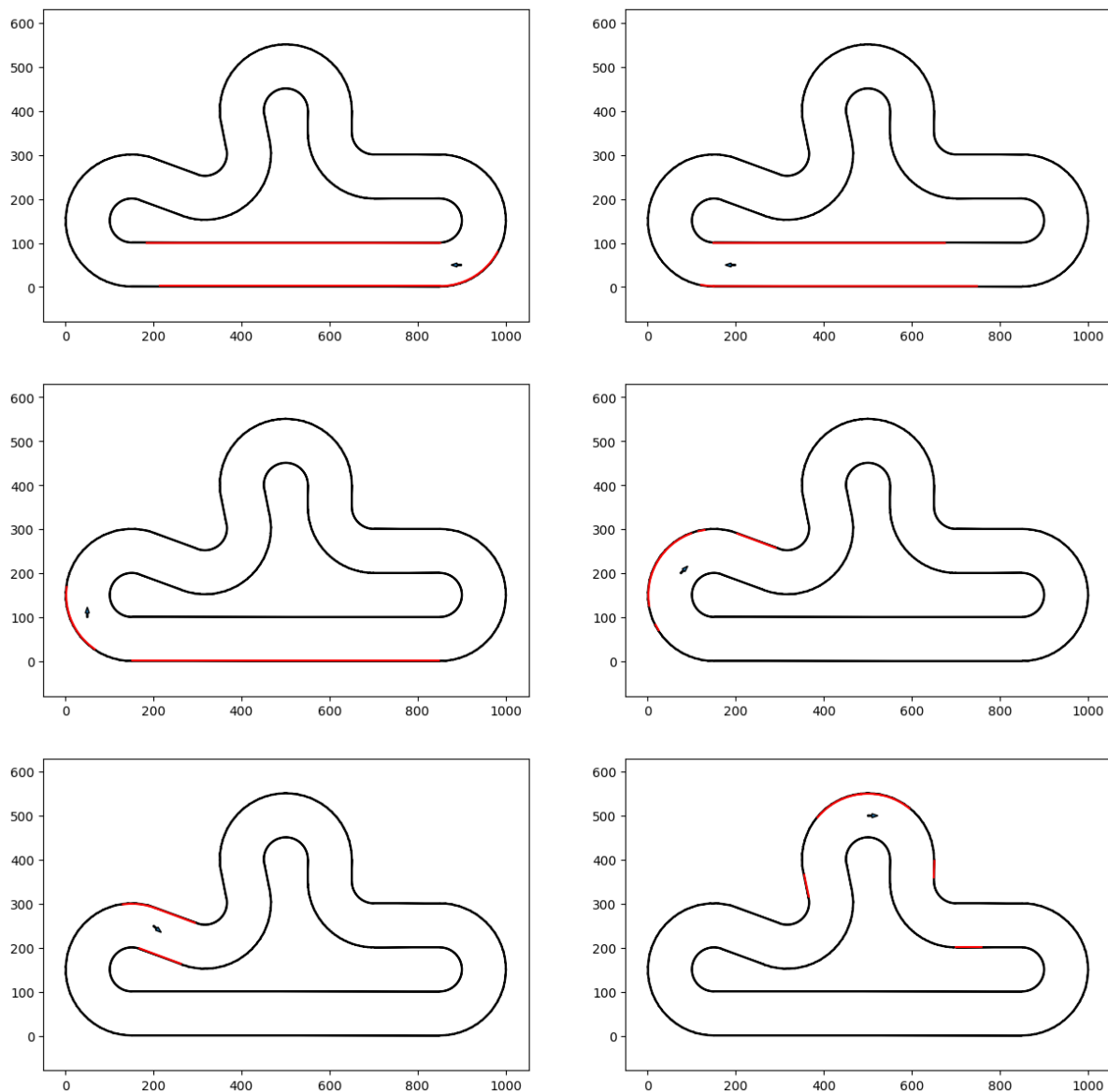
```
pose = [[900, 50, 180 ],
        [200, 50, 180 ],
        [50, 100, 90 ],
        [75, 200, 45 ],
        [200, 250, -40 ],
        [500, 500, 0 ]]

fig, axs = plt.subplots(3, 2, figsize=(15, 15))
for i in range(6):
    row_index = int(i / 2)
    column_index = i % 2
    axs[row_index, column_index].axis('equal')
    point = pose[i][0:2]
    yaw = np.radians(pose[i][2])
    plot_scan = lidar.get_lidar_points(point[0], point[1], yaw)
    # get cartesian coordinates
    x = []
    y = []
    for alpha, r in plot_scan:
        x.append(r * np.cos(alpha)+ point[0])
        y.append(r * np.sin(alpha) + point[1])

    sam = SplitAndMerge(line_point_dist_threshold=0.005, min_points_per_segment=5)

    axs[row_index, column_index].arrow(point[0], point[1], 10 * np.cos(yaw), 10 * np.
    sin(yaw), width=3.0)
    alpha_a, r_a, segend, seglen, pointIdx_a = sam.extractLines(x, y)
    for j in range(segend.shape[0]):
        x_p, y_p = utilities.rotate_segend(segend[j], point, yaw)
        axs[row_index, column_index].plot(x_p, y_p, 'r')
    for t in triangles:
        axs[row_index, column_index].fill(t[:, 0],t[:, 1],fill=False)
```







## ALGORITHM 2: LINE-REGRESSION

1. Initialize sliding window size  $N_f$
2. Fit a line to every  $N_f$  consecutive points
3. Compute a line fidelity array. Each element of the array contains the sum of   
→ Mahalanobis distances between every three adjacent windows
4. Construct line segments by scanning the fidelity array **for** consecutive elements   
→ having values less than a threshold
5. Merge overlapped line segments **and** recompute line parameters **for** each segment

### 14.1 Reference

Roland Siegwart, Illah Nourbakhsh, and Davide Scaramuzza. Introduction to Autonomous Mobile Robots. MIT Press, 2nd edition, 2011.

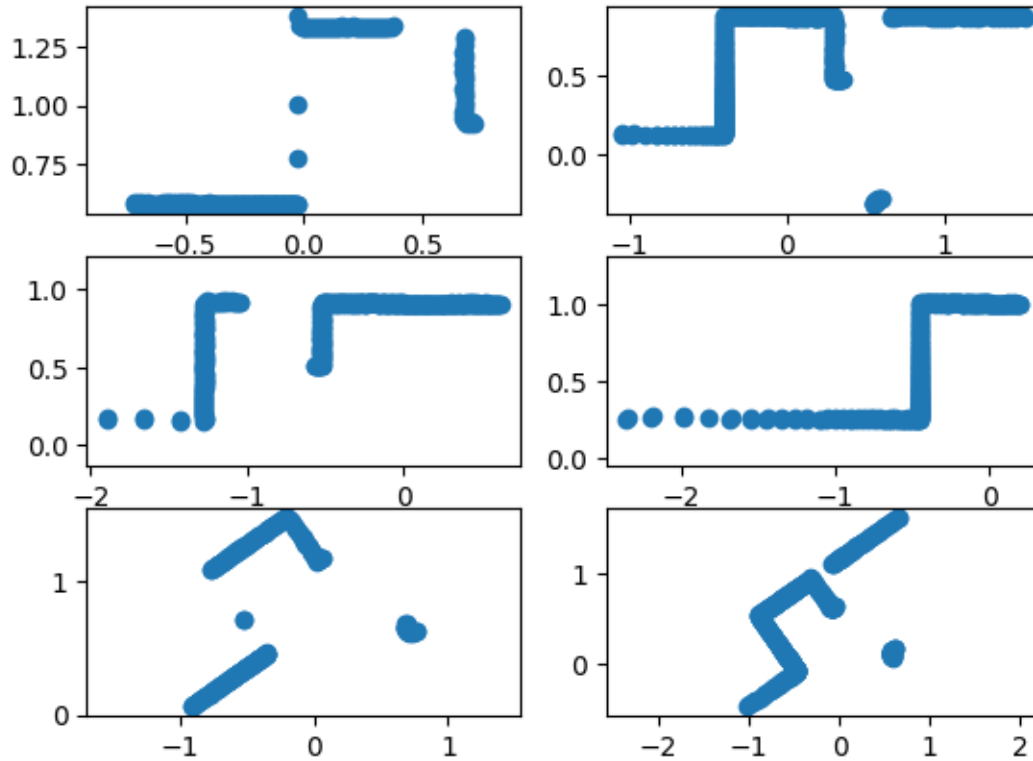
```
## ToDo
```

```
## Work in progress
```

```
%matplotlib inline
import matplotlib.pyplot as plt
import numpy as np
import pickle
from LineRegression.lineregression import LineRegression
```

```
testdata = []
for i in range(6):
    with open('/Users/hdumcke/git/lidar-simulator/jupyternb/data/testLineExtraction%s.
→mat.pickle' % str(i+1), 'rb') as f:
        testdata.append(pickle.load(f))
```

```
fig, axs = plt.subplots(3, 2)
for i in range(6):
    row_index = int(i / 2)
    column_index = i % 2
    axs[row_index, column_index].axis('equal')
    x = testdata[i]['rho'] * np.cos(testdata[i]['theta'])
    y = testdata[i]['rho'] * np.sin(testdata[i]['theta'])
    axs[row_index, column_index].scatter(x, y)
```

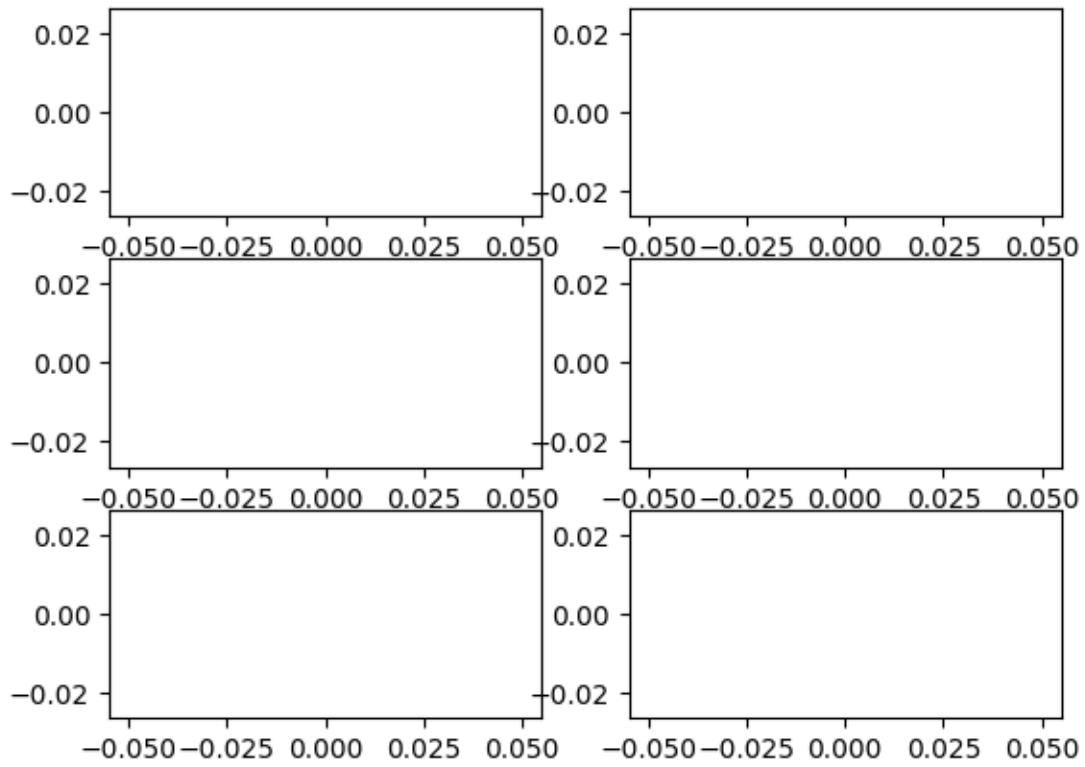


```

lr = LineRegression()

fig, axs = plt.subplots(3, 2)
for i in range(6):
    row_index = int(i / 2)
    column_index = i % 2
    axs[row_index, column_index].axis('equal')
    x = testdata[i]['rho'] * np.cos(testdata[i]['theta'])
    y = testdata[i]['rho'] * np.sin(testdata[i]['theta'])
    segend, seglen, pointIdx_a = lr.extractLines(x[0], y[0])
    for j in range(segend.shape[0]):
        axs[row_index, column_index].plot([segend[j,0], segend[j,2]], [segend[j,1],
↪segend[j,3]])

```





## ALGORITHM 3: INCREMENTAL

```
1. Start by the first 2 points, construct a line
2. Add the next point to the current line model
3. Recompute the line parameters by line fitting
4. If it satisfies the line condition, continue (go to step 2)
5. Otherwise, put back the last point, recompute the line parameters, return the line
6. Continue with the next two points, go to step 2
```

### 15.1 Reference

Roland Siegwart, Illah Nourbakhsh, and Davide Scaramuzza. Introduction to Autonomous Mobile Robots. MIT Press, 2nd edition, 2011.

```
Put all points on curve list, in order along the curve
Empty the line point list
Empty the line list
Until there are too few points on the curve
    Transfer first few points on the curve to the line point list
    Fit line to line point list
    While fitted line is good enough
        Transfer the next point on the curve to the line point list and refit the line
    end
    Transfer last point(s) back to curve
    Refit line
    Attach line to line list
end
```

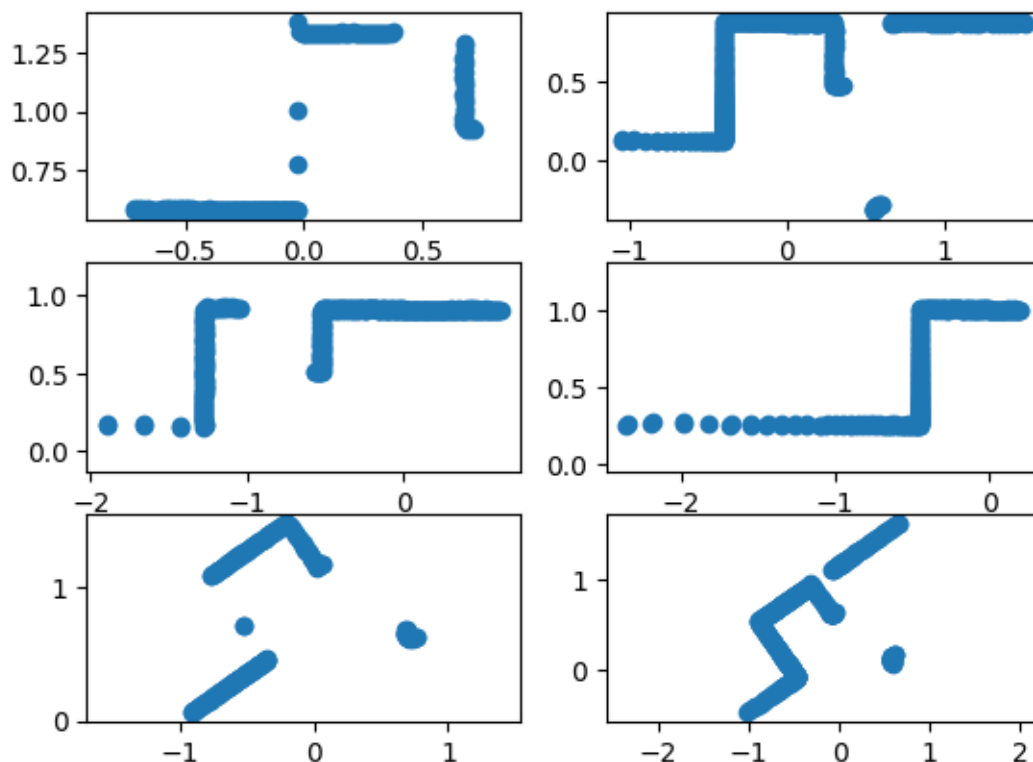
### 15.2 Reference

Forsyth, D. A., Ponce, J., Computer Vision: A Modern Approach. Upper Saddle River, NJ, Prentice Hall, 2003.

```
%matplotlib inline
import matplotlib.pyplot as plt
import numpy as np
import pickle
from Incremental.incremental import Incremental
```

```
testdata = []
for i in range(6):
    with open('/Users/hdumcke/git/lidar-simulator/jupyternb/data/testLineExtraction%s.
    ↪mat.pickle' % str(i+1), 'rb') as f:
        testdata.append(pickle.load(f))
```

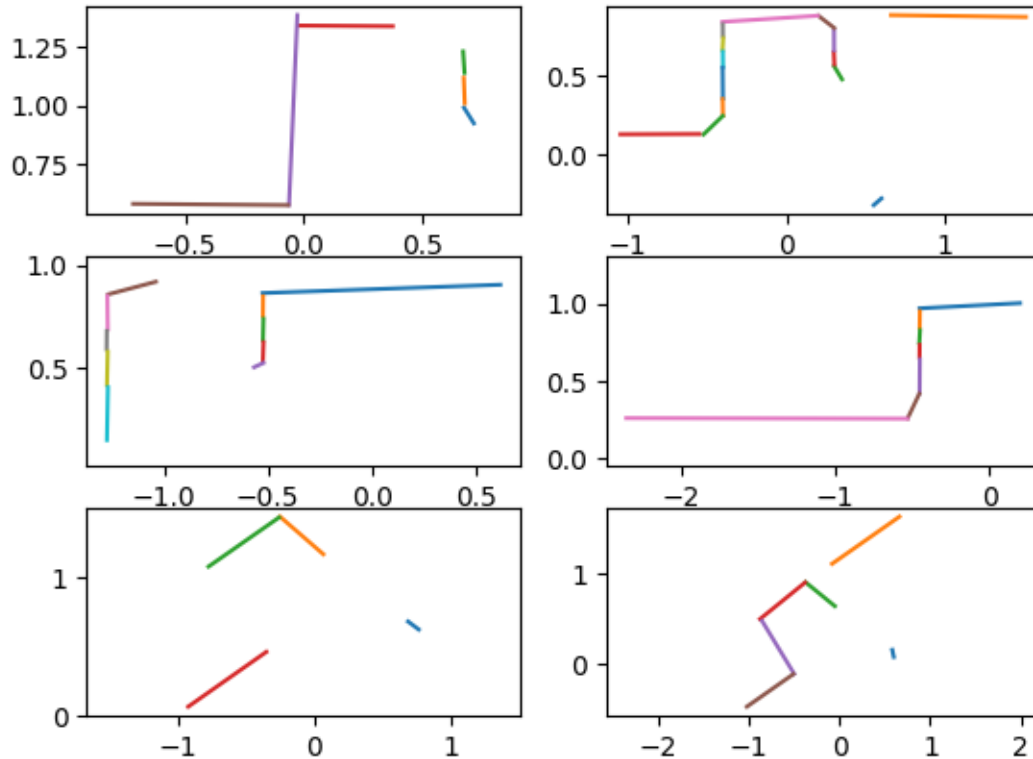
```
fig, axs = plt.subplots(3, 2)
for i in range(6):
    row_index = int(i / 2)
    column_index = i % 2
    axs[row_index, column_index].axis('equal')
    x = testdata[i]['rho'] * np.cos(testdata[i]['theta'])
    y = testdata[i]['rho'] * np.sin(testdata[i]['theta'])
    axs[row_index, column_index].scatter(x, y)
```



```
inc = Incremental(dist_threshold=0.05, min_points_per_segment=5)

fig, axs = plt.subplots(3, 2)
for i in range(6):
    row_index = int(i / 2)
    column_index = i % 2
    axs[row_index, column_index].axis('equal')
    x = testdata[i]['rho'] * np.cos(testdata[i]['theta'])
    y = testdata[i]['rho'] * np.sin(testdata[i]['theta'])
    segend, seglen, pointIdx_a = inc.extractLines(x[0], y[0])
    for j in range(segend.shape[0]):
        axs[row_index, column_index].plot([segend[j,0], segend[j,2]], [segend[j,1],
    ↪segend[j,3]])
```





```
from LidarSim.lidar_sim import LidarSimulator
lidar = LidarSimulator("rectangle.stl")
```

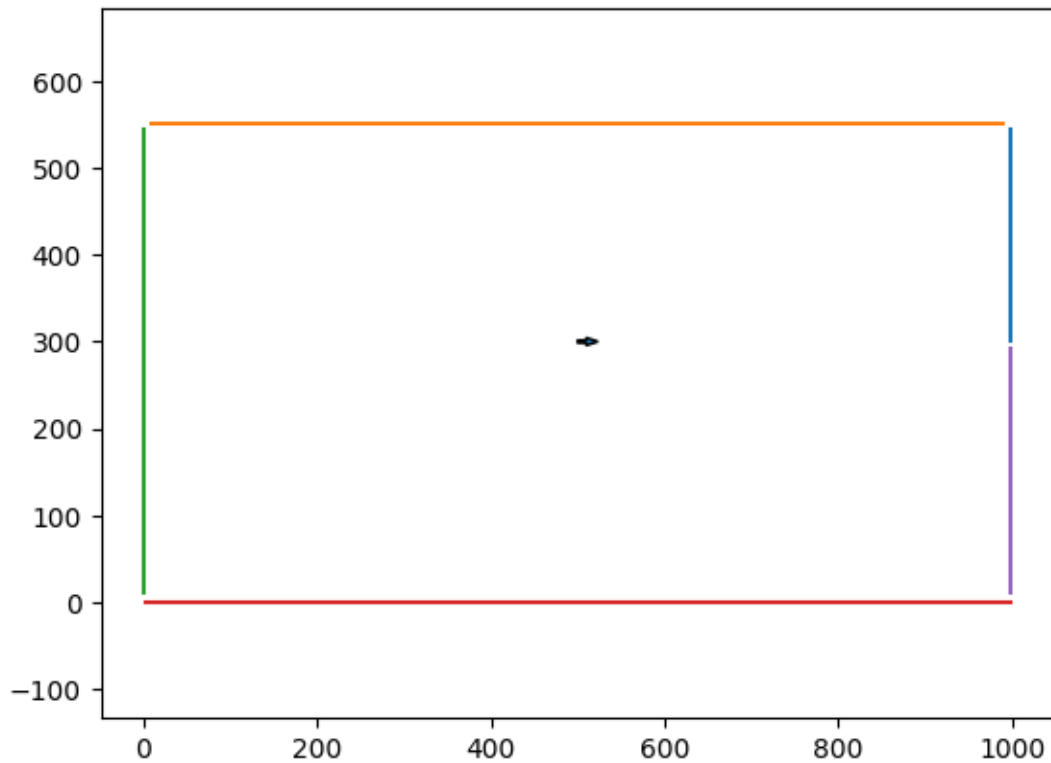
```
point = [500, 300]
yaw = np.radians(0)
plot_scan = lidar.get_lidar_points(point[0], point[1], yaw)
```

```
# get cartesian coordinates
x = []
y = []
for alpha, r in plot_scan:
    x.append(r * np.cos(alpha) + point[0])
    y.append(r * np.sin(alpha) + point[1])
```

```
inc = Incremental()

fig, ax = plt.subplots()
ax.axis('equal')
plt.arrow(point[0], point[1], 10 * np.cos(yaw), 10 * np.sin(yaw), width=3.0)
# plt.scatter(x[212:330], y[212:330])
segend, seglen, pointIdx_a = inc.extractLines(x, y)
for j in range(segend.shape[0]):
    ax.plot([segend[j,0], segend[j,2]], [segend[j,1], segend[j,3]])
```

```
/Users/hdumcke/git/lidar-simulator/python/incremental/src/Incremental/incremental.
py:20: RuntimeWarning: invalid value encountered in scalar divide
m = (X_d * Y_d).sum() / X_d_2.sum()
```



```
import Utilities.utilities as utilities
```

```
from LidarSim.lidar_sim import LidarSimulator
lidar = LidarSimulator("maze.stl")
triangles = lidar.get_map_triangles()
point = [25, 25]
yaw = np.radians(90)
```

```
pose = [[25, 25, 90 ],
        [25, 175, 90 ],
        [25, 175, 0 ],
        [75, 175, 0 ],
        [75, 125, 0 ],
        [25, 325, 90 ]]

fig, axs = plt.subplots(3, 2, figsize=(15, 15))
for i in range(6):
    row_index = int(i / 2)
    column_index = i % 2
    axs[row_index, column_index].axis('equal')
    point = pose[i][0:2]
    yaw = np.radians(pose[i][2])
    plot_scan = lidar.get_lidar_points(point[0], point[1], yaw)
    # get cartesian coordinates
    x = []
    y = []
    for alpha, r in plot_scan:
        x.append(r * np.cos(alpha) + point[0])
```

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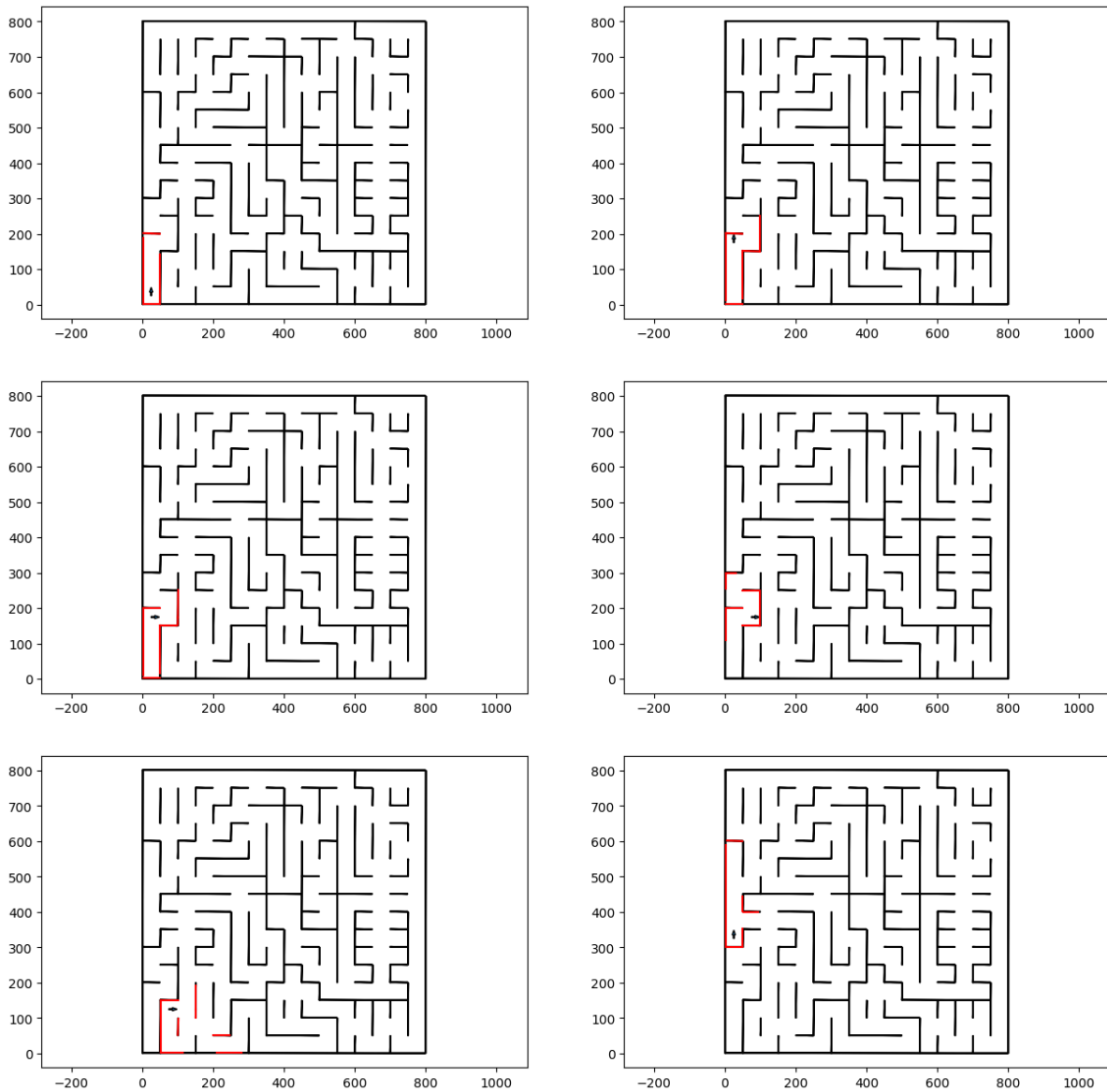
```

y.append(r * np.sin(alpha) + point[1])

inc = Incremental()

axs[row_index, column_index].arrow(point[0], point[1], 10 * np.cos(yaw), 10 * np.
sin(yaw), width=3.0)
segend, seglen, pointIdx_a = inc.extractLines(x, y)
for j in range(segend.shape[0]):
    x_p, y_p = utilities.rotate_segend(segend[j], point, yaw)
    axs[row_index, column_index].plot(x_p, y_p, 'r')
for t in triangles:
    axs[row_index, column_index].fill(t[:, 0], t[:, 1], fill=False)

```



```

from LidarSim.lidar_sim import LidarSimulator
lidar = LidarSimulator("racetrack.stl")

```

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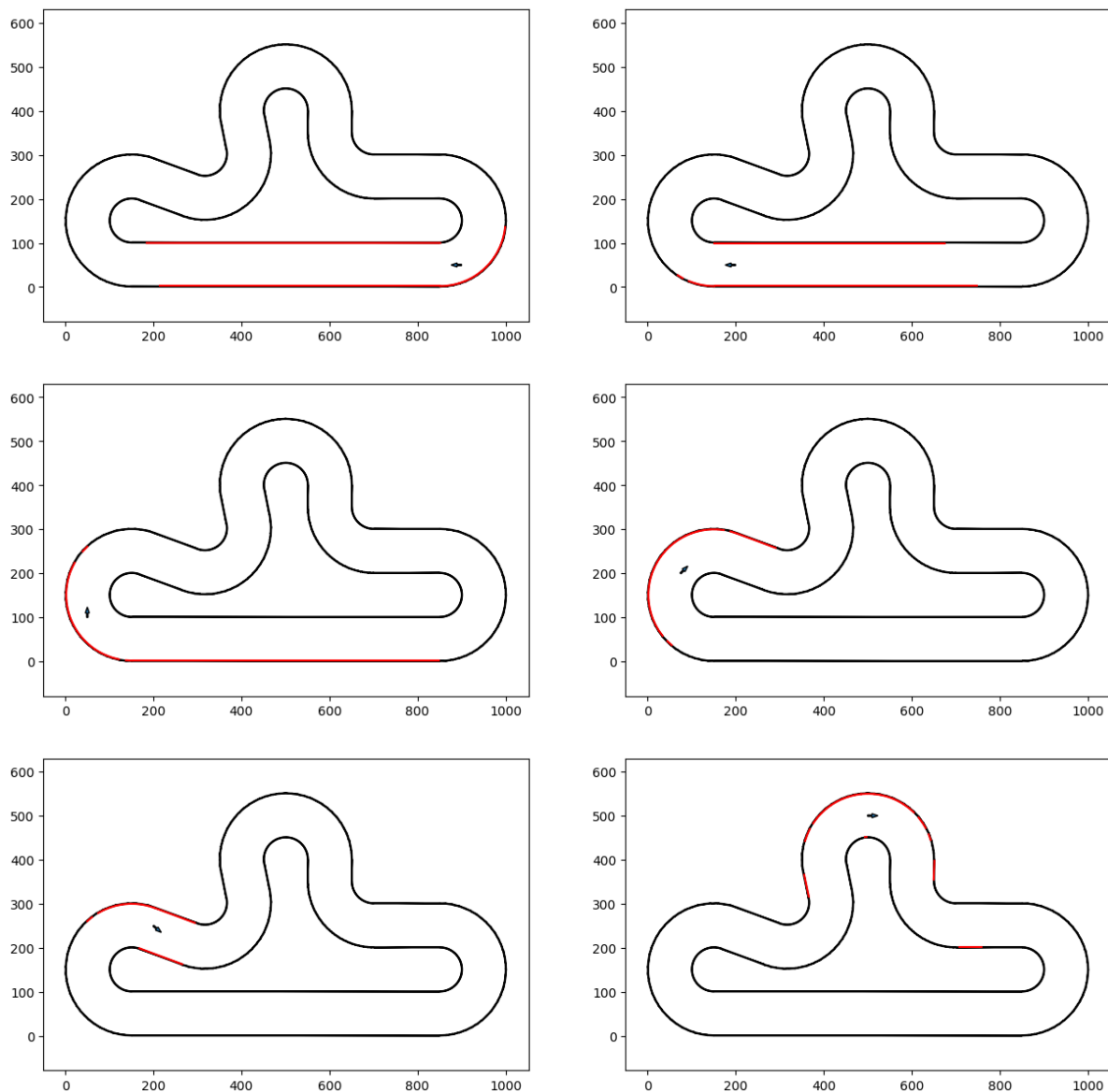
```
triangles = lidar.get_map_triangles()
yaw = np.radians(180)
```

```
pose = [[900, 50, 180 ],
        [200, 50, 180 ],
        [50, 100, 90 ],
        [75, 200, 45 ],
        [200, 250, -40 ],
        [500, 500, 0 ]]

fig, axs = plt.subplots(3, 2, figsize=(15, 15))
for i in range(6):
    row_index = int(i / 2)
    column_index = i % 2
    axs[row_index, column_index].axis('equal')
    point = pose[i][0:2]
    yaw = np.radians(pose[i][2])
    plot_scan = lidar.get_lidar_points(point[0], point[1], yaw)
    # get cartesian coordinates
    x = []
    y = []
    for alpha, r in plot_scan:
        x.append(r * np.cos(alpha)+ point[0])
        y.append(r * np.sin(alpha) + point[1])

    inc = Incremental()

    axs[row_index, column_index].arrow(point[0], point[1], 10 * np.cos(yaw), 10 * np.
    sin(yaw), width=3.0)
    segend, seglen, pointIdx_a = inc.extractLines(x, y)
    for j in range(segend.shape[0]):
        x_p, y_p = utilities.rotate_segend(segend[j], point, yaw)
        axs[row_index, column_index].plot(x_p, y_p, 'r')
    for t in triangles:
        axs[row_index, column_index].fill(t[:, 0],t[:, 1],fill=False)
```





## ALGORITHM 4: RANSAC

```
1. Initial: let A be a set of N points
2. repeat
3. Randomly select a sample of 2 points from A
4. Fit a line through the 2 points
5. Compute the distances of all other points to this line
6. Construct the inlier set (i.e. count the number of points with distance to the
   line < d)
7. Store these inliers
8. until Maximum number of iterations k reached
9. The set with the maximum number of inliers is chosen as a solution to the problem
```

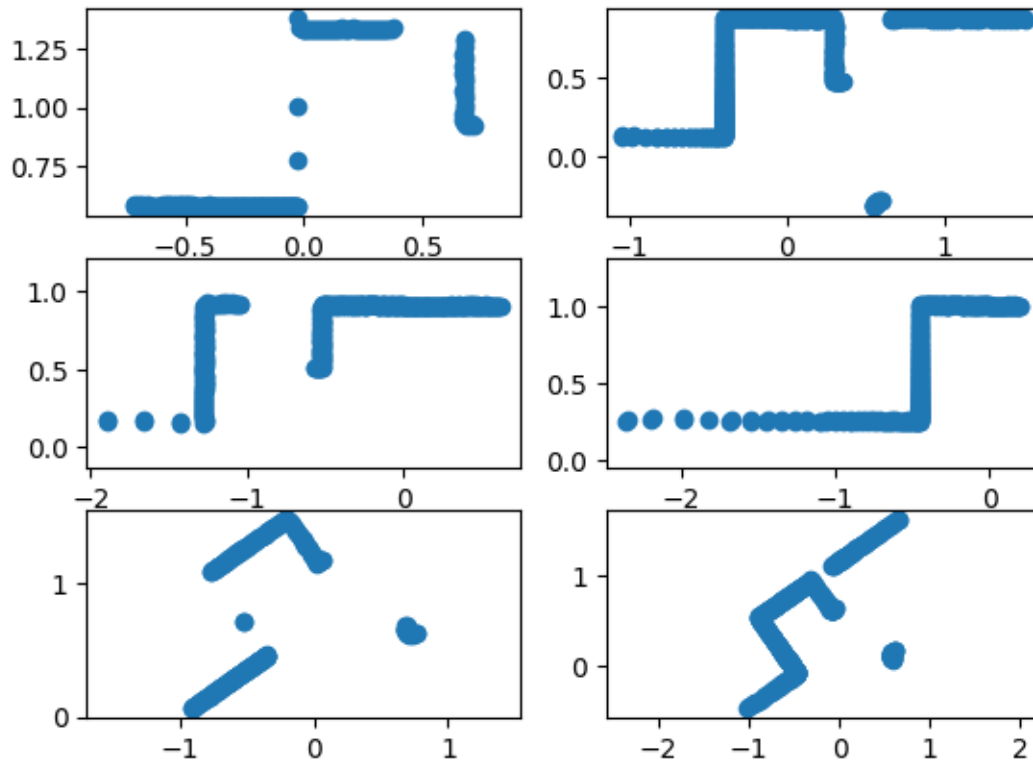
### 16.1 Reference

Roland Siegwart, Illah Nourbakhsh, and Davide Scaramuzza. Introduction to Autonomous Mobile Robots. MIT Press, 2nd edition, 2011.

```
%matplotlib inline
import matplotlib.pyplot as plt
import numpy as np
import pickle
from Ransac.ransac import Ransac
```

```
testdata = []
for i in range(6):
    with open('/Users/hdumcke/git/lidar-simulator/jupyternb/data/testLineExtraction%s.
mat.pickle' % str(i+1), 'rb') as f:
        testdata.append(pickle.load(f))
```

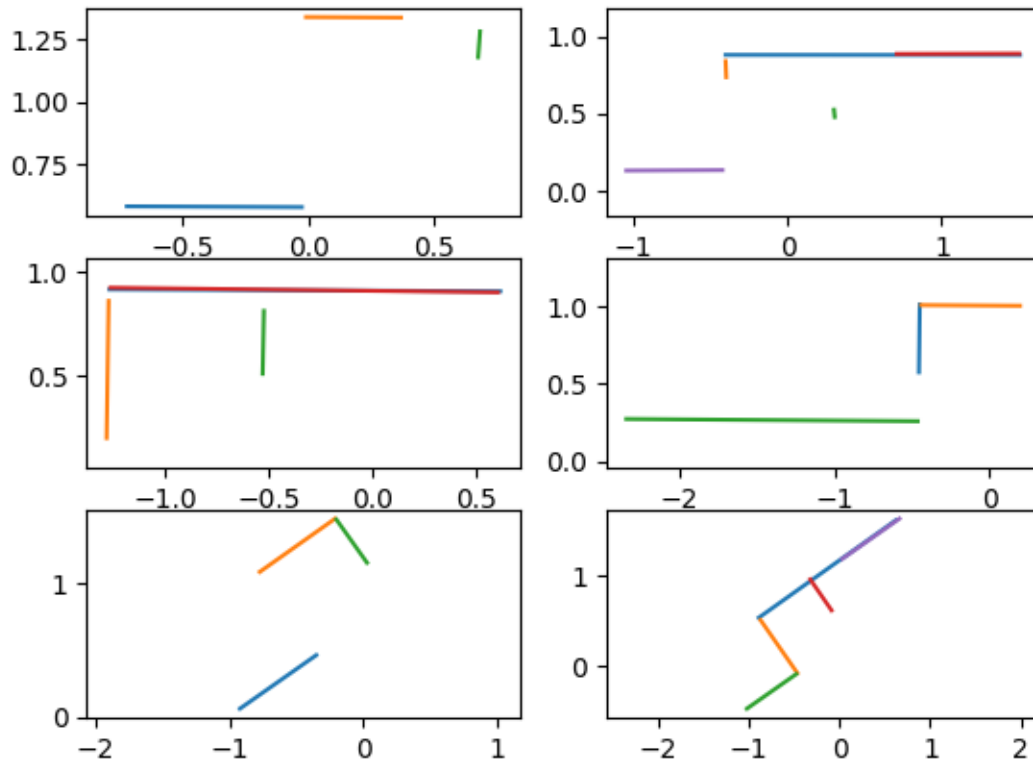
```
fig, axs = plt.subplots(3, 2)
for i in range(6):
    row_index = int(i / 2)
    column_index = i % 2
    axs[row_index, column_index].axis('equal')
    x = testdata[i]['rho'] * np.cos(testdata[i]['theta'])
    y = testdata[i]['rho'] * np.sin(testdata[i]['theta'])
    axs[row_index, column_index].scatter(x, y)
```



```
rs = Ransac(num_iterations=20, dist_threshold=0.005, min_points_per_segment=15)

fig, axs = plt.subplots(3, 2)
for i in range(6):
    row_index = int(i / 2)
    column_index = i % 2
    axs[row_index, column_index].axis('equal')
    x = testdata[i]['rho'] * np.cos(testdata[i]['theta'])
    y = testdata[i]['rho'] * np.sin(testdata[i]['theta'])
    segend, seglen, pointIdx_a = rs.extractLines(x[0], y[0])
    for j in range(segend.shape[0]):
        axs[row_index, column_index].plot([segend[j,0], segend[j,2]], [segend[j,1],
↪segend[j,3]])
```





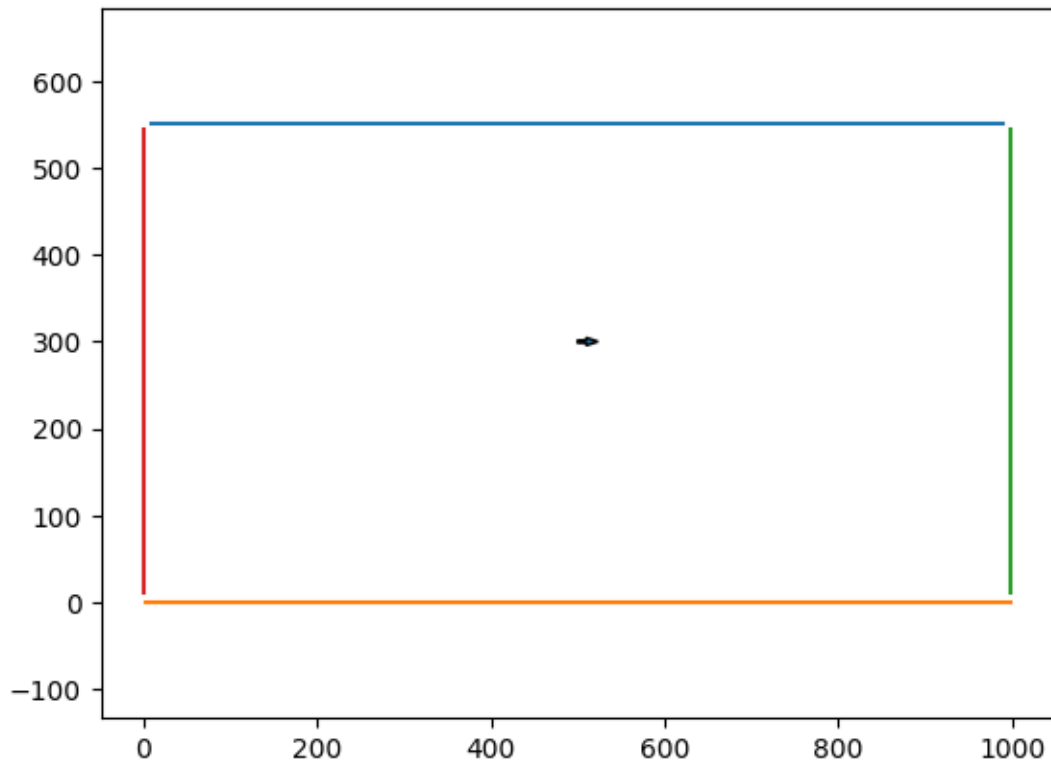
```
from LidarSim.lidar_sim import LidarSimulator
lidar = LidarSimulator("rectangle.stl")
```

```
point = [500, 300]
yaw = np.radians(0)
plot_scan = lidar.get_lidar_points(point[0], point[1], yaw)
```

```
# get cartesian coordinates
x = []
y = []
for alpha, r in plot_scan:
    x.append(r * np.cos(alpha) + point[0])
    y.append(r * np.sin(alpha) + point[1])
```

```
rs = Ransac()

fig, ax = plt.subplots()
ax.axis('equal')
plt.arrow(point[0], point[1], 10 * np.cos(yaw), 10 * np.sin(yaw), width=3.0)
# plt.scatter(x[211:329], y[211:329])
segend, seglen, pointIdx_a = rs.extractLines(x, y)
for j in range(segend.shape[0]):
    ax.plot([segend[j,0], segend[j,2]], [segend[j,1], segend[j,3]])
```



```
import Utilities.utilities as utilities
```

```
from LidarSim.lidar_sim import LidarSimulator
lidar = LidarSimulator("maze.stl")
triangles = lidar.get_map_triangles()
point = [25, 25]
yaw = np.radians(90)
```

```
pose = [[25, 25, 90 ],
        [25, 175, 90 ],
        [25, 175, 0 ],
        [75, 175, 0 ],
        [75, 125, 0 ],
        [25, 325, 90 ]]

fig, axs = plt.subplots(3, 2, figsize=(15, 15))
for i in range(6):
    row_index = int(i / 2)
    column_index = i % 2
    axs[row_index, column_index].axis('equal')
    point = pose[i][0:2]
    yaw = np.radians(pose[i][2])
    plot_scan = lidar.get_lidar_points(point[0], point[1], yaw)
    # get cartesian coordinates
    x = []
    y = []
    for alpha, r in plot_scan:
        x.append(r * np.cos(alpha) + point[0])
```

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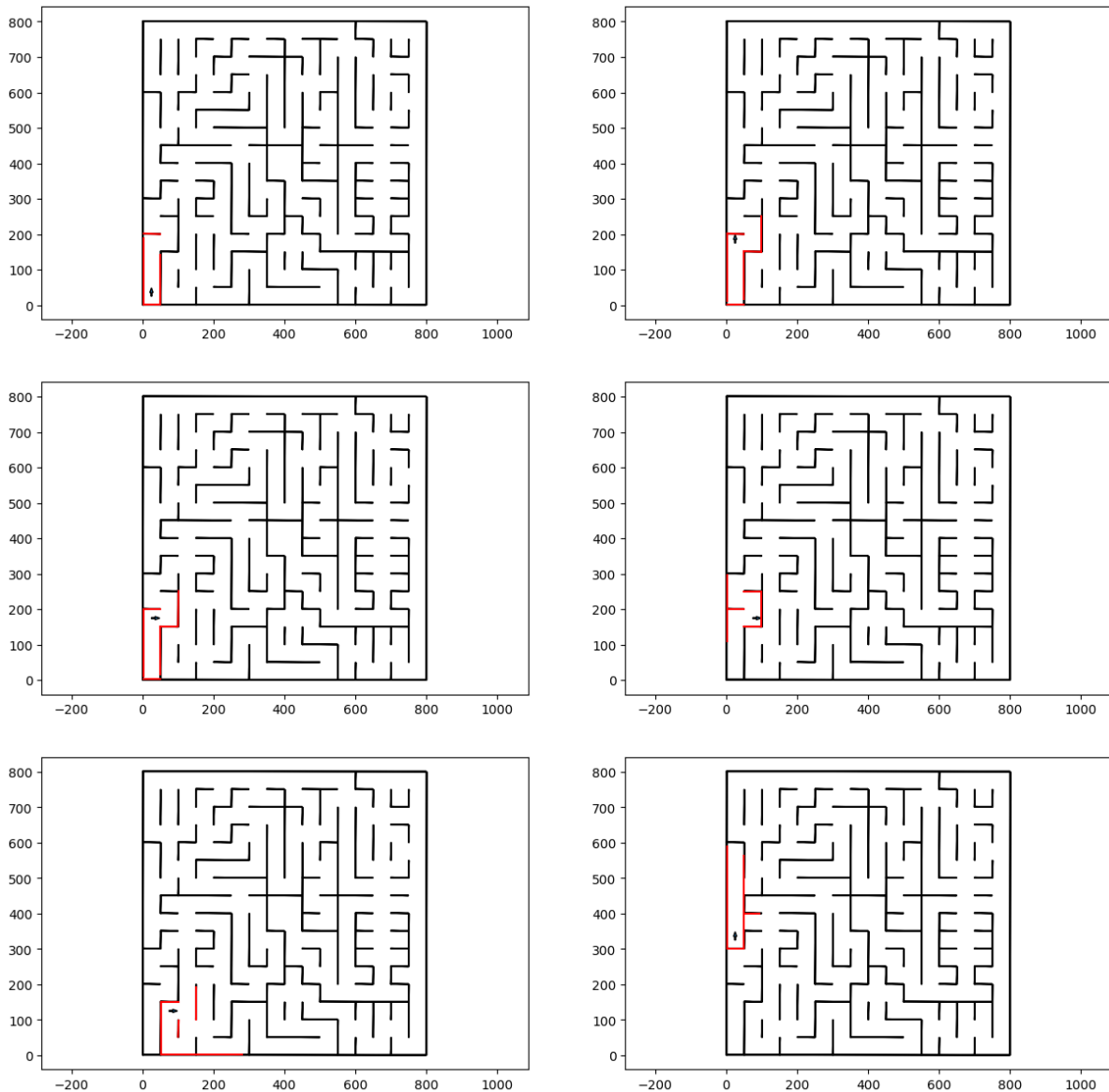
```

y.append(r * np.sin(alpha) + point[1])

rs = Ransac()

axs[row_index, column_index].arrow(point[0], point[1], 10 * np.cos(yaw), 10 * np.
sin(yaw), width=3.0)
segend, seglen, pointIdx_a = rs.extractLines(x, y)
for j in range(segend.shape[0]):
    x_p, y_p = utilities.rotate_segend(segend[j], point, yaw)
    axs[row_index, column_index].plot(x_p, y_p, 'r')
for t in triangles:
    axs[row_index, column_index].fill(t[:, 0], t[:, 1], fill=False)

```



```

from LidarSim.lidar_sim import LidarSimulator
lidar = LidarSimulator("racetrack.stl")

```

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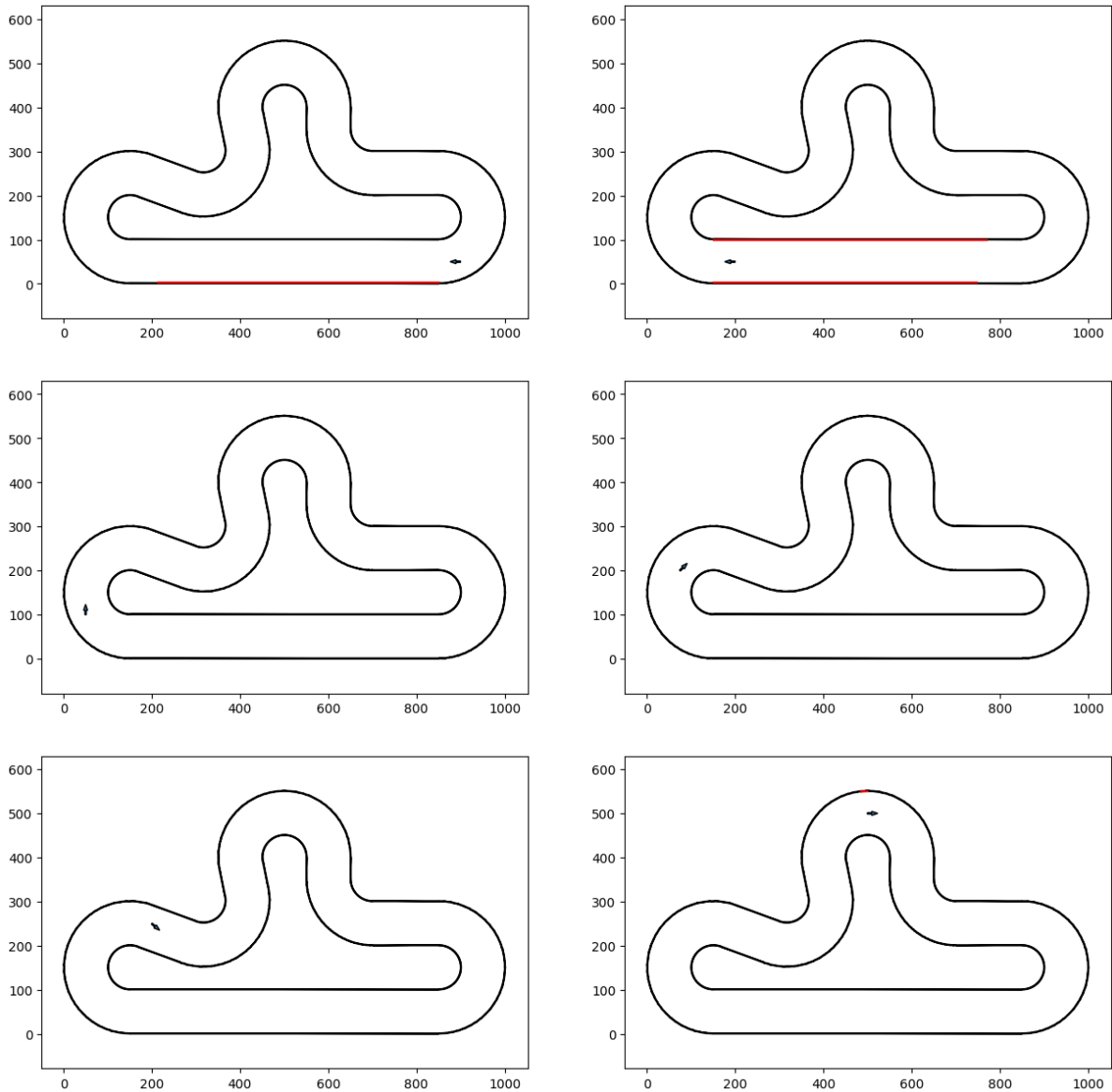
```
triangles = lidar.get_map_triangles()
yaw = np.radians(180)
```

```
pose = [[900, 50, 180 ],
        [200, 50, 180 ],
        [50, 100, 90 ],
        [75, 200, 45 ],
        [200, 250, -40 ],
        [500, 500, 0 ]]

fig, axs = plt.subplots(3, 2, figsize=(15, 15))
for i in range(6):
    row_index = int(i / 2)
    column_index = i % 2
    axs[row_index, column_index].axis('equal')
    point = pose[i][0:2]
    yaw = np.radians(pose[i][2])
    plot_scan = lidar.get_lidar_points(point[0], point[1], yaw)
    # get cartesian coordinates
    x = []
    y = []
    for alpha, r in plot_scan:
        x.append(r * np.cos(alpha)+ point[0])
        y.append(r * np.sin(alpha) + point[1])

    rs = Ransac()

    axs[row_index, column_index].arrow(point[0], point[1], 10 * np.cos(yaw), 10 * np.
    sin(yaw), width=3.0)
    segend, seglen, pointIdx_a = rs.extractLines(x, y)
    for j in range(segend.shape[0]):
        x_p, y_p = utilities.rotate_segend(segend[j], point, yaw)
        axs[row_index, column_index].plot(x_p, y_p, 'r')
    for t in triangles:
        axs[row_index, column_index].fill(t[:, 0],t[:, 1],fill=False)
```



## 16.2 Issues with Ransac

Ransac finds infinite lines. If two segments are on the same line they are considered the same as we can see with the maze example

Ransac is also non deterministic due to the random sample selection. Run the examples several times to see the changes



## ALGORITHM 5: HOUGH TRANSFORM

1. Initial: let A be a set of N points
2. Initialize the accumulator array by setting all elements to 0
3. Construct values for the array
4. Choose the element with max. votes  $V_{\max}$
5. If  $V_{\max}$  is less than a threshold, terminate
6. Otherwise, determine the inliers
7. Fit a line through the inliers and store the line
8. Remove the inliers from the set, go to step 2

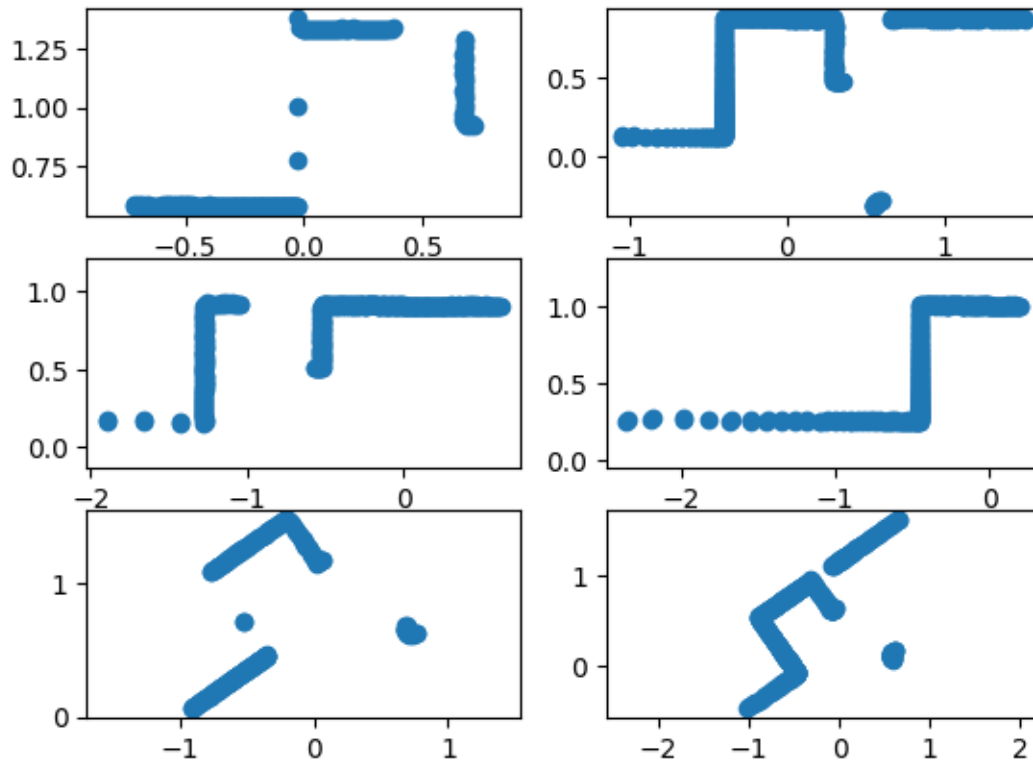
## 17.1 Reference

Roland Siegwart, Illah Nourbakhsh, and Davide Scaramuzza. Introduction to Autonomous Mobile Robots. MIT Press, 2nd edition, 2011.

```
%matplotlib inline
import matplotlib.pyplot as plt
import numpy as np
import pickle
from HoughTransform.hough_transform import HoughTransform
```

```
testdata = []
for i in range(6):
    with open('/Users/hdumcke/git/lidar-simulator/jupyternb/data/testLineExtraction%s.
    mat.pickle' % str(i+1), 'rb') as f:
        testdata.append(pickle.load(f))
```

```
fig, axs = plt.subplots(3, 2)
for i in range(6):
    row_index = int(i / 2)
    column_index = i % 2
    axs[row_index, column_index].axis('equal')
    x = testdata[i]['rho'] * np.cos(testdata[i]['theta'])
    y = testdata[i]['rho'] * np.sin(testdata[i]['theta'])
    axs[row_index, column_index].scatter(x, y)
```



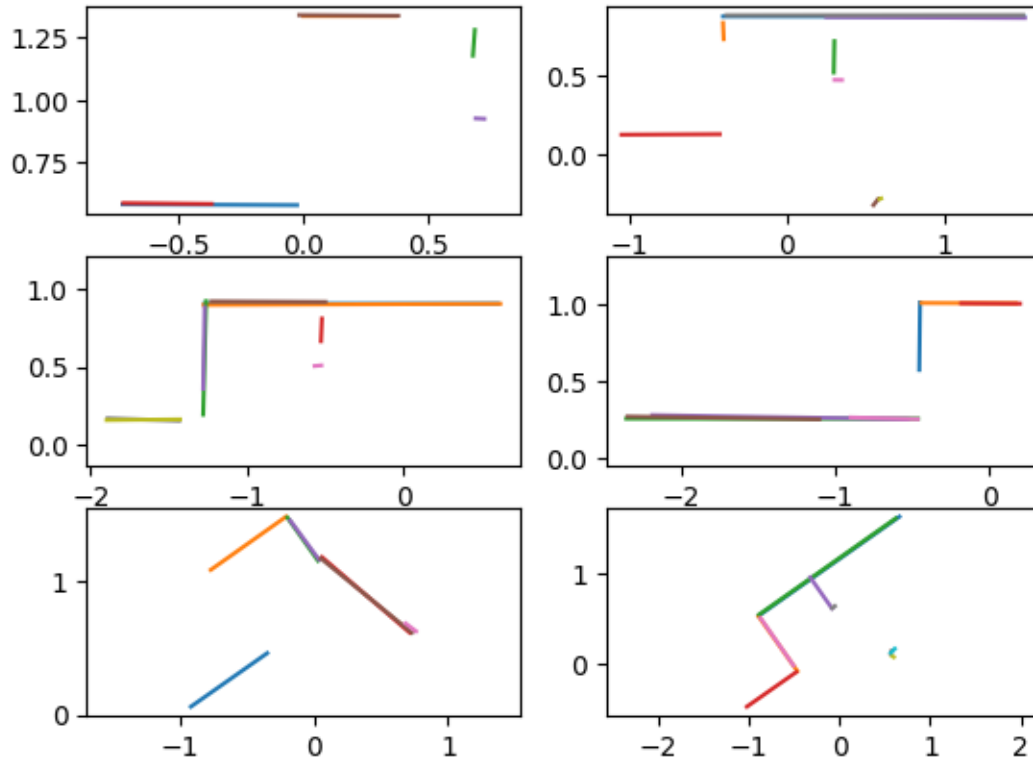
```

ht = HoughTransform()

fig, axs = plt.subplots(3, 2)
for i in range(6):
    row_index = int(i / 2)
    column_index = i % 2
    axs[row_index, column_index].axis('equal')
    x = testdata[i]['rho'] * np.cos(testdata[i]['theta'])
    y = testdata[i]['rho'] * np.sin(testdata[i]['theta'])
    segend, seglen, pointIdx_a = ht.extractLines(x[0], y[0])
    for j in range(len(segend)):
        axs[row_index, column_index].plot([segend[j][0], segend[j][2]], [segend[j][1],
↪ segend[j][3]])

```





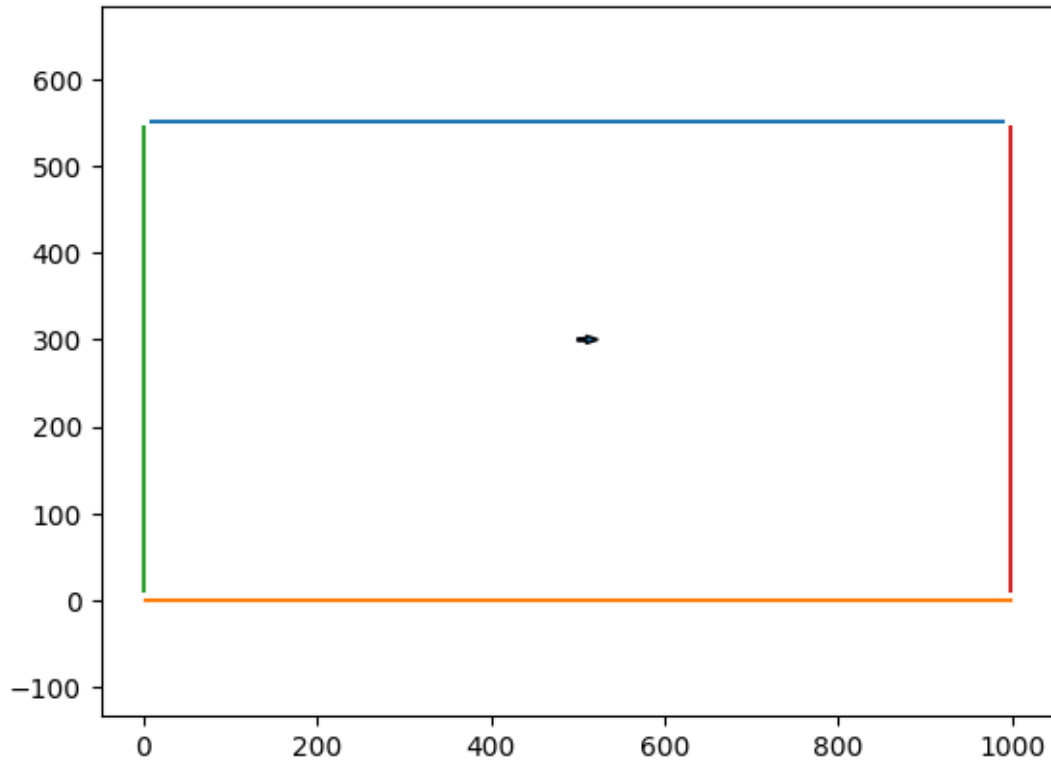
```
from LidarSim.lidar_sim import LidarSimulator
lidar = LidarSimulator("rectangle.stl")
```

```
point = [500, 300]
yaw = np.radians(0)
#plot_scan = lidar.get_lidar_points(point[0], point[1], yaw, theta=0, view_range=25)
plot_scan = lidar.get_lidar_points(point[0], point[1], yaw)
```

```
# get cartesian coordinates
x = []
y = []
for alpha, r in plot_scan:
    x.append(r * np.cos(alpha) + point[0])
    y.append(r * np.sin(alpha) + point[1])
```

```
ht = HoughTransform(rho_samples=200000)

fig, ax = plt.subplots()
ax.axis('equal')
plt.arrow(point[0], point[1], 10 * np.cos(yaw), 10 * np.sin(yaw), width=3.0)
#plt.scatter (x[211:329], y[211:329])
seglen, seglen, pointIdx_a = ht.extractLines(x, y)
for j in range(len(seglen)):
    ax.plot([seglen[j][0], seglen[j][2]], [seglen[j][1], seglen[j][3]])
```



```
import Utilities.utilities as utilities
```

```
from LidarSim.lidar_sim import LidarSimulator
lidar = LidarSimulator("maze.stl")
triangles = lidar.get_map_triangles()
point = [25, 25]
yaw = np.radians(90)
```

```
pose = [[25, 25, 90 ],
        [25, 175, 90 ],
        [25, 175, 0 ],
        [75, 175, 0 ],
        [75, 125, 0 ],
        [25, 325, 90 ]]

fig, axs = plt.subplots(3, 2, figsize=(15, 15))
for i in range(6):
    row_index = int(i / 2)
    column_index = i % 2
    axs[row_index, column_index].axis('equal')
    point = pose[i][0:2]
    yaw = np.radians(pose[i][2])
    plot_scan = lidar.get_lidar_points(point[0], point[1], yaw)
    # get cartesian coordinates
    x = []
    y = []
    for alpha, r in plot_scan:
        x.append(r * np.cos(alpha) + point[0])
```

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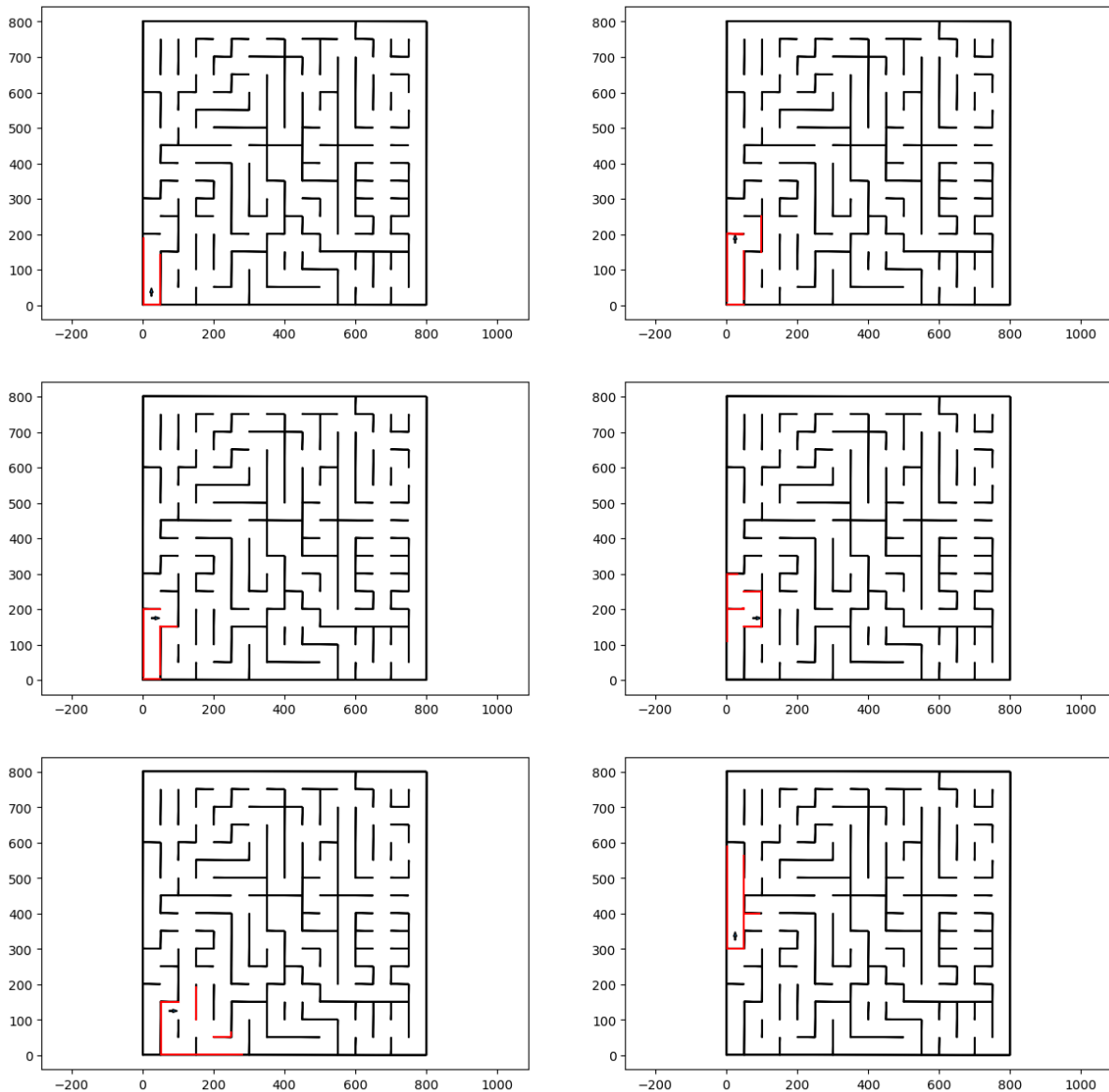
```

y.append(r * np.sin(alpha) + point[1])

ht = HoughTransform(rho_samples=200000)

axs[row_index, column_index].arrow(point[0], point[1], 10 * np.cos(yaw), 10 * np.
sin(yaw), width=3.0)
segend, seglen, pointIdx_a = ht.extractLines(x, y)
for j in range(len(segend)):
    x_p, y_p = utilities.rotate_segend(segend[j], point, yaw)
    axs[row_index, column_index].plot(x_p, y_p, 'r')
for t in triangles:
    axs[row_index, column_index].fill(t[:, 0], t[:, 1], fill=False)

```



```

from LidarSim.lidar_sim import LidarSimulator
lidar = LidarSimulator("racetrack.stl")

```

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```
triangles = lidar.get_map_triangles()
yaw = np.radians(180)
```

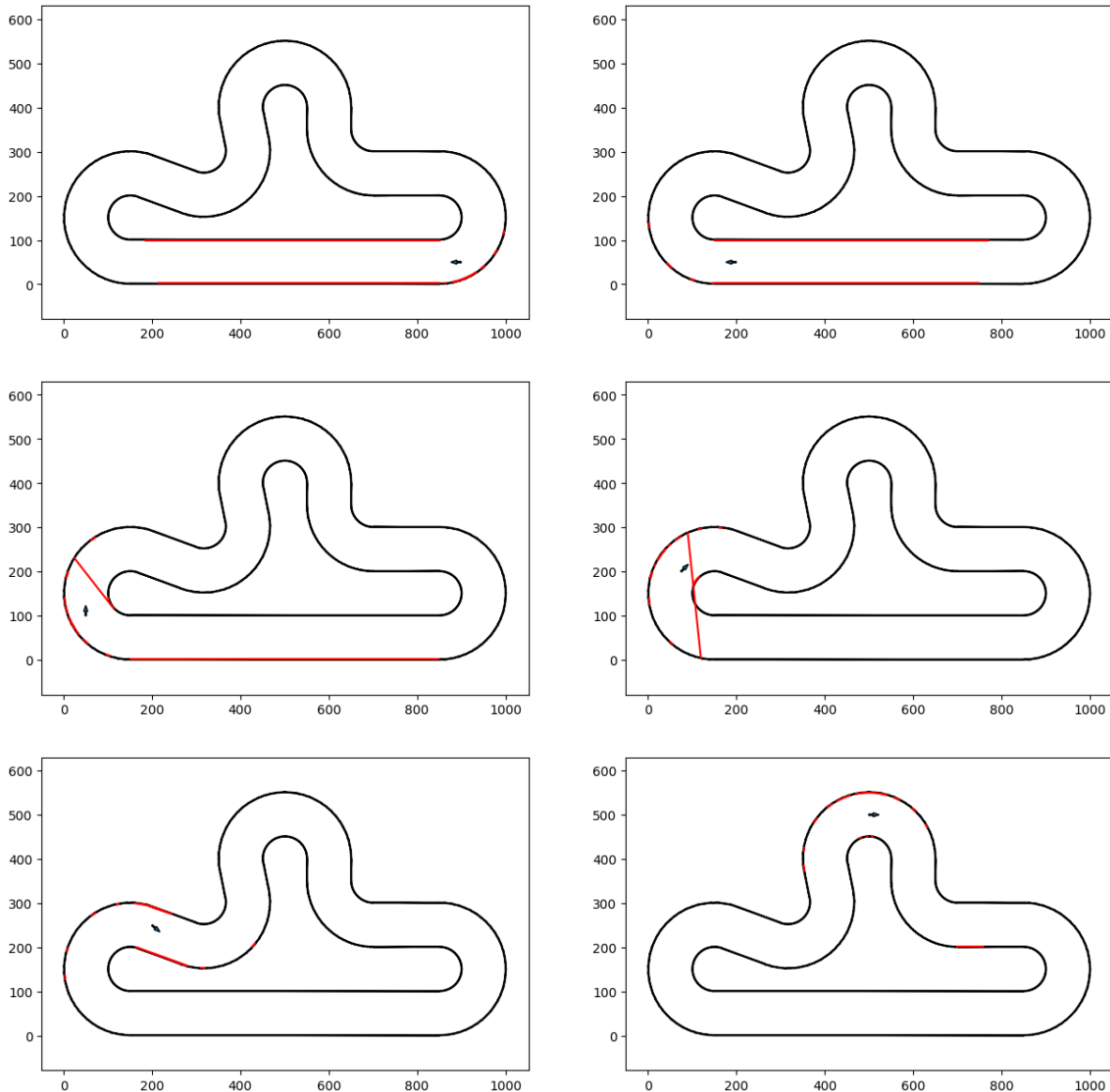
```
pose = [[900, 50, 180 ],
        [200, 50, 180 ],
        [50, 100, 90 ],
        [75, 200, 45 ],
        [200, 250, -40 ],
        [500, 500, 0 ]]

fig, axs = plt.subplots(3, 2, figsize=(15, 15))
for i in range(6):
    row_index = int(i / 2)
    column_index = i % 2
    axs[row_index, column_index].axis('equal')
    point = pose[i][0:2]
    yaw = np.radians(pose[i][2])
    plot_scan = lidar.get_lidar_points(point[0], point[1], yaw)
    # get cartesian coordinates
    x = []
    y = []
    for alpha, r in plot_scan:
        x.append(r * np.cos(alpha)+ point[0])
        y.append(r * np.sin(alpha) + point[1])

    segend, seglen, pointIdx_a = ht.extractLines(x, y)

    axs[row_index, column_index].arrow(point[0], point[1], 10 * np.cos(yaw), 10 * np.
    sin(yaw), width=3.0)
    for j in range(len(segend)):
        x_p, y_p = utilities.rotate_segend(segend[j], point, yaw)
        axs[row_index, column_index].plot(x_p, y_p, 'r')
    for t in triangles:
        axs[row_index, column_index].fill(t[:, 0],t[:, 1],fill=False)

    #axs[row_index, column_index].scatter (x, y)
```



## 17.2 Issues with Hough Transform

The Hough Transform finds infinite lines. If two segments are on the same line they are considered the same as we can see with the maze example

The Hough Transform allows to detect circles. In the case of the race track we know that we are only looking for circles with a radius of either 50 or 150 which simplifies the parameter space significantly.

For the circle detection I used <https://github.com/PavanGJ/Circle-Hough-Transform.git> as inspiration

More work would be required to make this useful like filtering relevant arc of the circles and detecting smaller lines when exiting a circle

```
pose = [[900, 50, 180 ],
        [200, 50, 180 ],
```

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```

        [50, 100, 90 ],
        [75, 200, 45 ],
        [200, 250, -40 ],
        [500, 500, 0 ]]

circles = [[], [], [], [], [], []]

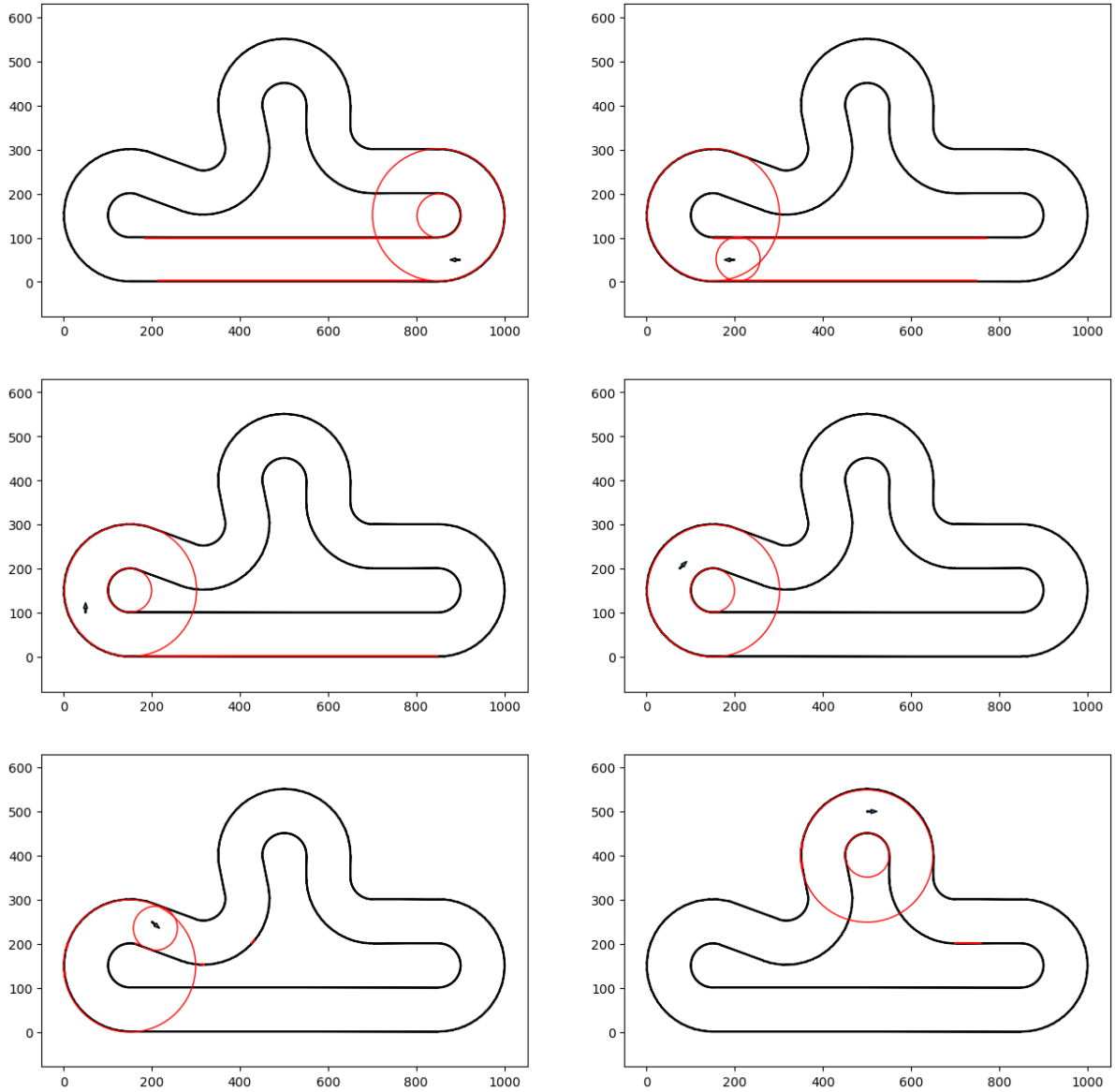
fig, axs = plt.subplots(3, 2, figsize=(15, 15))
for i in range(6):
    row_index = int(i / 2)
    column_index = i % 2
    axs[row_index, column_index].axis('equal')
    point = pose[i][0:2]
    yaw = np.radians(pose[i][2])
    plot_scan = lidar.get_lidar_points(point[0], point[1], yaw)
    # get carthesian coordinates
    x = []
    y = []
    for alpha, r in plot_scan:
        x.append(r * np.cos(alpha)+ point[0])
        y.append(r * np.sin(alpha) + point[1])

    segend, seglen, pointIdx_a = ht.extractCircles(x, y)

    axs[row_index, column_index].arrow(point[0], point[1], 10 * np.cos(yaw), 10 * np.
    ↪sin(yaw), width=3.0)
    for j in range(len(segend)):
        if len(segend[j]) == 3:
            x_c, y_c = utilities.rotate_segend([segend[j][1], segend[j][2], 0, 0], ↪
            ↪point, yaw)
            circles[i].append(plt.Circle((x_c[0],y_c[0]),segend[j][0],color=(1,0,0),
            ↪fill=False))
        else:
            x_p, y_p = utilities.rotate_segend(segend[j], point, yaw)
            axs[row_index, column_index].plot(x_p, y_p, 'r')
    for t in triangles:
        axs[row_index, column_index].fill(t[:, 0],t[:, 1],fill=False)
    for c in circles[i]:
        axs[row_index, column_index].add_patch(c)

    #axs[row_index, column_index].scatter (x, y)

```







## ALGORITHM 6: EXPECTATION MAXIMIZATION

```
1. Initial: let A be a set of N points
2. repeat
3.   Randomly generate parameters for a line
4.   Initialize weights for remaining points
5.   repeat
6.     E-Step: Compute the weights of the points from the line model
7.     M-Step: Recompute the line model parameters
8.     until Maximum number of steps reached or convergence
9.   until Maximum number of trials reached or found a line
10. If found, store the line, remove the inliers, go to step 2
11. Otherwise, terminate
```

### 18.1 Reference

Forsyth, D. A., Ponce, J., Computer Vision: A Modern Approach. Upper Saddle River, NJ, Prentice Hall, 2003.

### 18.2 ToDo



## RANGE HISTOGRAM FEATURES

### 19.1 ToDO



## LOCALIZATION

### 20.1 ToDo



## ODOMETRY SIMULATOR

### 21.1 ToDo

```
%matplotlib inline
import matplotlib.pyplot as plt
import numpy as np
import Utilities.utilities as utilities
```

```
pose = [0.0, 0.0, 0.0]
cmd_vel = {}
cmd_vel["linear_x"] = 1.0
cmd_vel["linear_y"] = 1.0
cmd_vel["angular_z"] = 0.0
dt = 0.1
```

```
fig,ax = plt.subplots()
position = pose
ax.axis('equal')
plt.arrow(position[0], position[1], 10 * np.cos(position[2]), 10 * np.
    ↪sin(position[2]), width=3.0)
for i in range(10):
    position = utilities.get_odom(cmd_vel, position, dt)
    print(position)
    plt.arrow(position[0], position[1], 10 * np.cos(position[2]), 10 * np.
    ↪sin(position[2]), width=3.0)
```

```
[0.1, 0.0, 0.0]
[0.2, 0.0, 0.0]
[0.30000000000000004, 0.0, 0.0]
[0.4, 0.0, 0.0]
[0.5, 0.0, 0.0]
[0.6, 0.0, 0.0]
[0.7, 0.0, 0.0]
[0.7999999999999999, 0.0, 0.0]
[0.8999999999999999, 0.0, 0.0]
[0.9999999999999999, 0.0, 0.0]
```

