Scientific Experimentation and Evaluation
Assignment 4

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1 Experiment 1: Manual Motion Observation

Aim: Manually measure the observable pose variation for three different constant velocity motions(Straight line, an arc to the left and an arc to the right) of LEGO NXT differential drive robot.

1.1 Setup

- The measurement system for the experiment includes a white sheet, a LEGO NXT differential drive robot, two pencils and a scale.
- The **device under test(DUT)** is the LEGO NXT differential drive robot.
- Clamp the sheet on a flat surface.
- Construct a robot equipped with two pencils in the front, such that the start and end positions can be marked. This constitutes the measurement facility.
- Install *Lejos* OS on the robot.
- Program the robot for it to run in three different constant velocity motions.

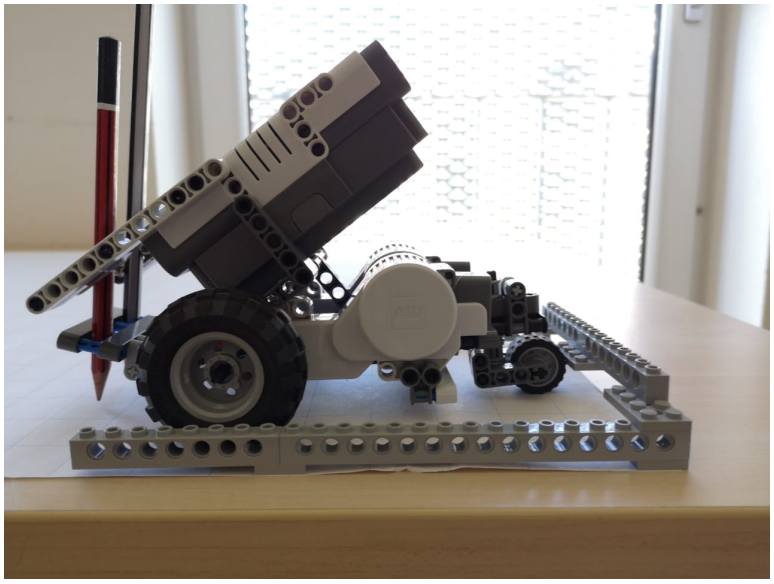

Figure 1: Side view

Figure 2: Isometric view

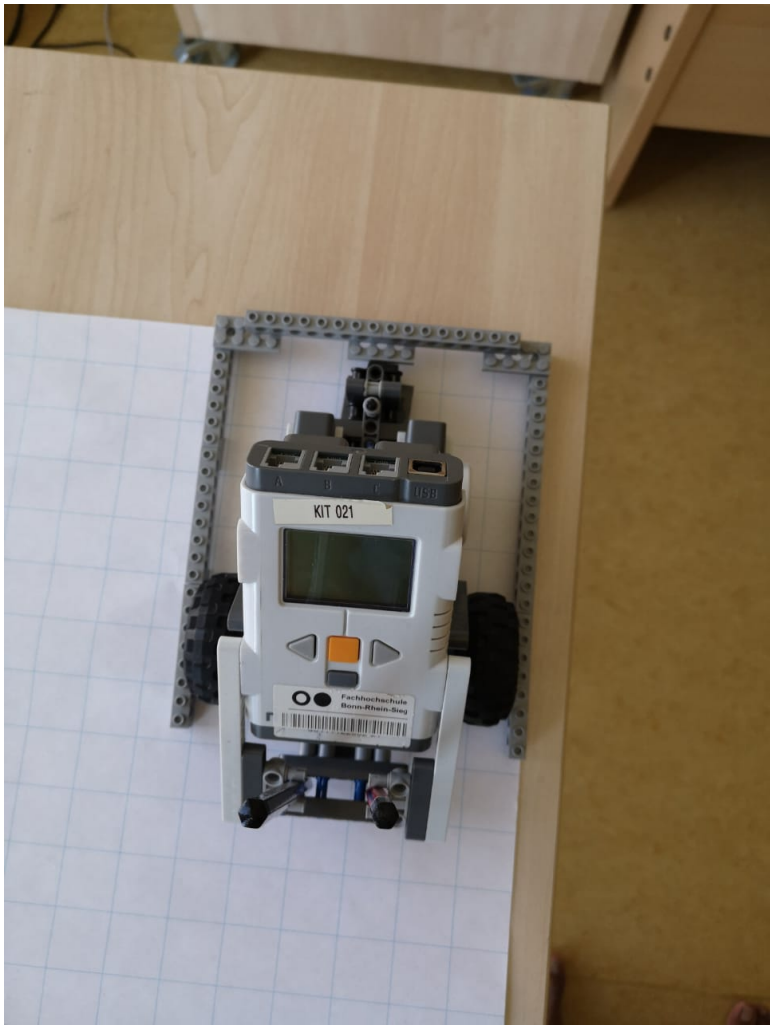

Figure 3: Top view

Figure 4: front view

1.2 Procedure

- The experimental setup is prepared.
- The starting position of the robot is marked with the help of the points drawn by the pencils.
- The center of the line joining the two points is recorded which will be the starting position of the robot.
- The robot is programmed to run in a straight line forward motion at a constant speed.
- Once the robot stops, the center of the line joining the last points drawn by the pencils will be the stop position.
- The **Measurand** i.e., the relative coordinates of the robot are measured by the difference in the coordinates of the starting position and the stopping position.
- **Measurement System:** Our measurement system consists of robot placed on a cardboard sheet. Two pencils are attached in the front to mark the position. As the robot is moved, the path is traced on the sheet. After reaching the end position, the final pose is marked. The variation in the path is observed and the error in pose is measured.
- The **Measurement result** i.e., the pose variation is obtained by finding the difference between the observed values and their mean.
- Repeat the above experiment with a program to run the robot with constant angular and translational velocities for a fixed time period, such that it describes an arc to the left.
- Repeat the above experiment with a program to run the robot with constant angular and translational velocities for a fixed time period, such that it describes an arc to the right.

1.3 Expected problems

- Parallax error.
- Positional errors.
- There might be slipping of wheels.
- The tip of the pencils might break during the motion of the robot.
- The pencils will have to be sharpened as the nibs might get blunt due to continuous drawing of lines which might result in measurement errors.

1.4 Expected performance

- The Rotation sensor measures the motor rotations with the accuracy of +/- one degree for one rotation(360 degree). Hence the accuracy in distance traveled for one rotation will be +/- 0.05 cm.
- Reference: LEGO NXT: Features & Limitations.
- Assuming experiment distance = 70.4 cm.
- Expected accuracy is ±2 and expected precision is 70.4±0.2cm.

1.4.1 Reasons

- The slippage of the wheels may cause poor accuracy and precision.
- The internal errors from the DUT may cause poor accuracy but the precision won't be affected significantly, since the DUT won't be changed while doing the measurement.

1.5 Execution of the experiment:

- The starting position of the robot was marked using two pencils which were placed on either side of the robot.
- The same points were used as the starting points while repeating the experiment.
- The position of the back wheel was also kept constant by drawing a line and using the same line as reference for future run of experiments.
- The position of the back wheel impacted the outcome of the experiment as a small deviation from the reference line would result in a considerable deviation of the outcome.
- The theta values were calculated by drawing a perpendicular to the line joining the two observed end points and calculating the angle using the formula $tan^{-1}(\frac{y}{x})$. As shown in the below figure.

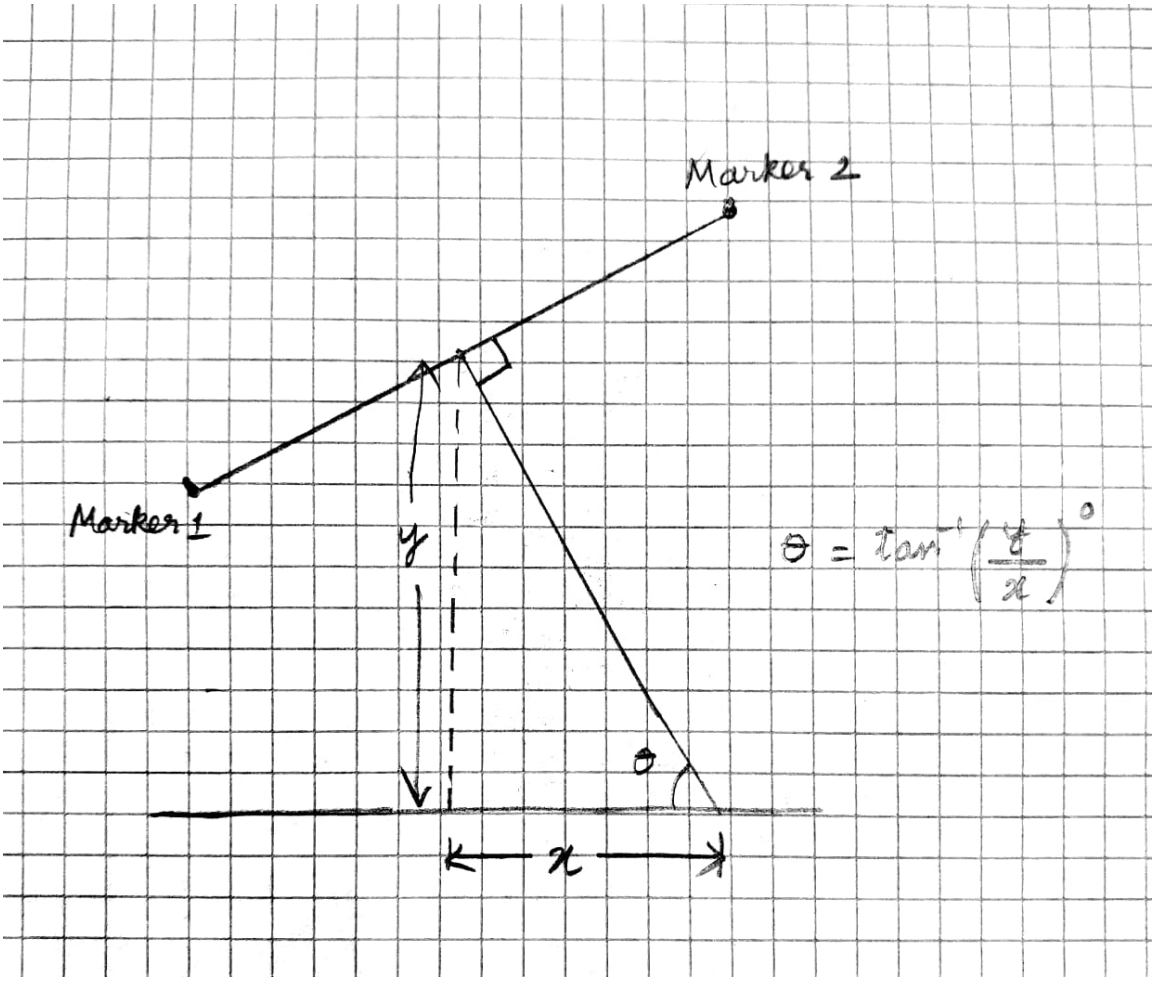

Figure 5: Finding out θ for robot orientation

- Lego NXT 2.0 software was used to program the robot with the following parameters.
  No. of rotations = 4
  Power = 40
- The battery voltage before the experiment = 8.2V.
  The battery voltage after the experimenter = 4.2V.
- The observed data has been recorded in the ".csv" format.
- There was no pre-processing of data required as no outliers were detected.
- Graph has been plotted for all three motions.

Note : All x,y measurements are in centimeters (cm)

1.5.1 Straight

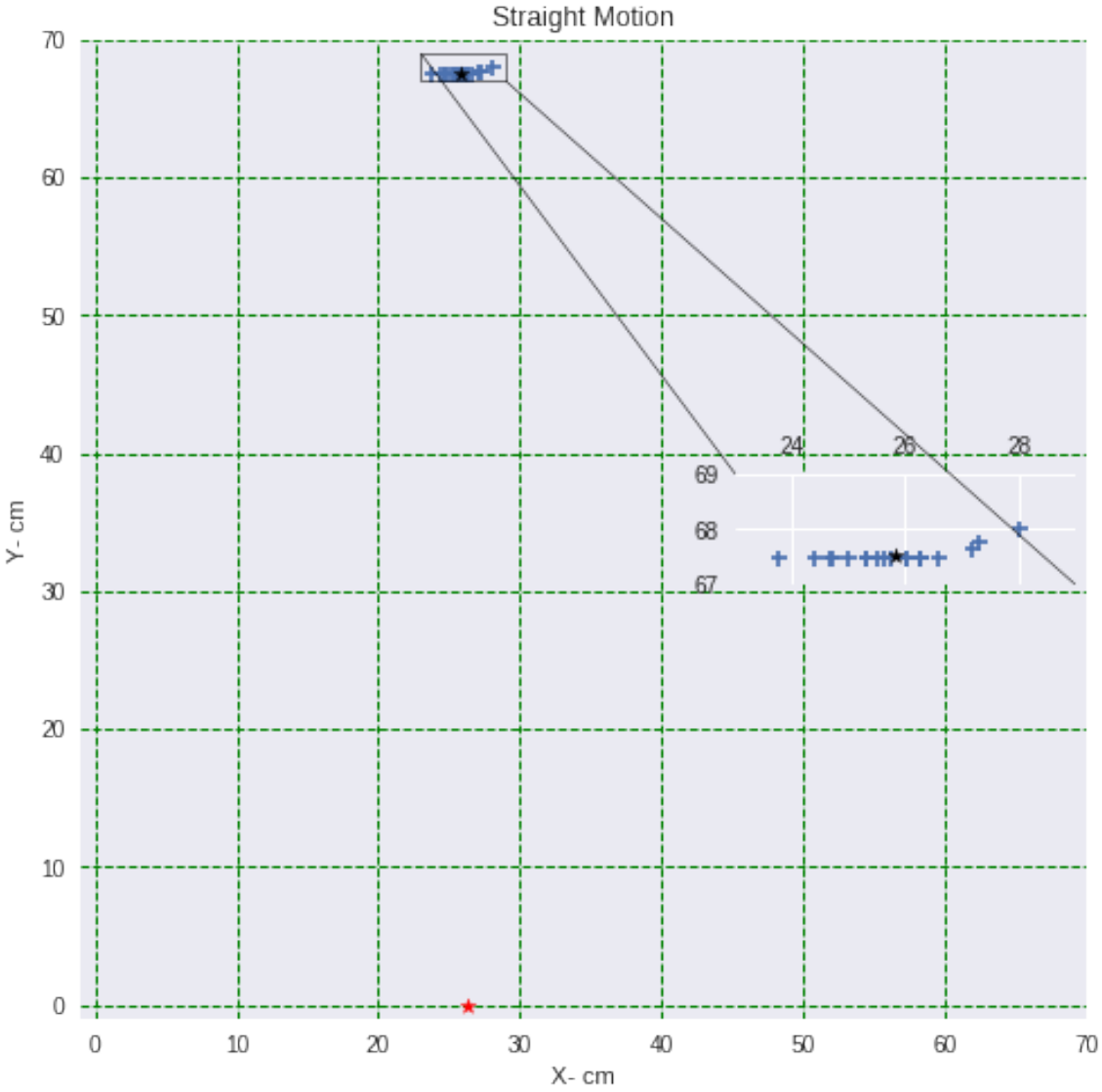

Figure 6: 20 iteration of moving straight

Fitting the distribution for the obtained data:
(X, Y) points for straight line

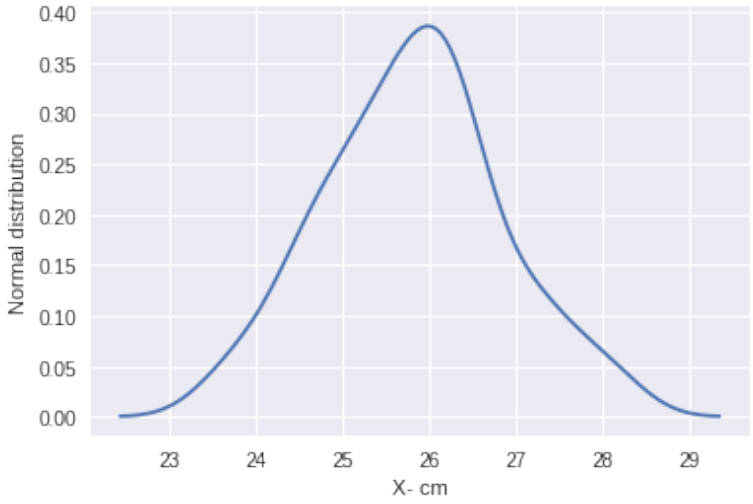

Figure 7: Normal distribution for straight motion (X in cm)

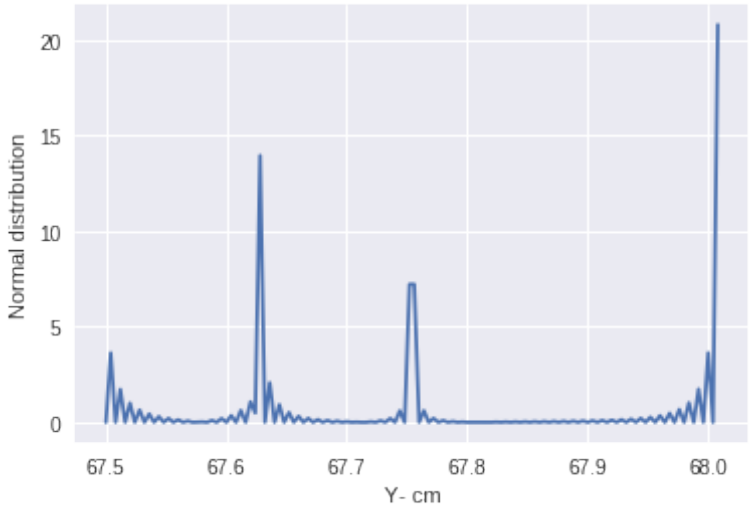

Figure 8: Normal distribution for straight motion (Y in cm)

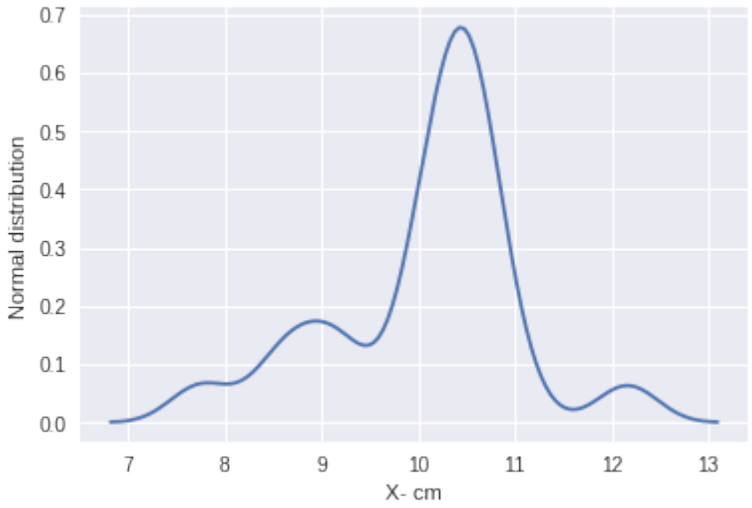

Figure 9: Normal distribution for left motion (X in cm)

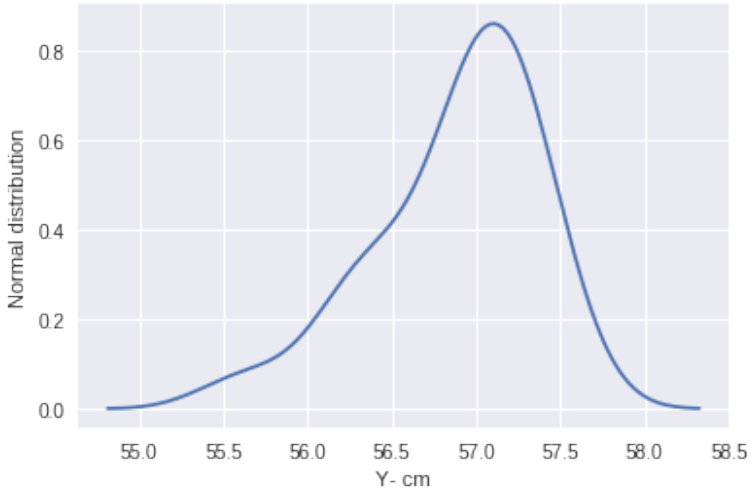

Figure 10: Normal distribution for left motion (Y in cm)

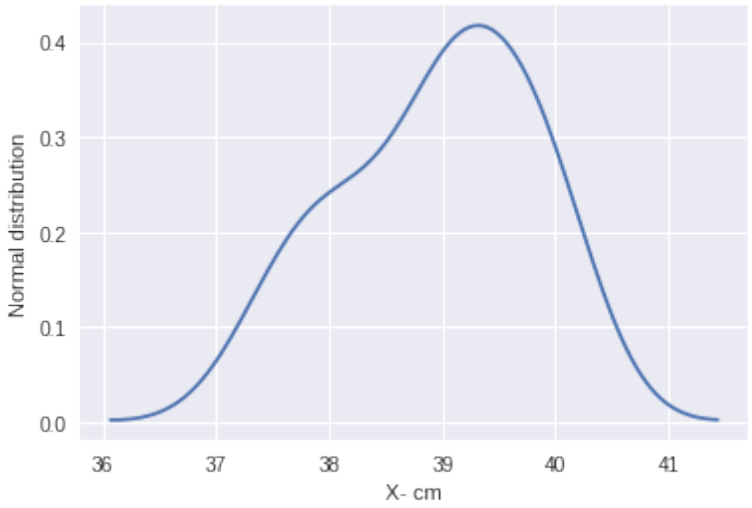

Figure 11: Normal distribution for right motion (X in cm)

- $Range(x, y) = ([23.8 - 28.1], [67.5 - 68.1])$
- $Mean(x, y) = 25.8, 67.5$
- $\sigma(x, y) = 1.1, 0.1$
- $Precision(x, y) = 25.8 \pm 1.1, 67.5 \pm 0.1$
- $Accuracy(x, y) = x \pm 0.9, y \pm 0.1$

1.5.2 Right

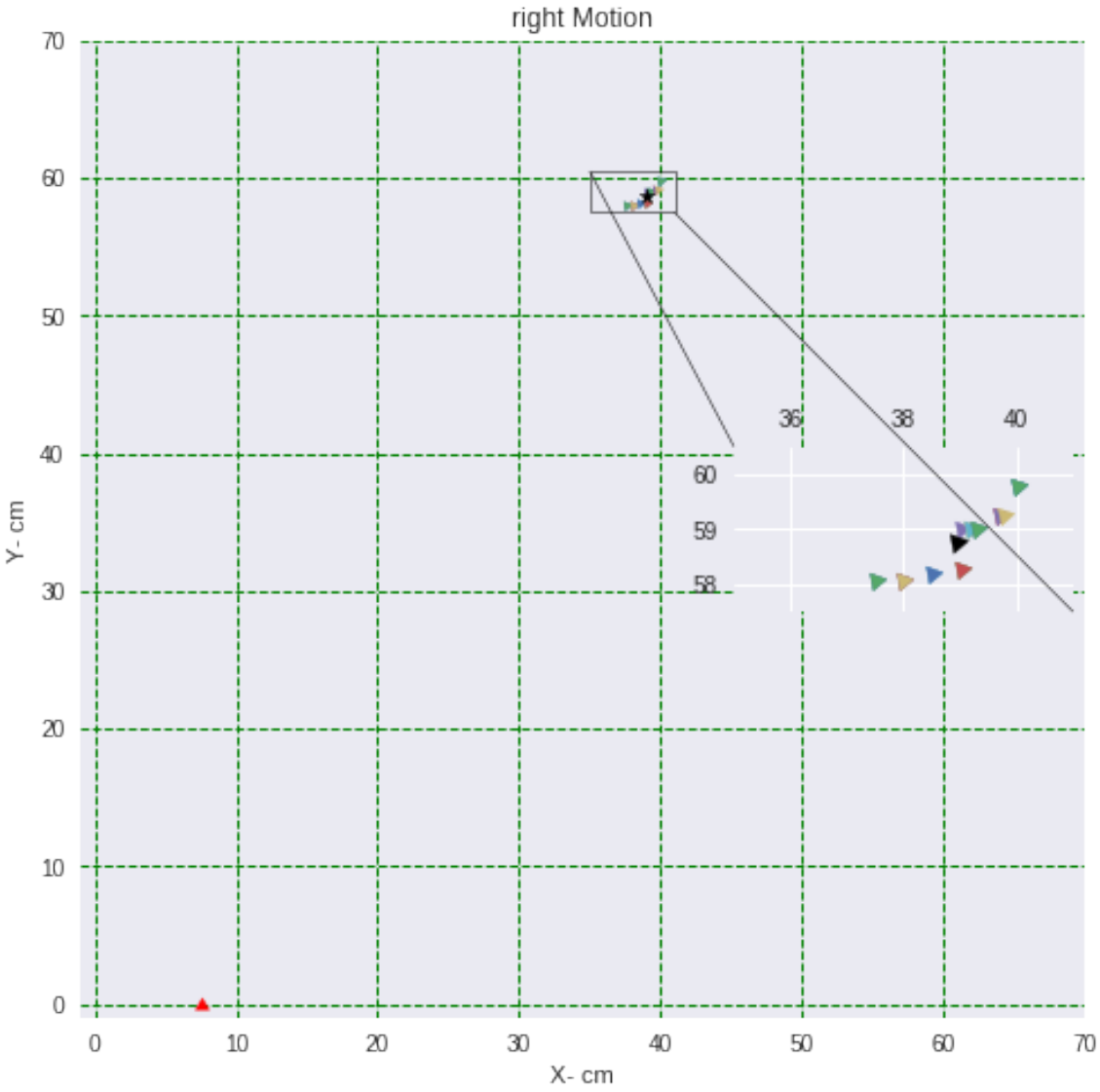

Figure 12: 20 observations of right arc motion

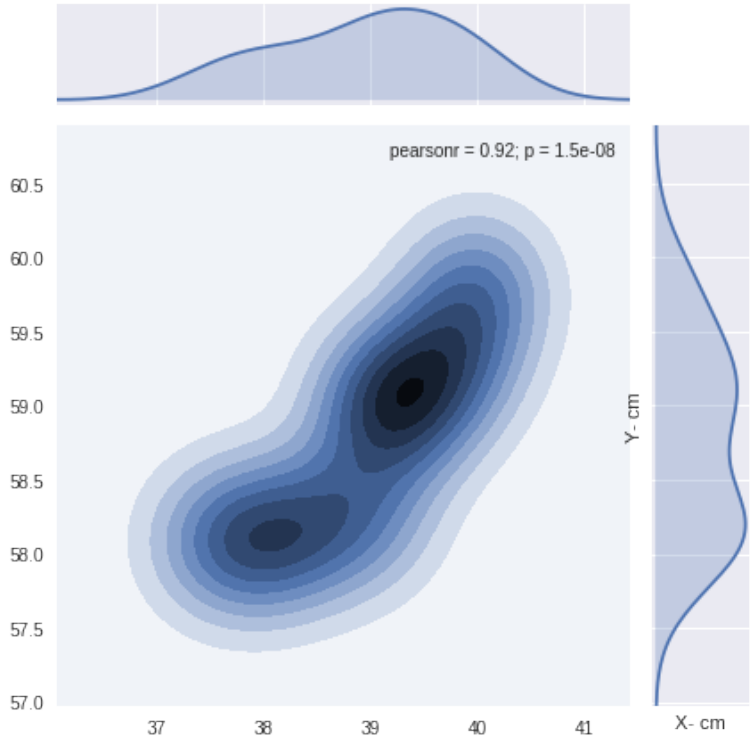

Figure 13: Density plot of (X, Y) points for right arc motion

In the above plot it can be observed that the distribution of X and Y positions for right arc motion is gaussian.

- $Range(x, y) = ([37.5 - 40.0], [58.0 - 59.8])$
- $Mean(x, y) = 38.9, 58.8$
- $Precision(x, y) = 38.9 \pm 0.8, 58.8 \pm 0.6$
- $Accuracy(x, y) = x \pm 0.5, y \pm 0.5$
- $\sigma(x, y) = 0.8, 0.6$
- $Mean(\theta) = 40.6 degree$
- $Accuracy(\theta) = \theta \pm 5 deg$
- $\sigma = 1.1 deg$
- $range(\theta) = 38.8 - 43.3$

1.5.3 Left

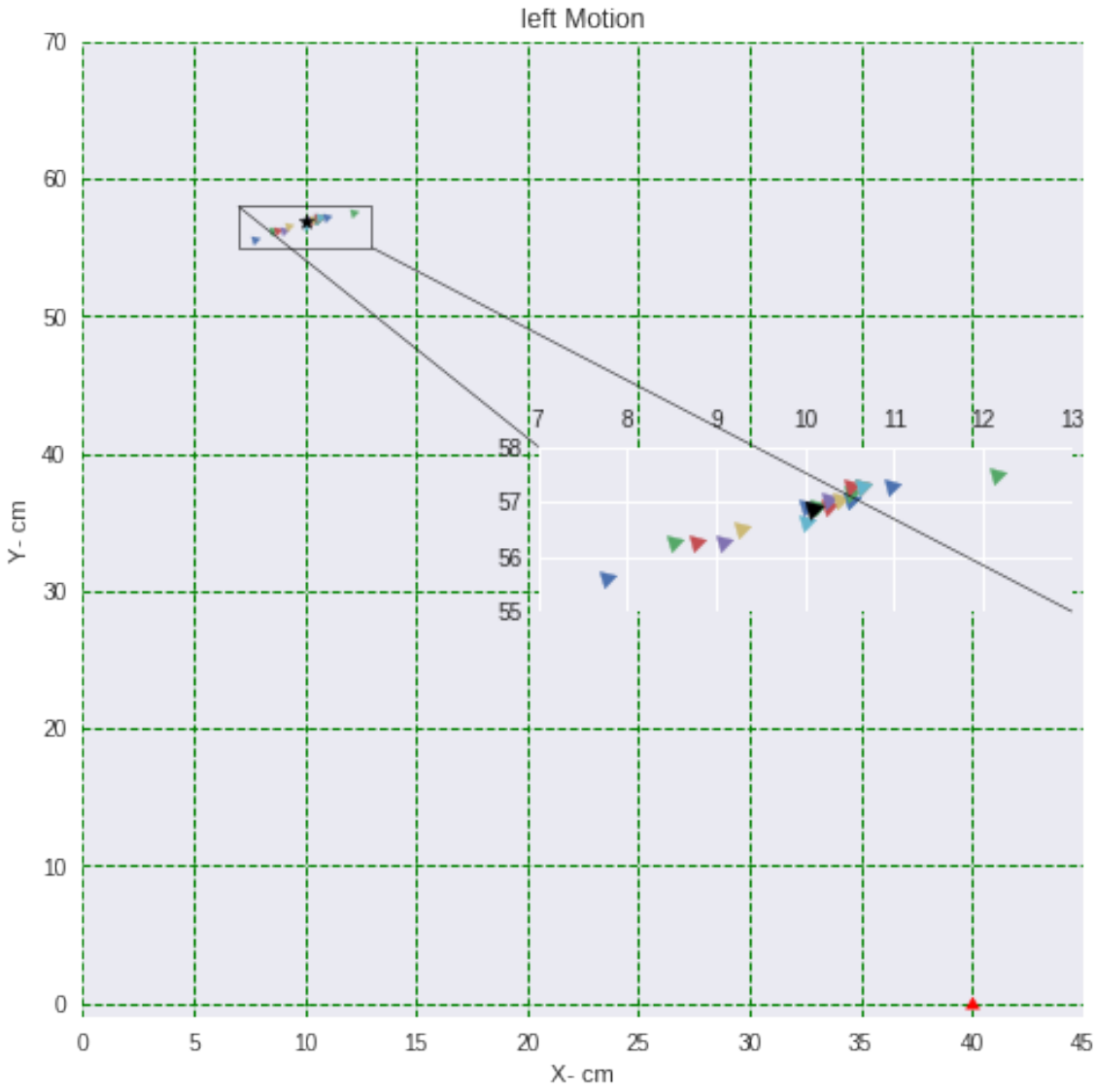

Figure 14: 20 iteration of moving left

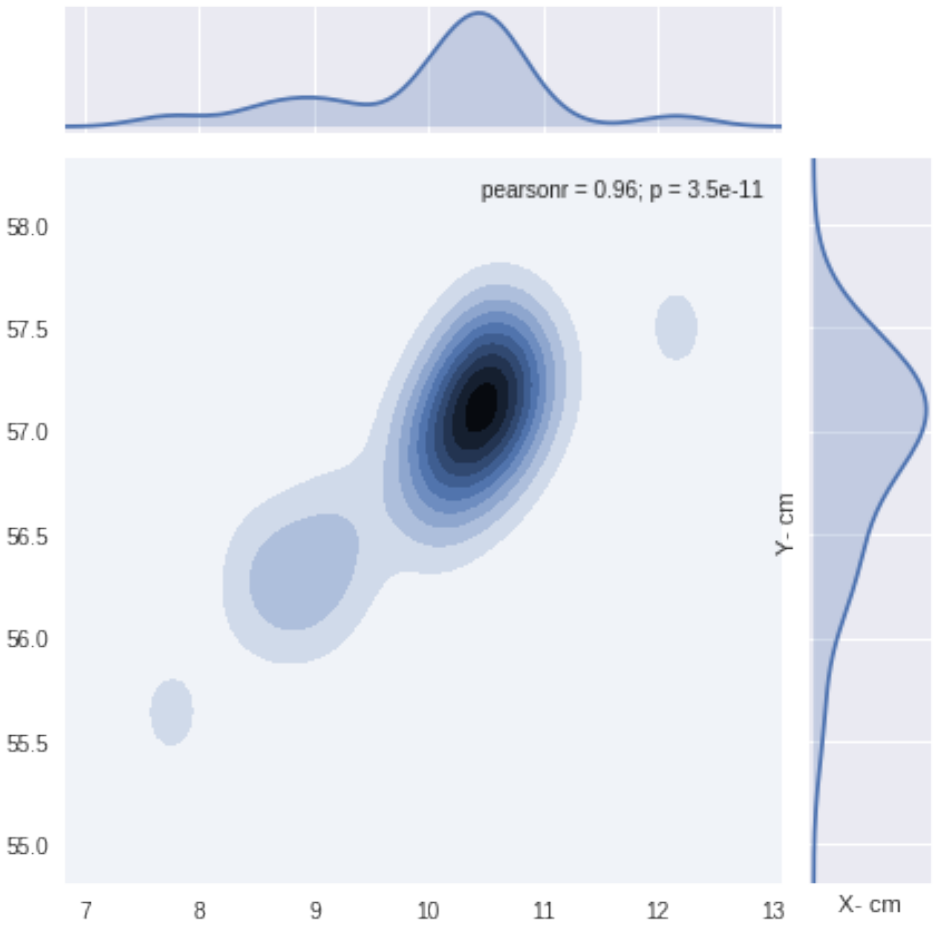

Figure 15: Density plot of (X, Y) for moving left

In the above figure it can be observed that the (X, Y) points obtained for left arc motion fit a gaussian distribution.

- $Range(x, y) = ([7.8 - 12.2], [55.6 - 57.5])$
- $Mean(x, y) = 10.1, 56.9$
- $Precision(x, y) = 10.1 \pm 2.0, 56.9 \pm 1.5$
- $Accuracy(x, y) = x \pm 0.5, y \pm 0.5$
- $\sigma(x, y) = 1.0, 0.5$
- $Mean(\theta) = 45.7 degree$
- $Accuracy(\theta) = \theta 7 \pm deg$
- $\sigma = 1.3 deg$
- $range(\theta) = 43.2 - 48.3$

Boxplots

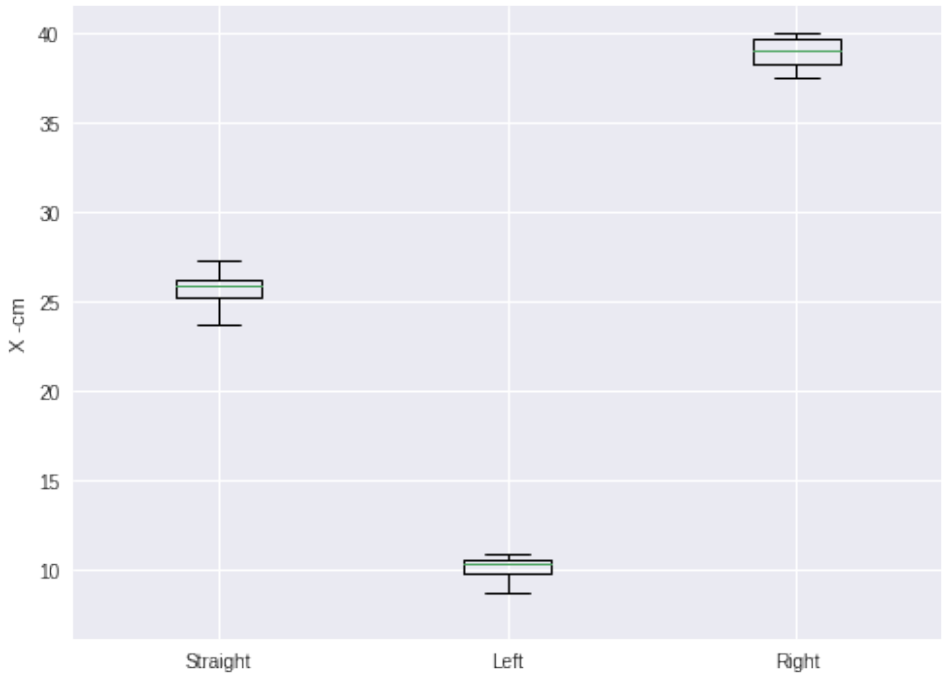

Figure 16: Box plot of X-values for straight line motion, left and right turn

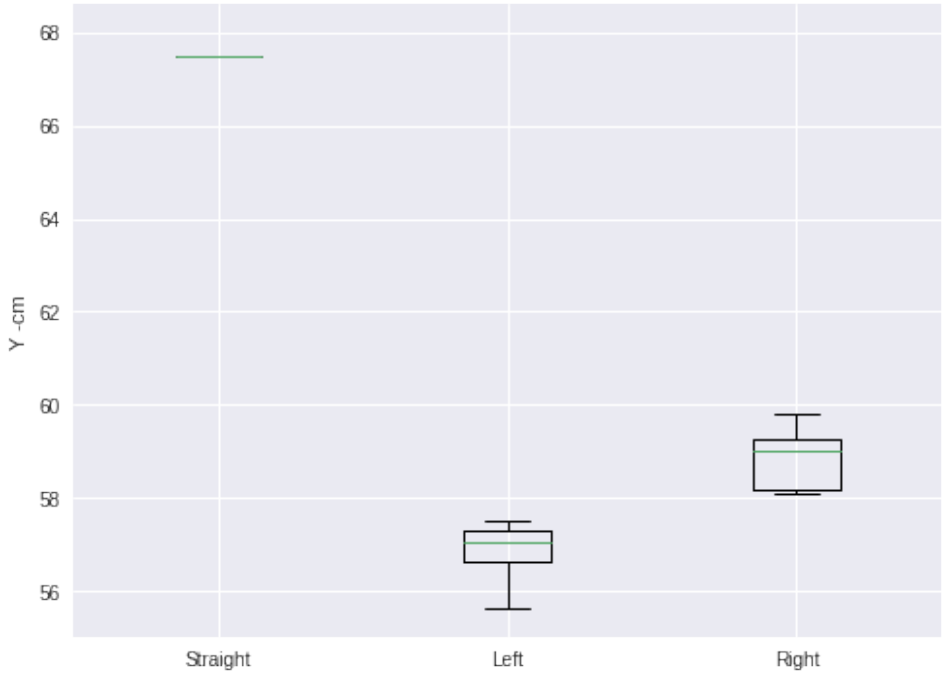

Figure 17: Box plot of Y-values for straight line motion, left and right turn

- 2.3 The chi-square test indicates that the observed data is a normal distribution.

```
In [9]: ks , p = stats.normaltest(straight_xy)
alpha = 1e-3
# null hypothesis: x comes from normal distribution
if p.all() < alpha:
print("The null hypothesis can be rejected")
else:
print("The null hypothesis cannot be rejected")

[Out]: The null hypothesis cannot be rejected

In [10]: ks , p = stats.normaltest(right_xy)
alpha = 1e-3
if p.all() < alpha:
print("The null hypothesis can be rejected")
else:
print("The null hypothesis cannot be rejected")
[Out]: The null hypothesis cannot be rejected
```

```
In [11]: ks , p = stats.normaltest(left_xy)
alpha = 1e-3
if p.all() < alpha:
print("The null hypothesis can be rejected")
else:
print("The null hypothesis cannot be rejected")
[Out]: The null hypothesis cannot be rejected
```

- Deliverable 2.2: The functions used for plotting are from the seaborn library and use a kernel density method to obtain a 2-D density plot of the x and y coordinates.
- Deliverable 2.5:

  Straight: In the case of straight line motion the readings obtained for Y coordinates appear to be both precise and accurate, whereas the X coordinates appear to be precise but not very accurate.

  Right: The readings obtained for Y coordinates appear to be both highly precise and accurate, whereas the X coordinates appear to be precise but not very accurate.

  Left: The readings obtained for X coordinates appear to have low precision and accuracy, whereas the Y coordinates appear to be precise with a low accuracy.

- **Comparison of Actual performance with expected:**
  - The expected distance to be traveled was 70.4 cm.
  - The mean actual distance travelled was calculated by applying distance formula on initial and end points, which is equal to 67.5 cm.
  - Estimated error = 2 cm
  - Actual error = 2.9 cm
  - The difference in the actual and expected error can be due to systematic or/and random errors.

2 Calibration of an Optical Tracking System

2.1 Setup for camera calibration

- The camera calibration process is done on a checkerboard (calibration tool).
- The *MATLAB support package for USB webcam* for camera calibration is installed.
- The camera calibration toolbox for MATLAB is installed.
- The webcam (Microsoft Lifecam) is connected to the laptop and the connectivity is checked by running *webcam list* command in MATLAB, which shows the list of all webcams connected to PC.
- The webcam is mounted on a static paper punch, while capturing the images.
- The Auto-focus is turned on and the camera is made to adjust to the distance of the checkerboard.
- Then the Auto-focus is turned *off* by running the following command in MATLAB,

  ```
  cam.FocalMode = manual
  ```

- The images of checkerboard are captured from different perspectives (different position and orientation in the camera's field of view). A total of 25 images are captured and saved for the calibration process.

2.2 Estimation of number of images

- Since the object in the world is 3-dimensional and the captured images are 2D (co-planar), we loose one dimension. In order to regain the actual dimensionality during the calibration, various number of images are captured in different positions and orientations.
- As a rule of thumb, 20-40 images is quite enough for the process. For our calibration we have captured 25 images in different position and orientations.
- The idea behind the capturing enough images is to cover the field of view, and to have a good distribution of 3D orientations of the board.

2.3 Description of the camera parameters

After the calibration, the list of parameters obtained are of two types.

- **Intrinsic parameters**
  - **Focal length:** Distance between the lens and the image sensor when the object is in focus, usually stated in millimeters. The focal length in pixels is stored in the 2x1 vector fc.
  - **Principal point:** A point at the intersection of the optical axis and the image plane. The principal point coordinates are stored in the 2x1 vector cc.
  - **Skew coefficient:** Defines the angle between x and y pixels axes and is stored in $alpha_c$
  - **Radial Distortions:** Symmetric distortion caused by the lens due to imperfections in curvature when the lens was ground.
- **Extrinsic parameters**
  - **Rotational Vectors:** Vector describing the rotation of camera plane with respect to the world frame.
  - **Translational vectors:** Vector describing the translation of camera plane with respect to the world frame.

2.4 Possible problems or error sources that can disturb the calibration process

- If the naming conventions of the images are different, then the calibration process gives error.
- The computer running the calibration toolbox must have enough RAM(128 MB or less). If this condition is not met, then it gives *OUT OF MEMORY* error.
- Enough number of images must be provided to get better calibration results.
- The images captured with same orientations will not provide enough inputs for the toolbox to calibrate efficiently.

2.5 Image poses used for calibration

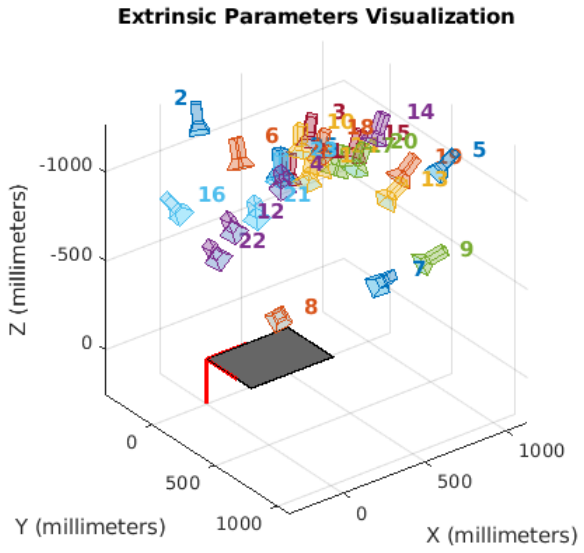

Figure 18: Pattern centric view

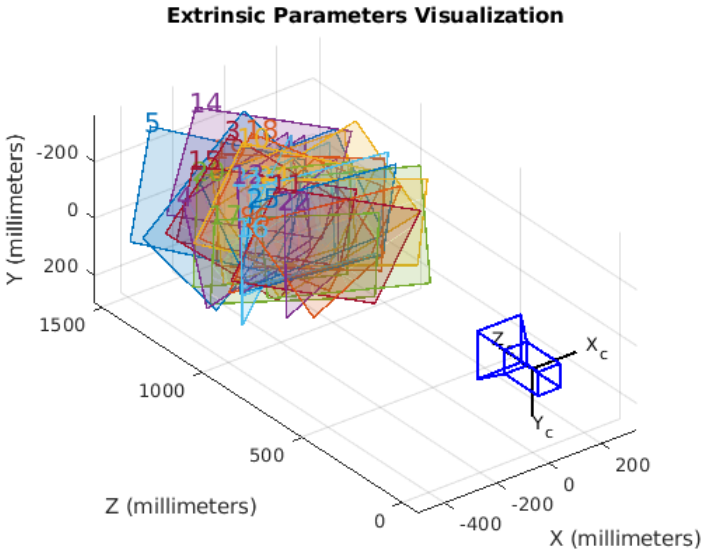

Figure 19: Camera centric view

The image poses can be approximated using the figures presented above which are obtained using the MATLAB camera calibration toolbox. How-

ever the exact values of each of these images poses is not possible to obtain due to a lack of a suitable measuring metric.

2.6 Camera parameters obtained from calibration

$$IntrinsicMatrix = \begin{bmatrix} 650.1601 & 0 & 0 \\ -0.4441 & 649.8506 & 0 \\ 417.1918 & 288.0535 & 1 \end{bmatrix}$$

$$FocalLength = \begin{bmatrix} 650.1601 & 649.8506 \end{bmatrix}$$

$$Principalpoint = \begin{bmatrix} 417.1918 & 288.0535 \end{bmatrix}$$

$$Skew = -0.4441$$

$$Radialdistortion = \begin{bmatrix} -0.0215 & 0.1424 \end{bmatrix}$$

$$TangentialDistortion = \begin{bmatrix} -9.82e-04 & 0.0010 \end{bmatrix}$$

$$Meanprojectionerror = 0.1031$$

2.7 Estimation errors

2.7.1 Intrinsic errors

$$Skewerror = 0.0641$$

$$Focallengtherror = \begin{bmatrix} 0.3066 & 0.2922 \end{bmatrix}$$

$$Principalpointerror = \begin{bmatrix} 0.4705 & 0.3823 \end{bmatrix}$$

$$Radialdistortionerror = \begin{bmatrix} 0.0025 & 0.0167 \end{bmatrix}$$

$$Tangentialdistortionerror \begin{bmatrix} 2.1093e-04 & 2.5468e-04 \end{bmatrix}$$

2.7.2 Extrinsic errors

They are available in the *estimationerrors.mat* file

2.8 Arguments for which camera model best fits

If the lens distortions are really too severe (for fish-eye lenses for example), the simple guiding tool based on a single distortion coefficient kc may not be sufficient to provide good enough initial guesses for the corner locations. then corner extraction has to be done manually.

3 Measurement of the accuracy and precision of a KUKA youBot arm

3.1 Description of the experiment

- The experiment involves a KUKA youBot arm performing different pick and place tasks.
- Three objects with different masses are to be placed in three different poses.
- The object poses (x, y,) are determined by tracking ArUco markers that are placed on top of the objects by an external vision system (Microsoft LifeCam) placed above the arm's workspace.
- The experiment is performed 20 times for each object-pose combination resulting in a total of 180 experimental trials.

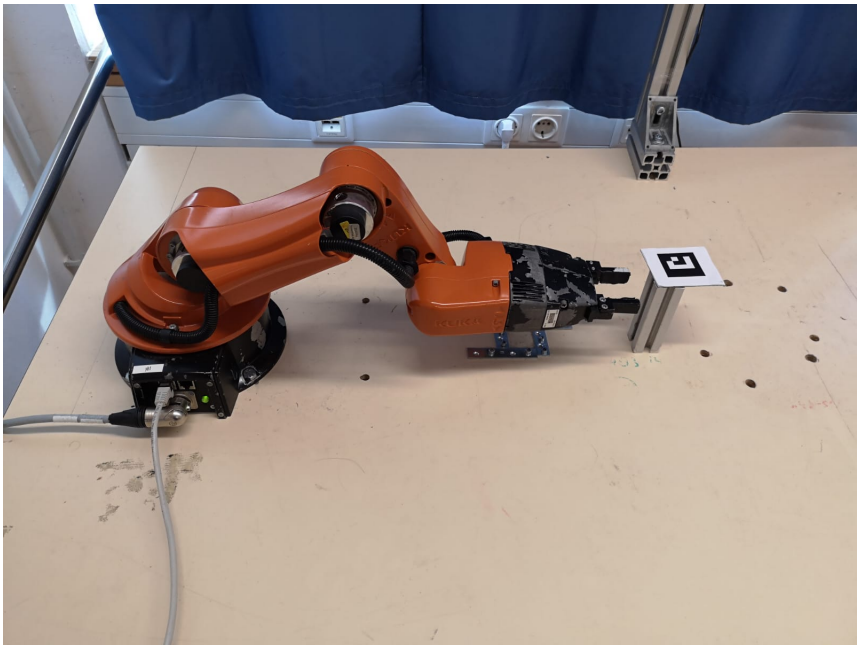

Figure 20: KUKA youBot arm after placing the object

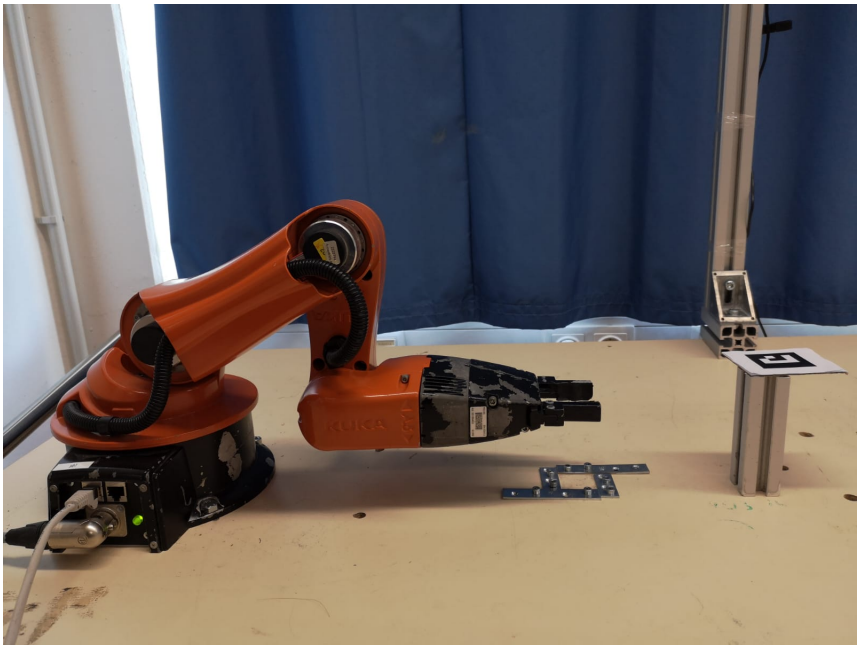

Figure 21: KUKA youBot arm at the initial position

3.2 Observations

3.2.1 Sources of error

- Changes in the position of the small/ medium object while being placed on the table.
- Changes in the position of the object while the arm retracts after placing the object.
- Small shakes in the table can cause changes in the pose readings.

3.2.2 Pose filtering procedure

- For the obtained pose values in each experiment we have fitted a Gaussian distribution.
- It can be seen that there are some outliers in the data.
- We have removed these outliers by using Chebyshev algorithm and the final data is plotted.
- On average, 3 outliers were detected for the entire experiment.

3.2.3 Distribution plots

- For small mass object

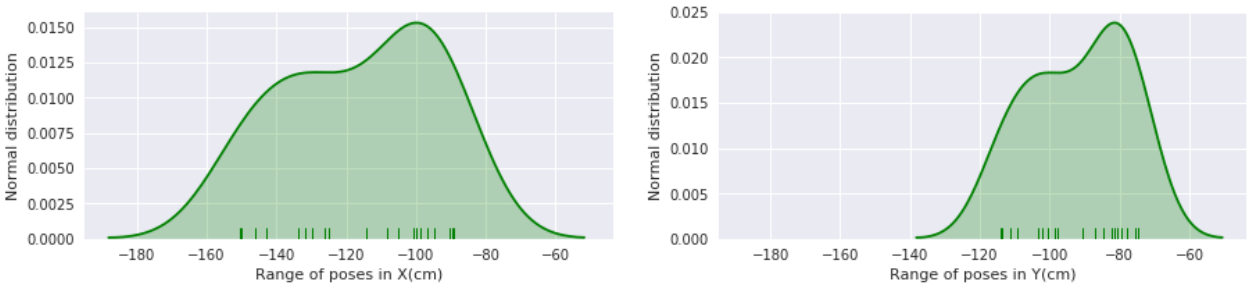

Figure 22: Plots before the removal of outliers for the poses of the straight motion with object of small mass

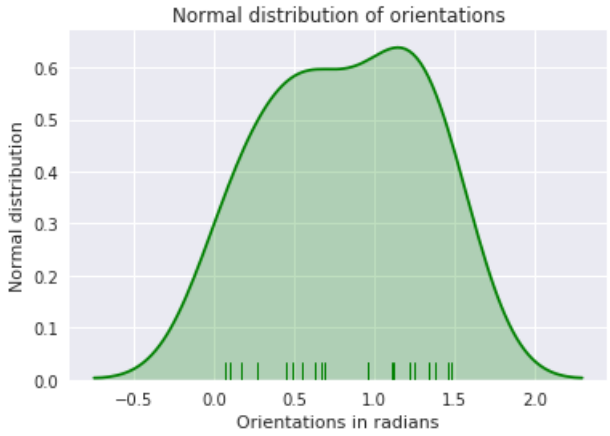

Figure 23: Plots before the removal of outliers for the orientation of the straight motion with object of small mass

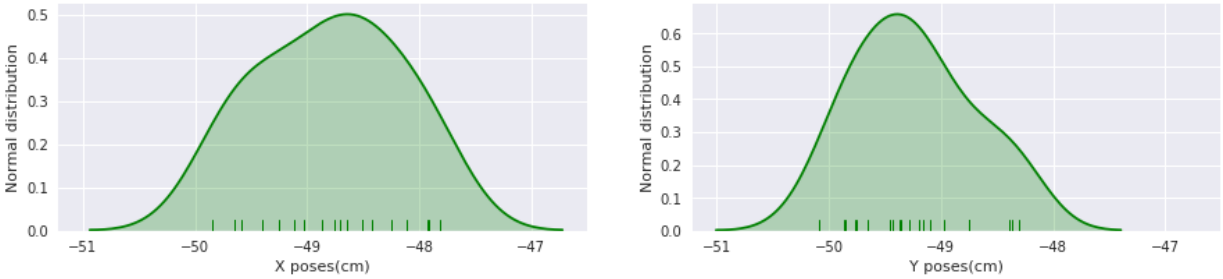

Figure 24: Plots before the removal of outliers for the poses of the left motion with object of small mass

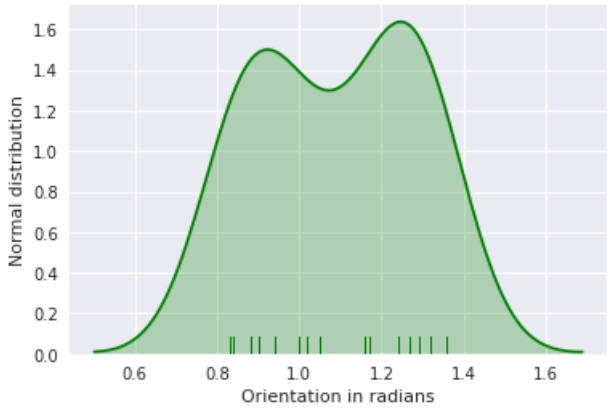

Figure 25: Normal distribution of orientation for placing marker left

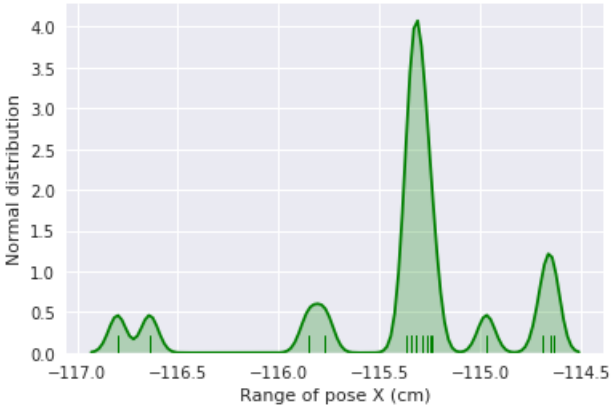

Figure 26: Plots before the removal of outliers for the poses of the right motion with object of small mass

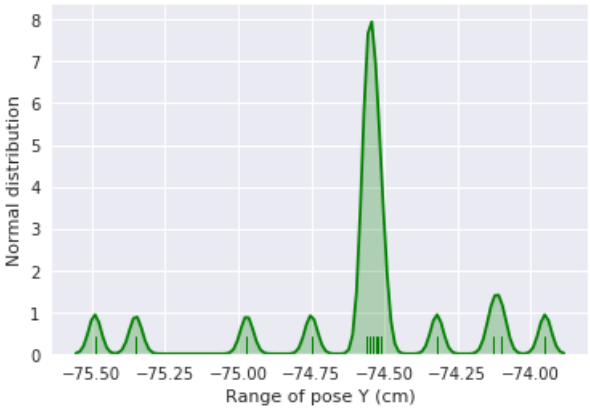

Figure 27: Plots before the removal of outliers for the poses of the right motion with object of small mass

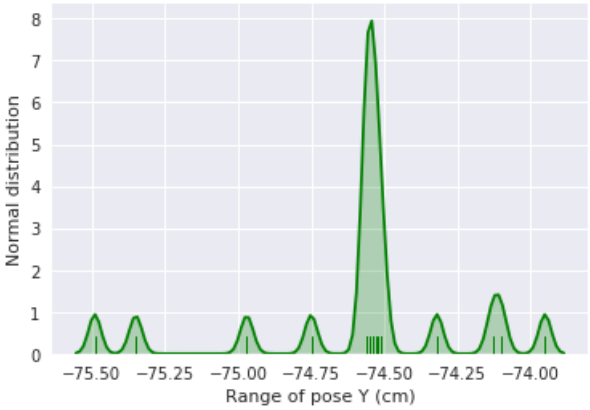

Figure 28: Plots before the removal of outliers for the poses of the right motion with object of small mass

- For the large mass object
  - For straight line motion

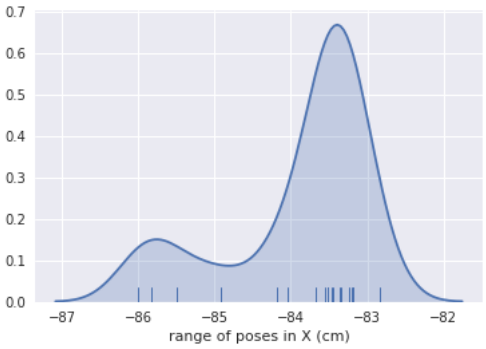

Figure 29: Distribution of X coordinates for straight line motion of large object

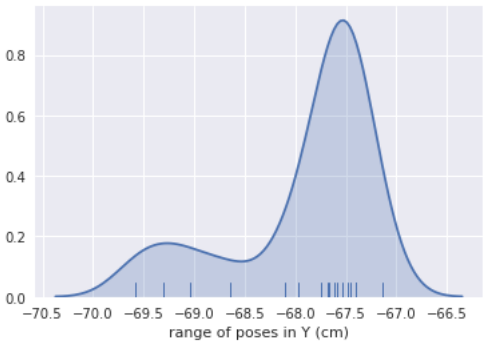

Figure 30: Distribution of Y coordinates for straight line motion of large object

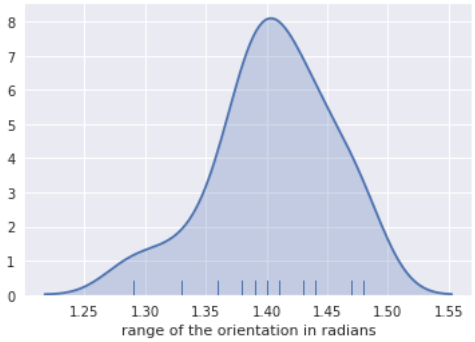

Figure 31: Distribution of orientations for straight line motion of large object

– For left arc motion

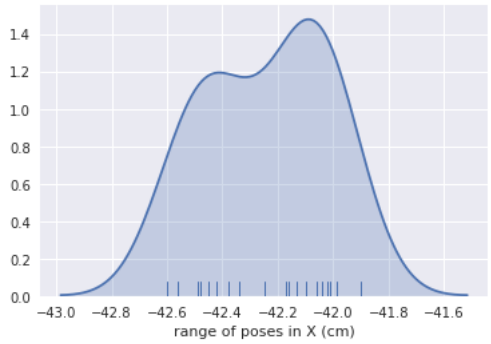

Figure 32: Distribution of orientations for left arc motion of large object

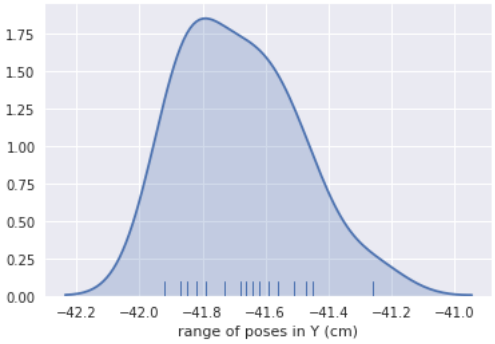

Figure 33: Distribution of orientations for left arc motion of large object

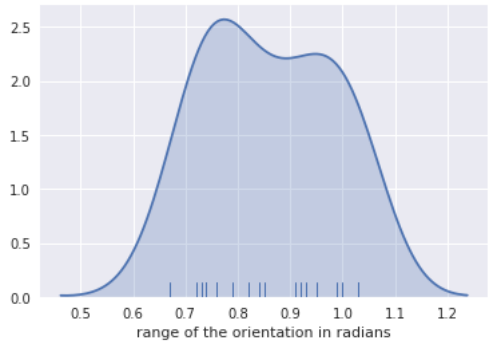

Figure 34: Distribution of orientations for left arc motion of large object

– For right arc motion

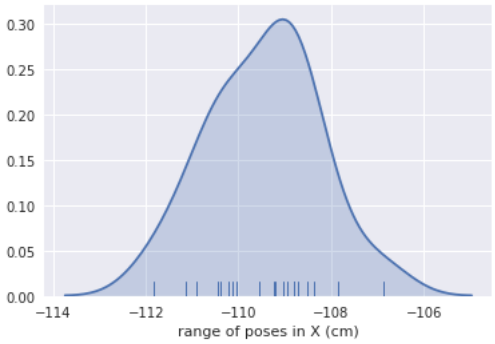

Figure 35: Distribution of orientations for right arc motion of large object

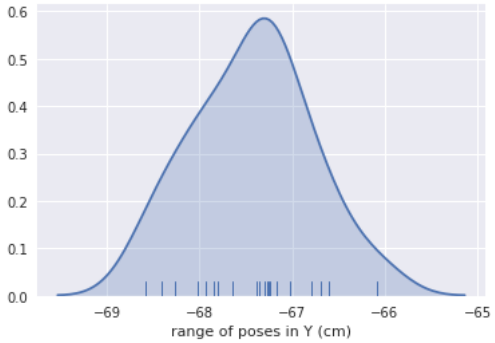

Figure 36: Distribution of orientations for right arc motion of large object

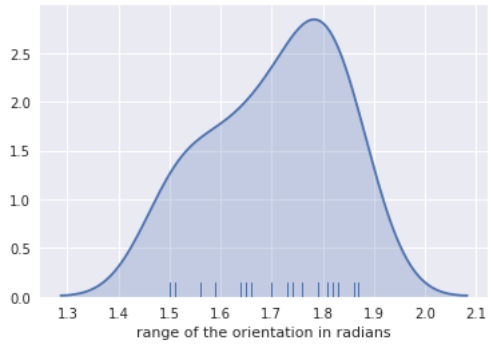

Figure 37: Distribution of orientations for right arc motion of large object

4 Appendix

4.1 Softwares used

- Python 2.7

- Jupyter Notebook

4.2 Packages used

- Matplotlib
- Numpy
- Scipy
- Seaborn
- Pylab

4.3 Tools used

- Onscreen ruler is used to mark scale on the sheet (metrics in inches)
- `https://www.piliapp.com/actual-size/inch-ruler/`

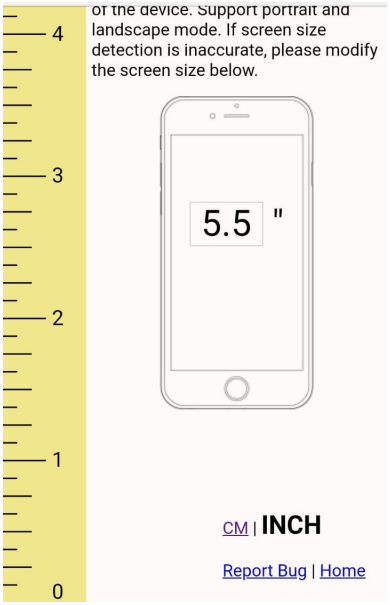

Figure 38: Online ruler

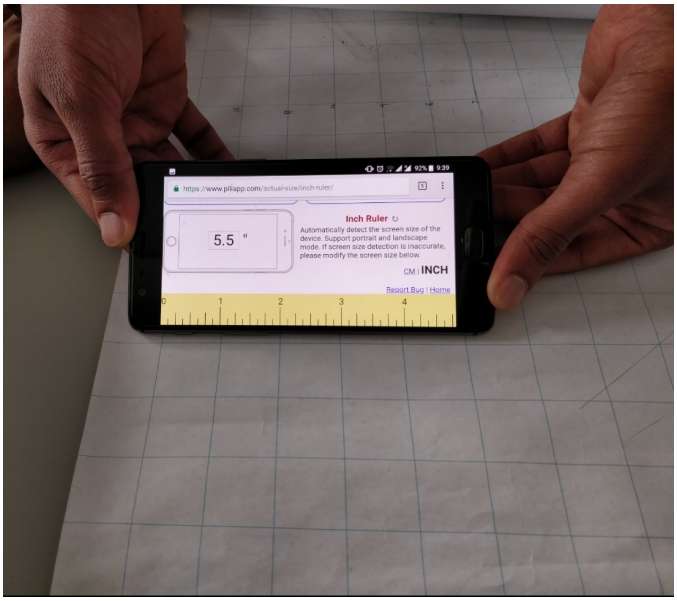

Figure 39: Marking scale on the sheet using onscreen ruler