

# Utilizing Object Detection of Humanoid Robots for Implementation in Color Recognition and Motion Planning



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## Abstract

Algorithms and operating systems are continually being evolved in need to develop more equipped and reliable robotic systems capable of performing in many real-world dynamics. The NAO robot is used to observe the ability of robots to perform specific tasks and the accuracy and reliability in the robot's performance. The robot is tasked with color detection, accurately identifying object distances and motion planning. Algorithms modeled in Python with use of OpenCV allowed for successful object registration.

## Petri nets

**Definition:** A graphical tool used for modeling and simulating various complex systems. Petri nets are a 5-tuple:  $PN = (P, T, F, W, Mo)$  where  $P =$  Places,  $T =$  Transitions,  $F =$  Arcs,  $W =$  Arc Weight, and  $Mo =$  Initial Marking Tokens. These variables represent the states, operations, and connections within systems, where a tokens movement throughout the net represents the current state of the system.

**Firing Rules:** A token may only fire once a transition is enabled, meaning the correct amount of tokens must be present in the input place and the arc weight conditions must be satisfied. The firing of a token moves the token from input place through the transition, then finally to the output place.



## Object Detection and Motion Planning

Using OpenCV and Choregraphe applications, the NAO Robot detects the color of each puzzle piece for use in the grasping and path motion behavior process.

An algorithm allows for the creation of trackbars for detection of hue, saturation, and value, HSV, of each corresponding color for further implementation with the NAO. In other words, trackbars allow for the masking of unwanted colors, thus allowing for the accurately measured HSV values seen in Table 1.

Table 1. Minimum and Maximum HSV Values

|                     | Hue Minimum | Hue Maximum | Saturation Minimum | Saturation Maximum | Value Minimum | Value Maximum |
|---------------------|-------------|-------------|--------------------|--------------------|---------------|---------------|
| Navy Blue Diamond   | 105         | 165         | 185                | 255                | 100           | 255           |
| Red Oval            | 0           | 7           | 195                | 253                | 138           | 255           |
| Purple Rectangle    | 113         | 150         | 101                | 255                | 134           | 255           |
| Forest Green Circle | 48          | 98          | 156                | 210                | 80            | 164           |
| Orange Square       | 8           | 28          | 154                | 209                | 195           | 255           |
| Periwinkle Hexagon  | 67          | 110         | 89                 | 255                | 190           | 255           |
| Pink Pentagon       | 130         | 173         | 64                 | 255                | 207           | 255           |
| Lime Green Triangle | 35          | 60          | 115                | 235                | 170           | 255           |
| Yellow Star         | 23          | 45          | 90                 | 245                | 164           | 255           |

Masks are used simultaneously with the bitwise\_and operator to create the resulting image. This operation adds two images together to detect the color desired by filtering out all the unwanted pixels, as shown below in Figure 1.



Figure 1. OpenCV Masked and Resulting Image

For motion planning, the Choregraphe timeline application stores the arm movements. NAO will grasp the puzzle pieces positioned at the marked location on the table by the user. The main goal is to store many small keyframes to return the puzzle piece to the correct location in one smooth movement, seen in Figure 2.

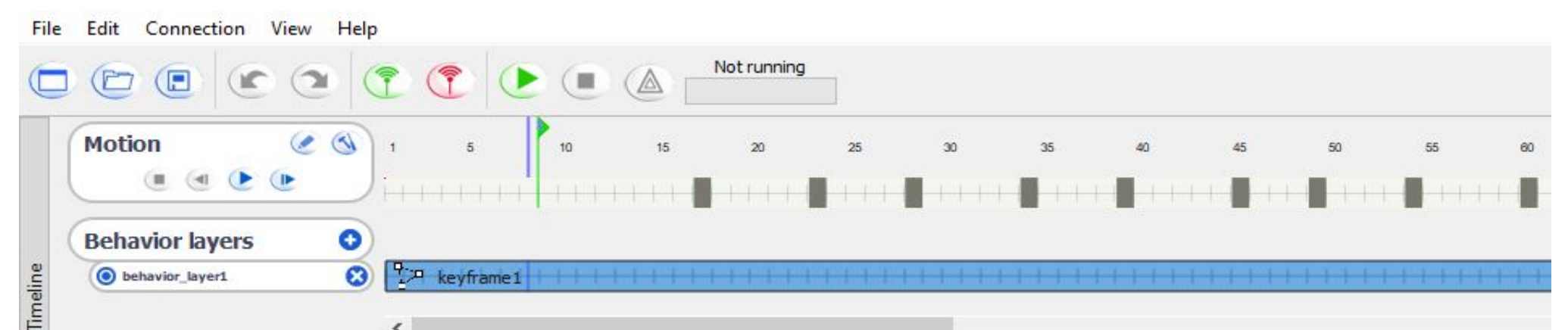


Figure 2. Choregraphe Timeline

## Results

The Color Detection algorithm created detects all nine colors represented in the puzzle board with the assistance of trackbars. The grasping and motion planning procedure using the NAO's left hand was achieved for the circle, square, pentagon, and triangle for a total of 220 completed trials. Storing various keyframes allowed NAO to correctly identify the predetermined location of the puzzle piece to be grasped to then fulfill the final motions of returning the puzzle piece to its respective location.

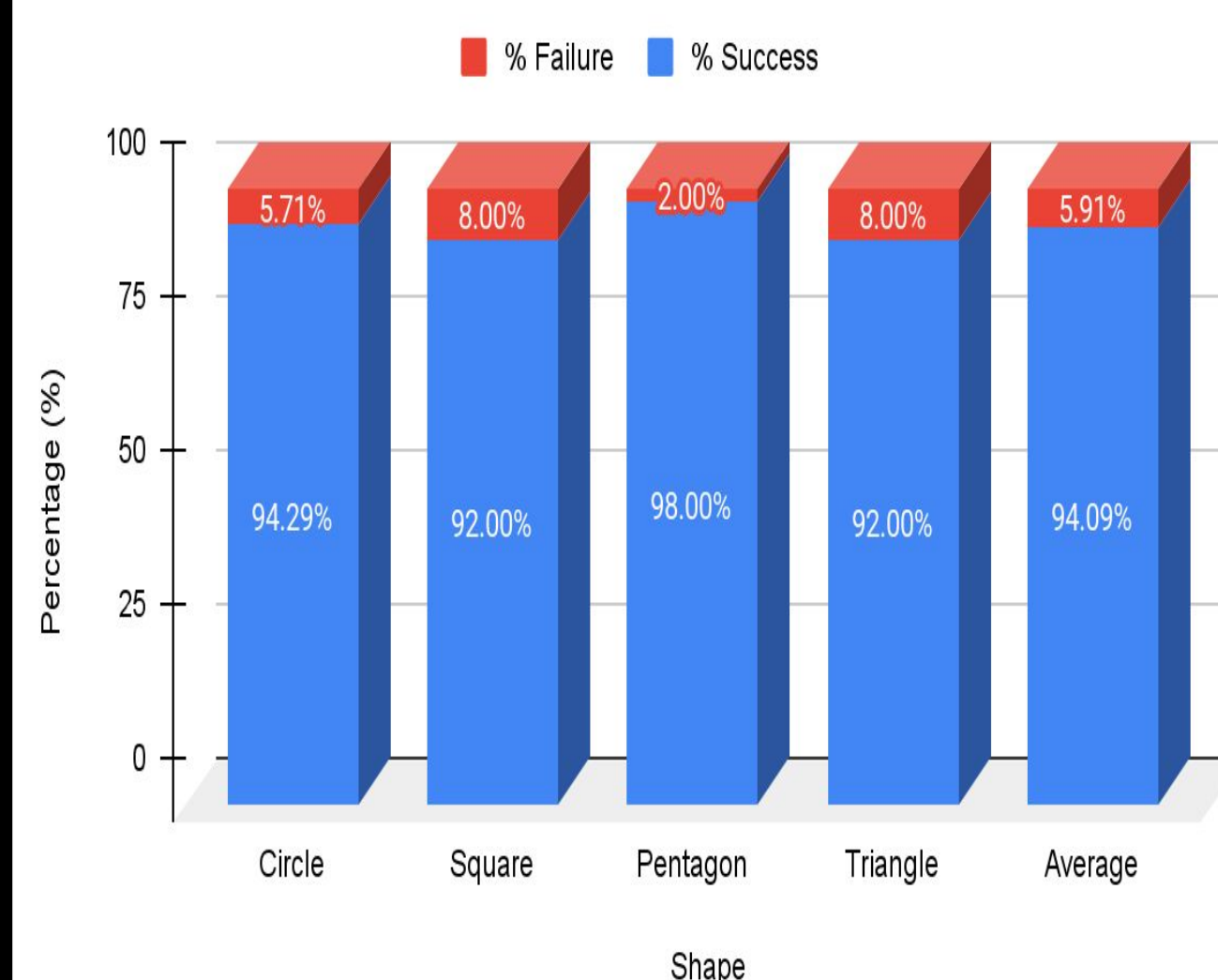


Figure 3: Success and Failure Rate

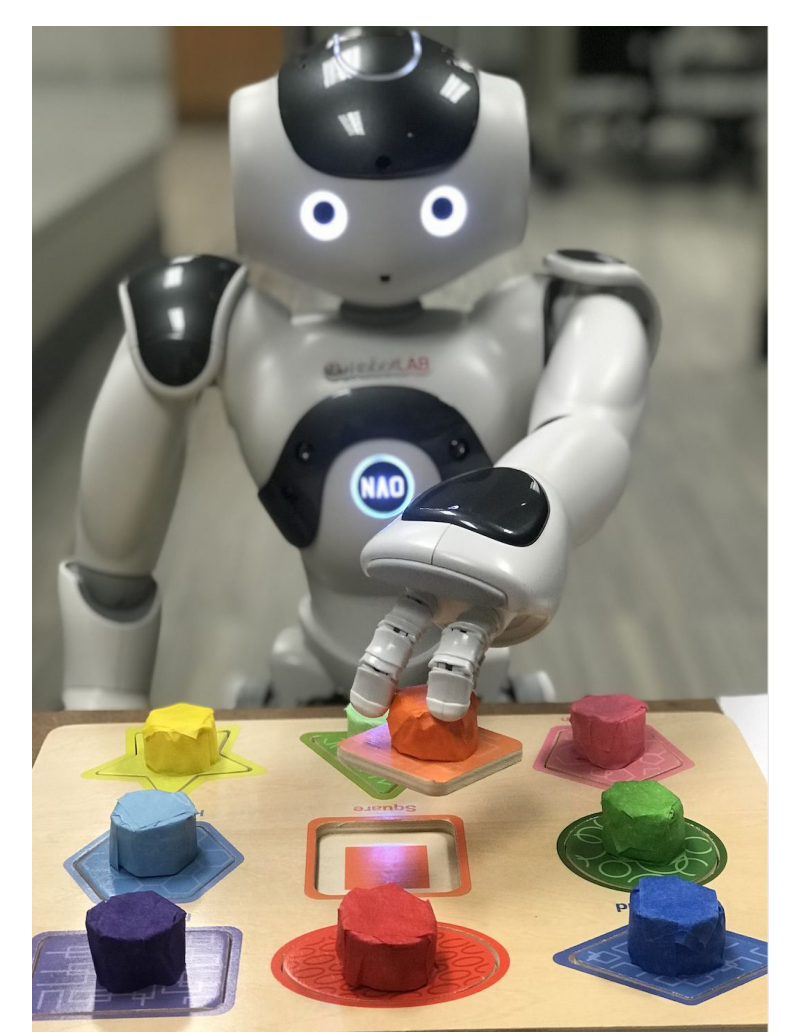


Figure 4. NAO Motion Planning

## Conclusion & Future Work

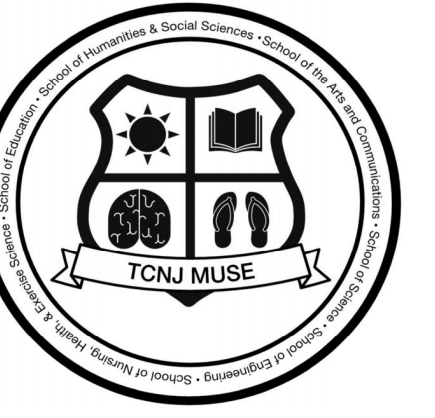
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## Acknowledgement

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# Utilizing Object Detection of Humanoid Robots for Implementation in Motion Planning and Stereo Camera Calibrations

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## Fuzzy Logic

**Definition:** Uses fuzzy set theory to describe, analyze, and process qualitative or imprecise data.

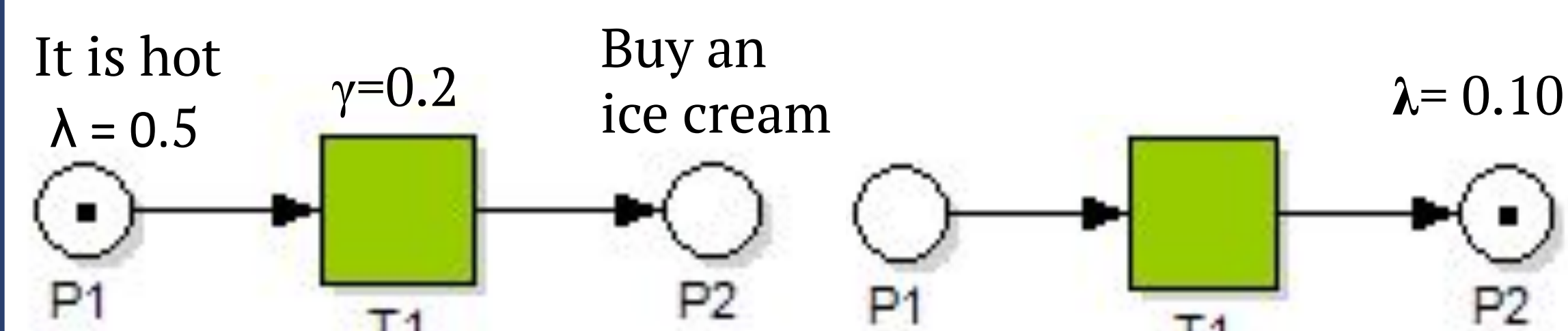
**Fuzzy Set:** A set that defines the truthfulness of a particular set. With the truth value ranging from 0 to 1, where 0 and 1 are completely false and true, respectively. Decimal values between 0 and 1 allow for partial membership indicating degree of truth.

$$\mu_x : X \rightarrow [0, 1]$$

## Fuzzy Petri nets

**Definition:** A subset of Petri nets that incorporates fuzzy logic. Decisions are made based on the truthfulness value and fuzzy logic allows for conclusions on qualitative situations lacking precise reasoning.

**Firing Rules:** Tokens are assigned a certainty factor ( $\lambda$ ) and transitions a threshold value ( $\gamma$ ). The firing of a token is enabled when  $\lambda > \gamma$ . The resulting truth value is  $\sigma = \lambda \times \gamma$ .



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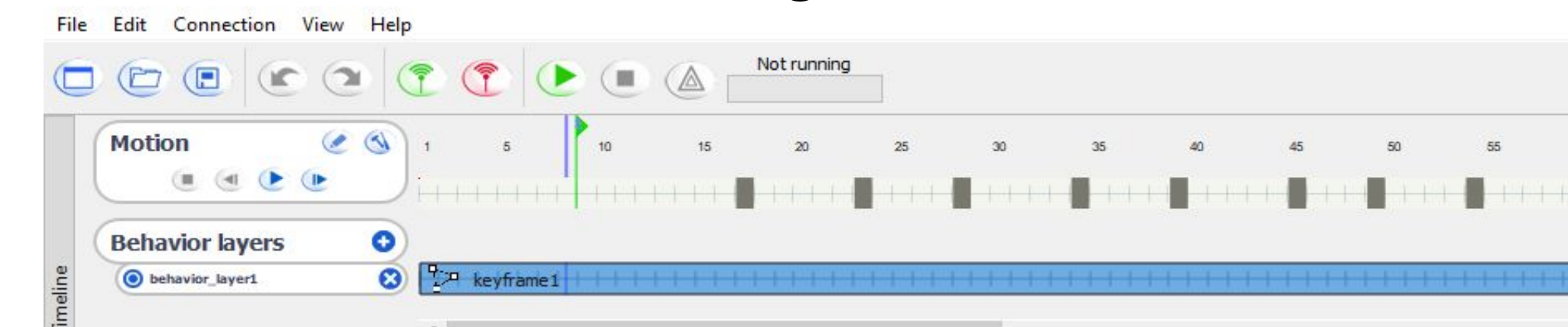


Figure 2. Choregraphe Timeline

## Object Detection and Stereo Camera Calibrations

Two web cameras with a baseline ( $b$ ) as shown in Figure 1 are used to create a stereo camera which takes multiple photos from varying angles of a checkerboard print-out. After running these images through a stereo camera calibration app, a camera calibration matrix ( $K$ ) as shown in (1) is produced, providing the camera's focal length ( $f$ ) in pixels.

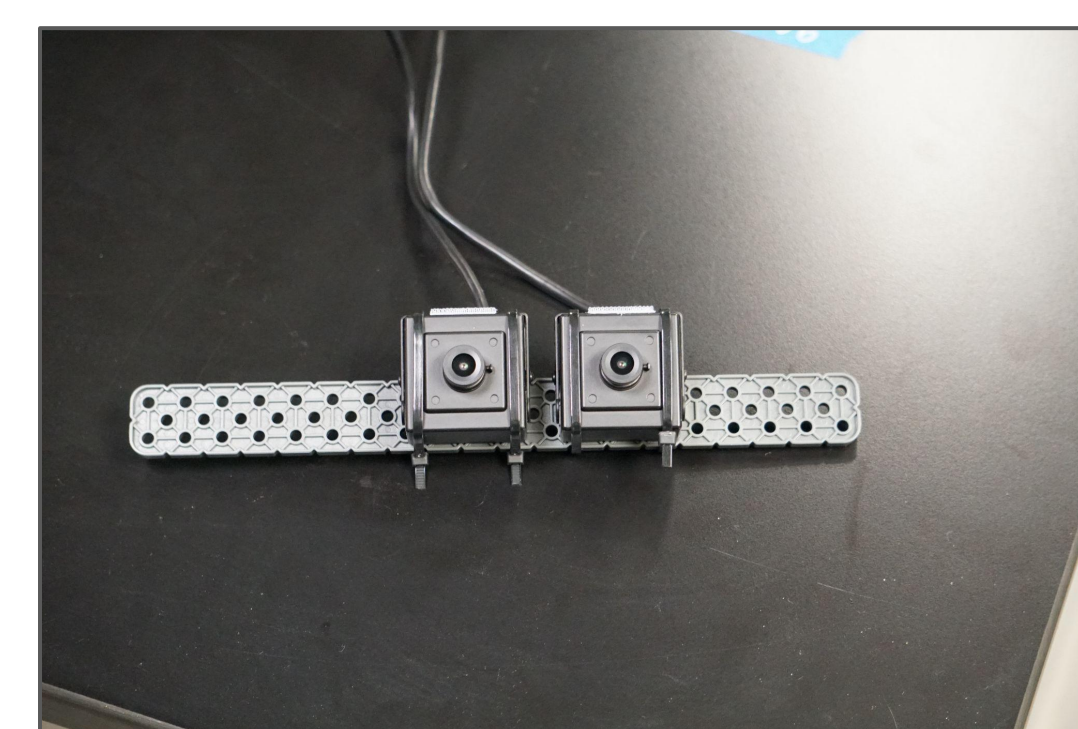


Figure 1. 55 mm baseline distanced cameras

$$K = \begin{bmatrix} f_x & 0 & c_x \\ 0 & f_y & c_y \\ 0 & 0 & 1 \end{bmatrix} \quad (1)$$

The cameras then take photos of a red ball of diameter 100 mm at calculated distances ( $d$ ). The images are passed through the created Reduction Filter method, which involves converting the RGB image to HSV to eliminate colors out of the red range from the image.

With only the red ball remaining in the image, a Hough Circle Transform circle is created, outlining the ball and locating its center point. Finding the difference between the ball's center point location on corresponding frames from the two cameras as shown in Figure 2 obtains the measurement of disparity in pixels ( $uL - uR$ ).

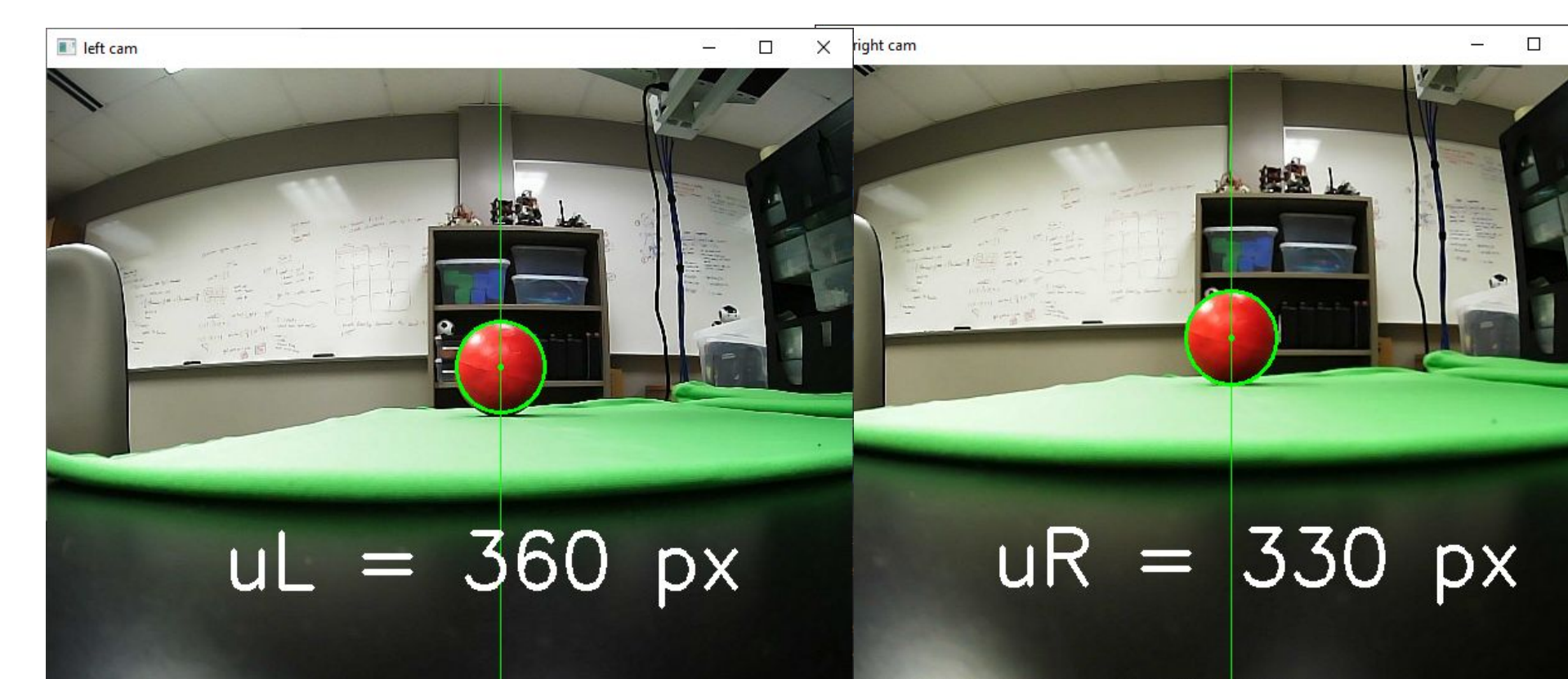


Figure 2. center points of red ball in both frames

Finally using the Stereo Camera Distance Ranging Method, the distance from the cameras to the red ball is calculated in (2).

$$d = \text{distance from camera to object} = \frac{\text{baseline (mm)} \cdot \text{focal length (pixels)}}{\text{disparity (pixels)}} = \frac{bf}{uL - uR} \quad (2)$$

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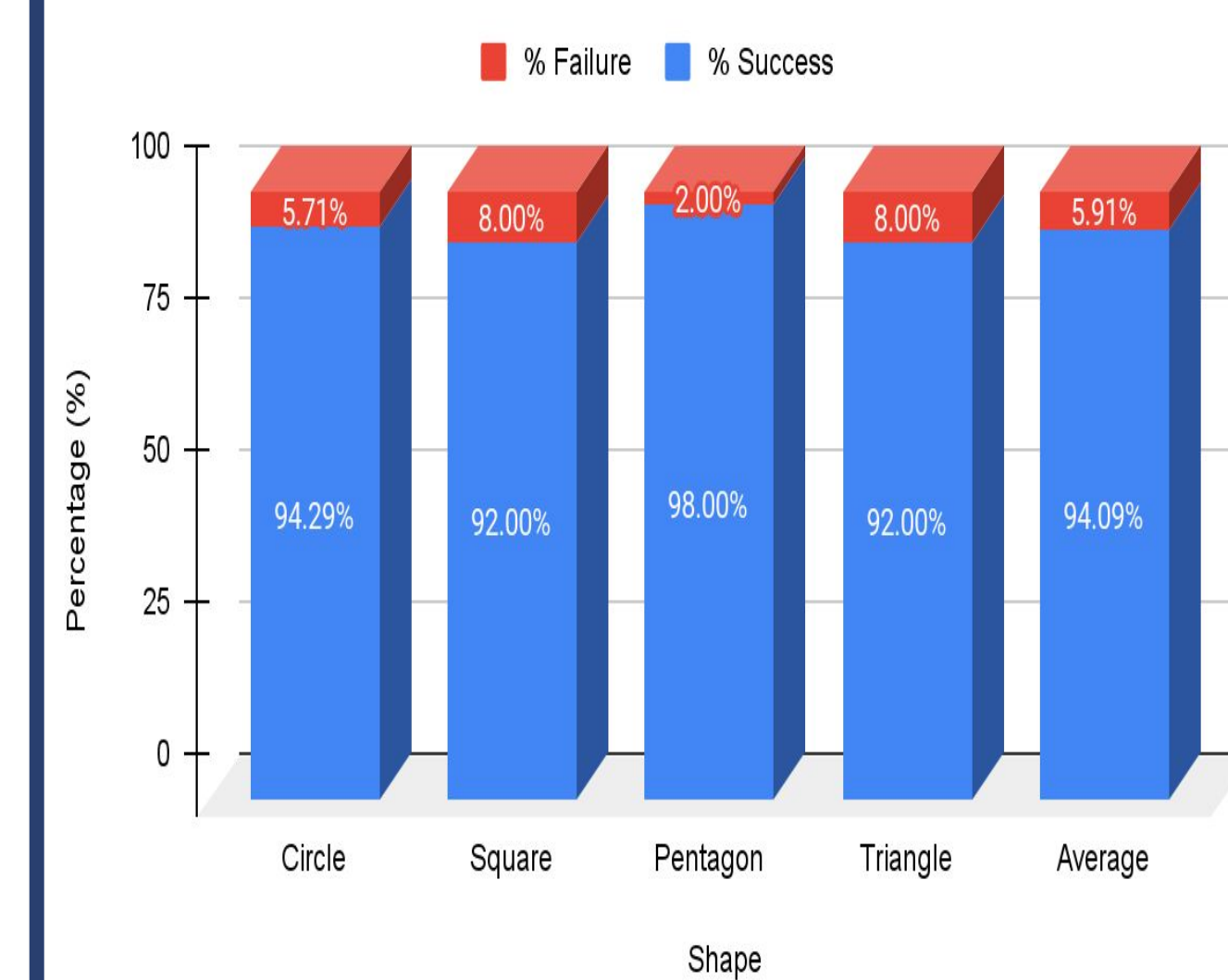


Figure 3: Success and Failure Rate



Figure 4. NAO Motion Planning

The Optimized Stereo Vision Distance Detection algorithm was finalized after multiple trials resulted in precision and accuracy. Creating the Reduction Filter method in collaboration with accurate camera calibration matrix data allowed for reliable results as shown in Figure 3. This algorithm can be implemented for the future of humanoid robots, by utilizing two cameras in the same location of their eyes to become more human-like.

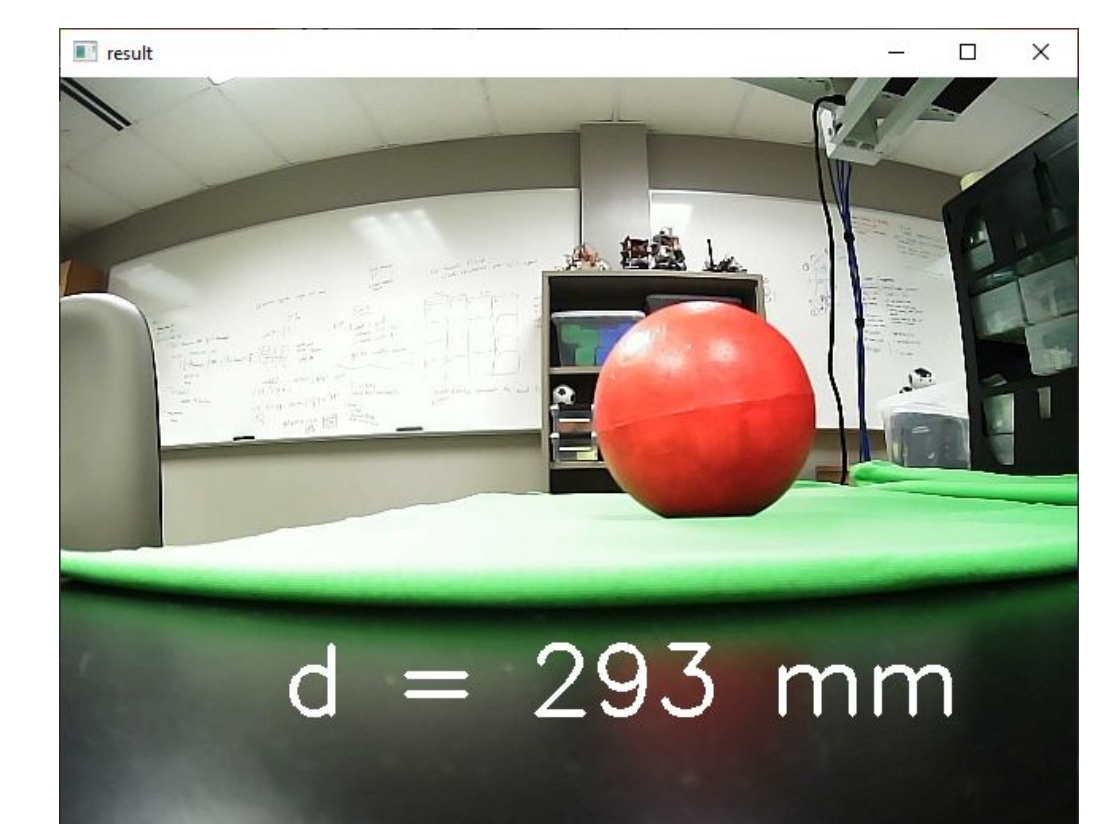


Figure 3. 300 mm distance from cameras

## Conclusion & Future Work

To discover an appropriate way of sending a ping from Python to Choregraphe for a connection between the color recognition and motion planning aspects. Also, to further improve the NAO's ability to locate objects. To improve the accuracy of longer ranged distances with the created stereo camera distance ranging algorithm and apply these methods to the NAO robot.

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