

Penalty Shootouts by Aldebran Nao

Lovish(Y9227309) and Rahul(10560)

Guide: Prof Amitabh Mukerjee

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Introduction

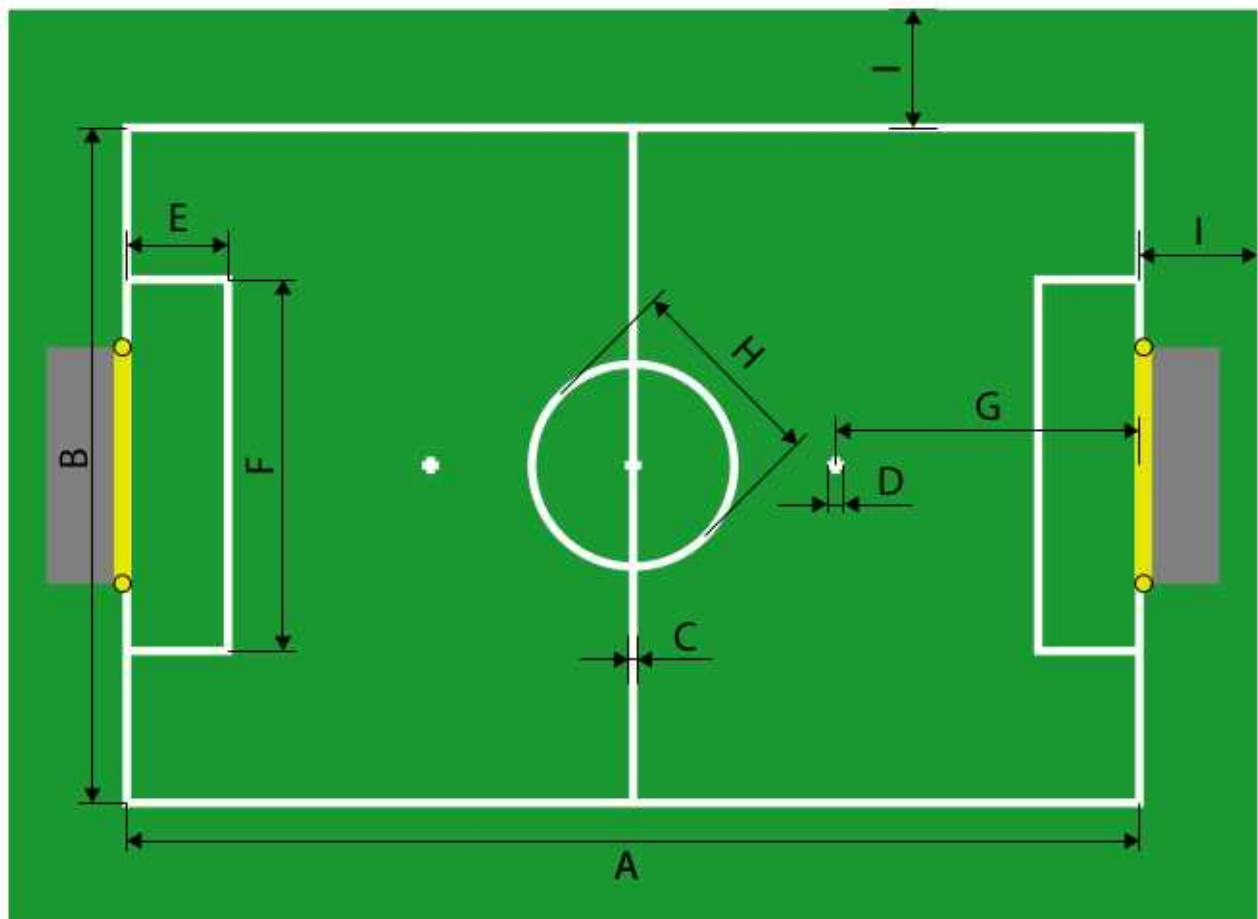


Figure 1: Field of RoboSoccer

Synonymous to human body, humanoid robots have carved a niche into the robotics world due to the fact that they can closely imitate humans in every task be it cleaning home, playing football or dancing. Rapid research in the area of robotics ie better hardware etc have given rise to some exciting competitions such as [RoboCup](#) where a team of four robots compete against other in a football match. The aim of the tournament is to defeat the best human soccer team by the year 2050.

Nao, from Aldebran Robotics is the robot used in the Robocup Standard Platform League. One of the task that it has to accomplish is penalty shootouts in case of tie. The distance at which robot is standing to goal is given by G in figure 1. G is kept constant, like in the real soccer match, at 180 cm. The goal dimensions are given in Figure 2.



Figure 2: Dimensions of Goal (in mm)

The goal width is 150 cm as you can see in figure 2. For our implementation goal of width 30cm is kept at a distance of 120cm.

Previous Work

As RoboCup takes place every year from 1998(2008 with Nao) there is plethora of work going on in the direction of increasing the probability for a positive penalty. In [1] authors have used Reinforcement Learning to increase this to 85%.

Apart from it stability becomes a major player when balancing on one leg. The stability is classified into the types:-

- **Static** - In static kicking there is no continuous feedback to the body from camera. Once the decision is taken it is not reconsidered. [2] and [1] discuss keyframe based approach using Bezier Curves which we have also followed.
- **Dynamic** - Dynamic kicking involves taking PID (proportional integro differential) feedback controllers or using the Nao's Gyroscopic feedback to decide for itself whether the next keyframe arriving after the present one is stable or not. The papers describing this are [3] and [4].

Approach

Our method can be decomposed into the following steps :-

- **Goal Detection:** For goal detection the RGB image derived from Nao's forehead camera is used. It is first converted to HSV (Hue Saturation Value) space and then morphological image processing operations (dilation followed by erosion) are performed on it. Finally the center of goal is found from it.
- **Adjust Angle:** In figure 5, we can see how Nao changes its body orientation towards the goal. We have curated a very simple model (following the KISS- Keep it Simple Stupid principle). It assumes that the ratio of pixels from a point to other in an image is roughly equal to the ratio of their distances in the same image i.e.

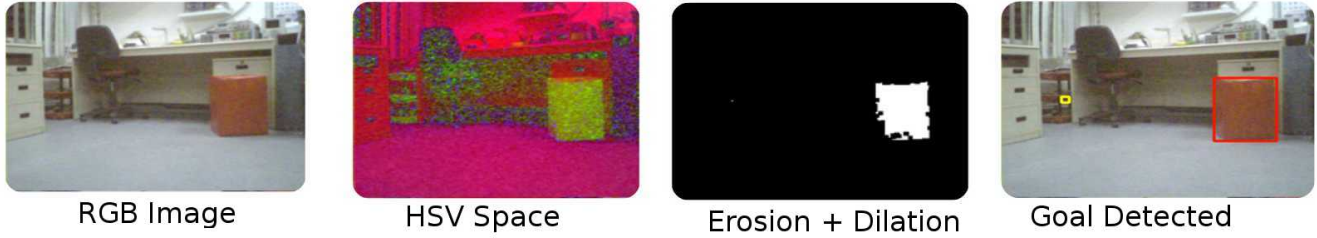


Figure 3: Steps for detection of Goal

$$P_{AB}/P_{AC} = D_{AB}/D_{AC}$$

D_{AC} is found out by using the field of view angle of upper camera. As D is fixed at 120 cm and the following diagram depicts the top view of Nao,

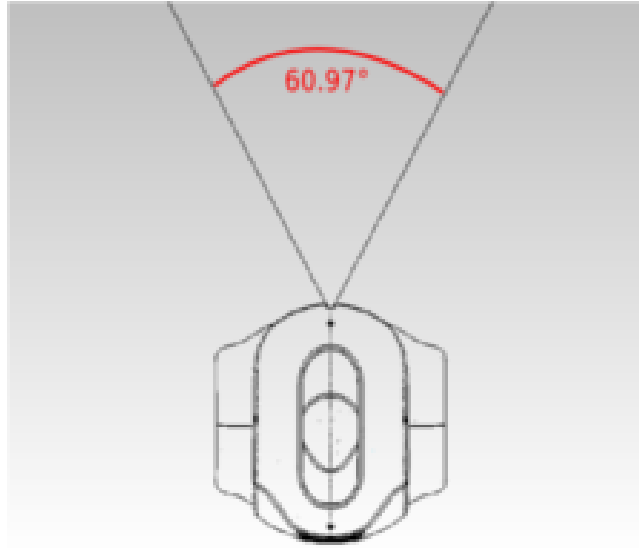


Figure 4: Top view of Nao showing the Field of view angle

Finally as the distance to goal is fixed ie in our case equal to 120 cm. We can find θ simply by

$$\tan \theta = D_{AB}/120$$

- **Ball Detection:** After the angle is adjusted Nao's HeadPitch angle is changed from 0 to 29.5 degrees so that it can look down completely. The ball is then detected in the same way as goal, the difference is only that the image is taken from Nao's Lower Camera. Figure 5 describes the same.
- **Adjust Position:** Looking at the position of the ball, Nao adjusts its position by moving front, left or right. In figure 7, two cases are shown, 7(a) is an image upon which nao decides to walk in front and in 7(b) nao needs to move right. The number of steps taken are a function of the abscissa and ordinate of ball in image.
- **Kick:** We have adopted a bezier curve interpolation between keyframes for kicking the ball. Below in figure 8 you can see the 3 keyframes (can be understood colloquially as "screenshots"). Each of the three keyframes consists of different HipPitch, KneePitch and AnklePitch angles. Here instead of using a motion engine as in [1] we have manually found out these frames by checking stability of Nao in various positions.

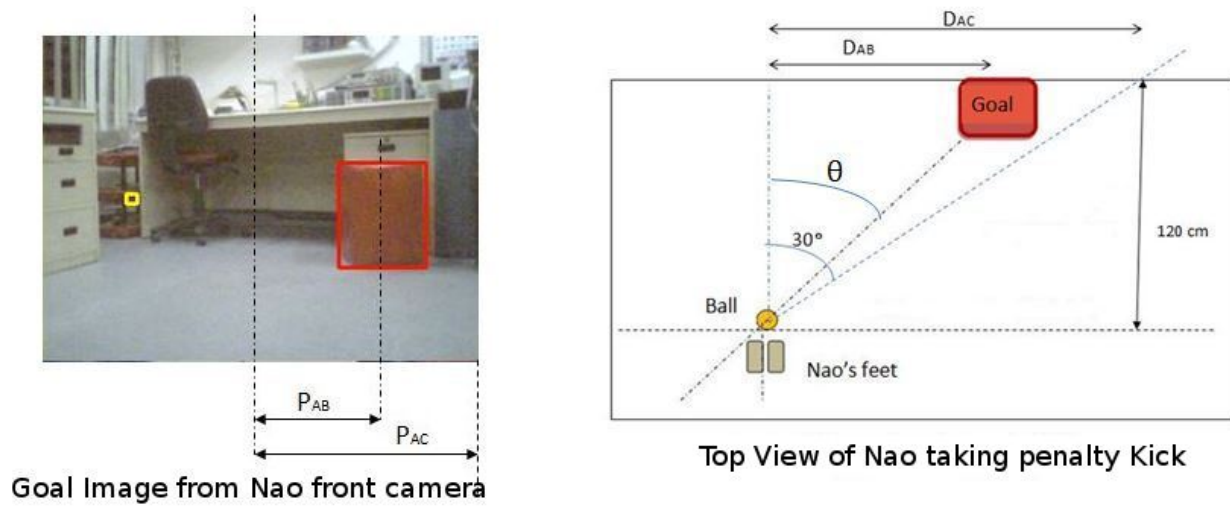


Figure 5: Calculation of Angle to be turned

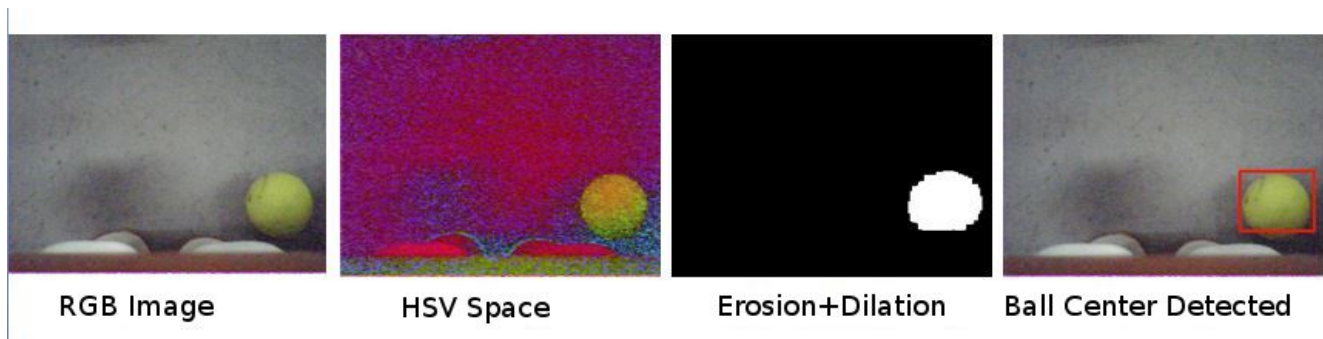


Figure 6: Steps for detection of ball

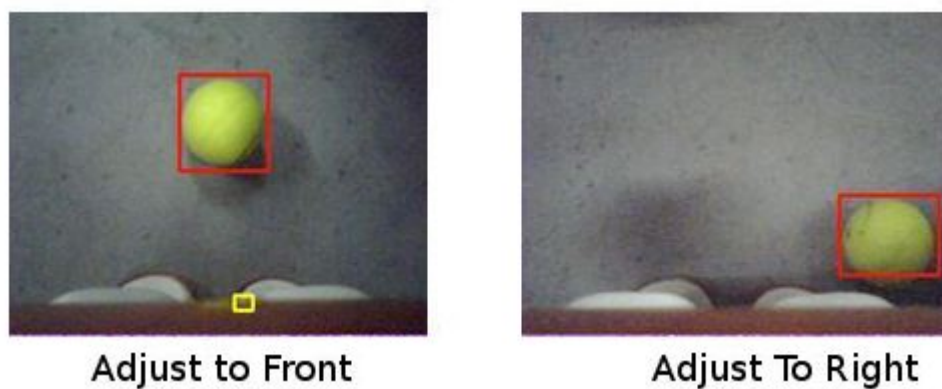


Figure 7: Scenarios for Adjusting Position

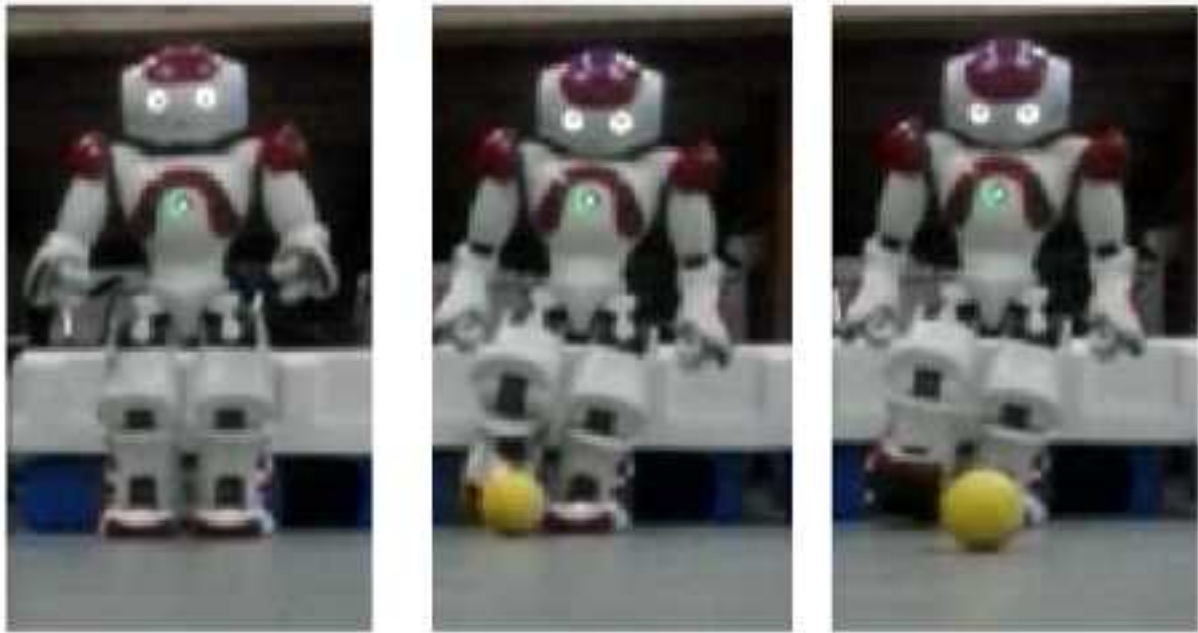


Figure 8: Keyframes while Kicking the ball

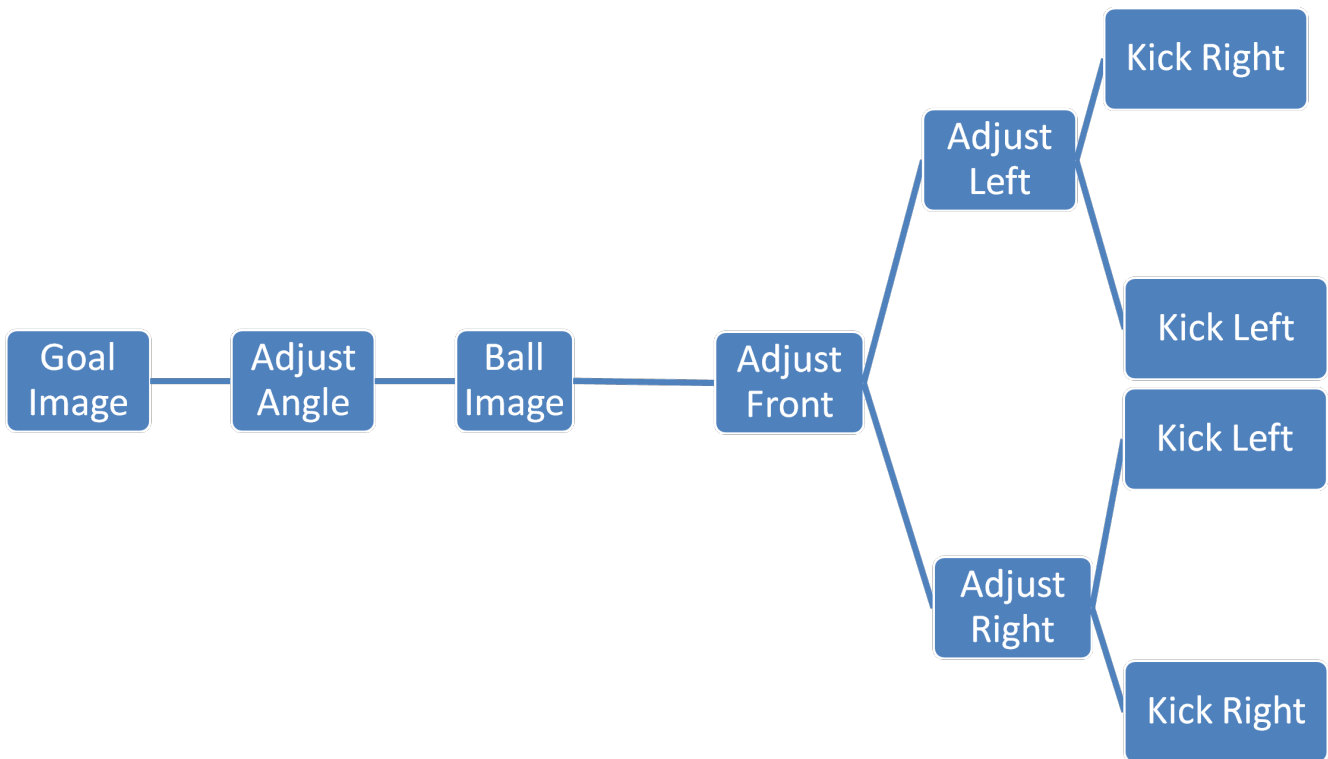


Figure 9: Wireframe of Approach

Results

The Goal Hit Percentage through our implementation is coming out to be 55-60%. Video of some attempts (both hits and misses) can be found at this (<http://home.iitk.ac.in/~lovishc/cs365/project/videos>)[link].

Problems Faced

Some of major problems we encountered in the project are as follows :-

- After few practice kicks, Nao's joints get heated up due to flow of current which makes the robot stiff. Hence it loses balance after 7-8 kicks as less current passes through these joints to prevent damage.
- Smooth floor is the biggest concern as the ball just spins around grooves, subtle inclinations and declinations. Due to them, it sometimes misses the goal. Also the ball we are using is a soft rubber one, hence it moves only by 150 cm no matter how hard Nao kicks it..

Future Work

As of now we have implemented only static kicking ie if the robot once take decision to kick the ball , it will then irrationally kick it even if the ball is displaced from the position when the decision was taken. Dynamic kicking involves taking the feedback from Gyroscope and balancing the robot based on that. It could be a path for future work on it.

Also right now, we have only one kind of kick ie no side kick etc. We can possibly find keyframes for maybe , a few more kick types and apply reinforcement learning [1] which will make the decision process rational and improve the results from 60% hit rate to 85% as stated in [1].

References

- [1] Hester, Todd, Michael Quinlan, and Peter Stone. "Generalized model learning for reinforcement learning on a humanoid robot." In Robotics and Automation (ICRA), 2010 IEEE International Conference on, pp. 2369-2374. IEEE, 2010.
- [2] Czarnetzki, Stefan, Sören Kerner, and Daniel Klagges. "Combining key frame based motion design with controlled movement execution." In RoboCup 2009: Robot Soccer World Cup XIII, pp. 58-68. Springer Berlin Heidelberg, 2010.
- [3] Xu, Yuan, and Heinrich Mellmann. "Adaptive motion control: Dynamic kick for a humanoid robot." In KI 2010: Advances in Artificial Intelligence, pp. 392-399. Springer Berlin Heidelberg, 2010.
- [4] Müller, Judith, Tim Laue, and Thomas Röfer. "Kicking a ball—modeling complex dynamic motions for humanoid robots." In RoboCup 2010: Robot Soccer World Cup XIV, pp. 109-120. Springer Berlin Heidelberg, 2011.
- [5] NAOqi API guide ,[Link](#)
- [6] Master's Thesis of Tomàs González Sánchez, Department of Computer Science and Mathematics, Universitat Rovira I Virgili, September 2009, 64-82

Figure References

Figure 1 and 2 are adopted from RoboCup 2013 official rulebook available at this [link](#). Figure 4 is taken from Nao's documentation online([link](https://developer.aldebaran-robotics.com/doc/1-14/))[<https://developer.aldebaran-robotics.com/doc/1-14/>].

Acknowledgements

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