

#### OBJECTIVE

- The offside rule in Football is notorious for being mishandled. Error of judgement, however, is a **human limitation** and beyond control. Our aim was to create a mechanism by which the **problem of incorrect offside decisions** made by referees in football can be addressed

#### INTRODUCTION

- Our project takes in a video feed, **analyses each pass** and takes a call on whether the situation was an offside
- The implementation involves two separate modules that track the ball and the players. The **integration of the modules** leads to the desired goal of offside detection
- The ball in play and the players are tracked using **separate algorithms** found to be most suitable for their respective purpose
- The project is implemented in **Python** and uses the **OpenCV library**
- This choice reflects the **intuitive nature** of the Image processing library when implemented in Python
- The project still has certain shortcomings which are highlighted in the section titled **Future Scope**
- Be sure to stick around for a **live presentation!** You can read through this less-lively poster in the meantime!

#### CHALLENGES

- Since our project was software based, we were fortunate in being able to forego the road-blocks that are inherent in procuring suitable components, nevertheless each step had its own complications
- Numerous approaches for ball detection were tested, namely **Colour Thresholding, Histogram Back-projection, Hough circle transform and dlib library's correlation tracker**
- None of these approaches were perfect. After much deliberation we **finalised on Colour Thresholding**
- Colour Thresholding generates significant **noise** if the colour of the ball and background are similar, this was solved using **background subtraction**
- We discovered that dlib library's correlation tracker is unsuitable for tracking more than 3 players in real time
- Our final choice for player tracking was **Histogram Back-projection**
- Another candidate was the **HOG descriptor with SVM** which is a method of detecting persons using machine learning but this turned out to be disappointing, both in terms of speed and accuracy

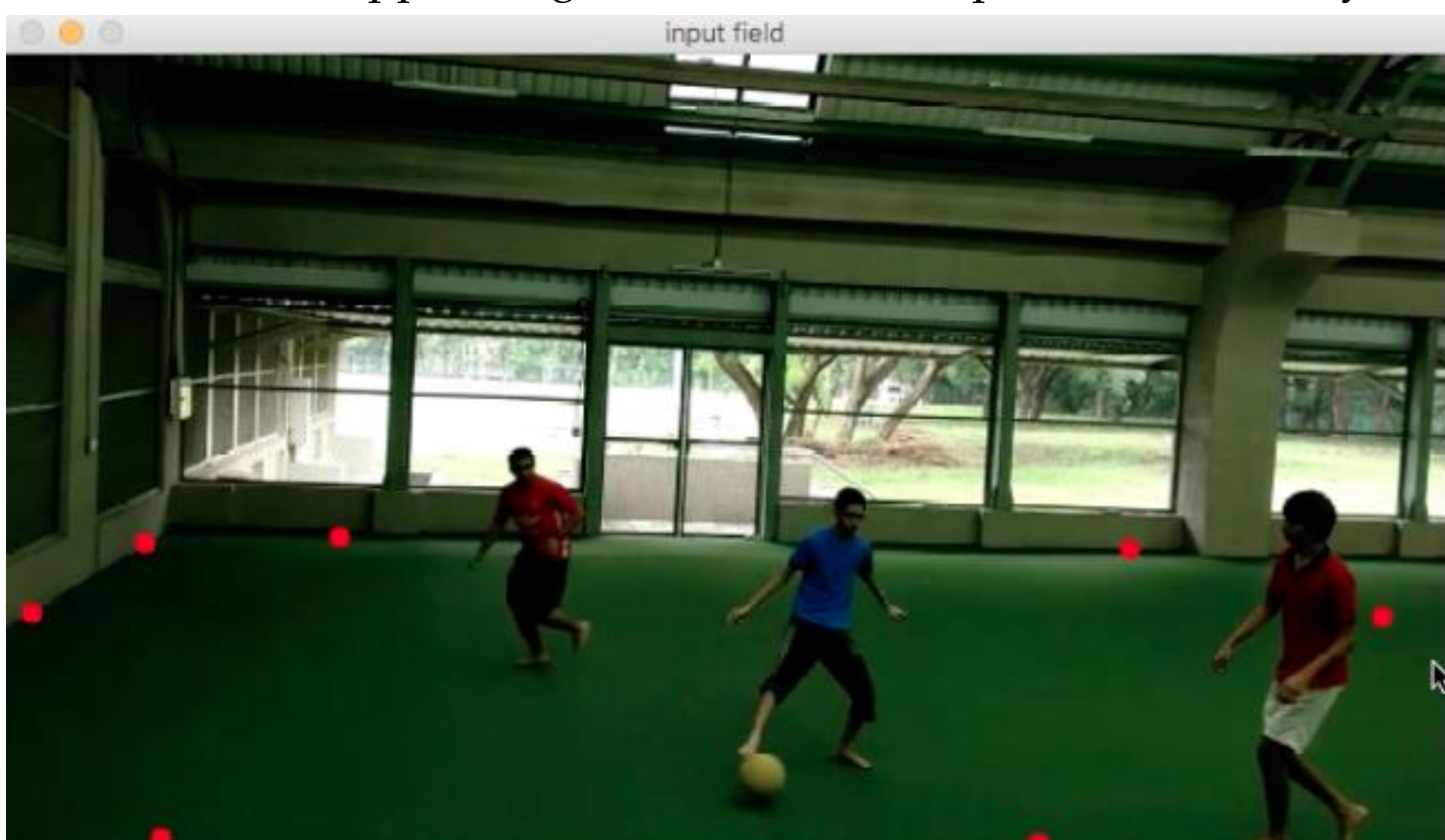
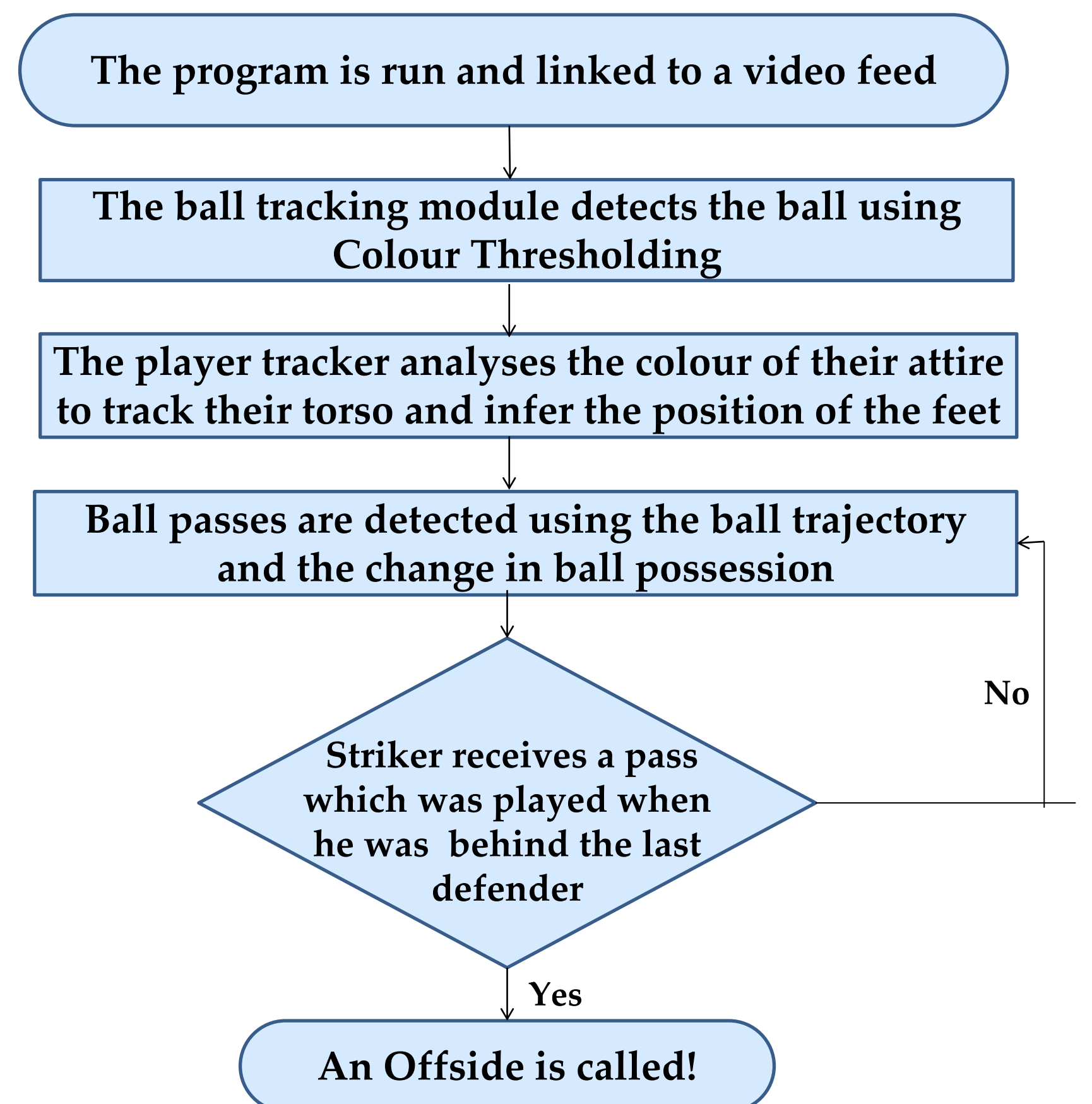


Figure 1: Demarcating the playing area, a one time activity



Figure 2: Image Processing in action! Seen here are the tracked elements with the pesky background subtracted

#### WORKING MECHANISM



#### FUTURE SCOPE

- Some of the ideas for improving the project are:
  - Increasing accuracy by **combining video feed from multiple cameras**  
A problem with our current model is **occlusion**, that is, when one player blocks the image of another player. This can be solved using multiple camera angles
  - **Increasing the field of view** of our video feed to cover the entire playing field. This will allow us to jump to the region of interest faster in case of a sudden long pass across the length of the playing field
  - Our current algorithm heavily employs the use of jersey colour to distinguish between teams. This may prove to be a shortcoming in certain cases and may be resolved by **using better algorithms** which are **not as dependent upon jersey colour**

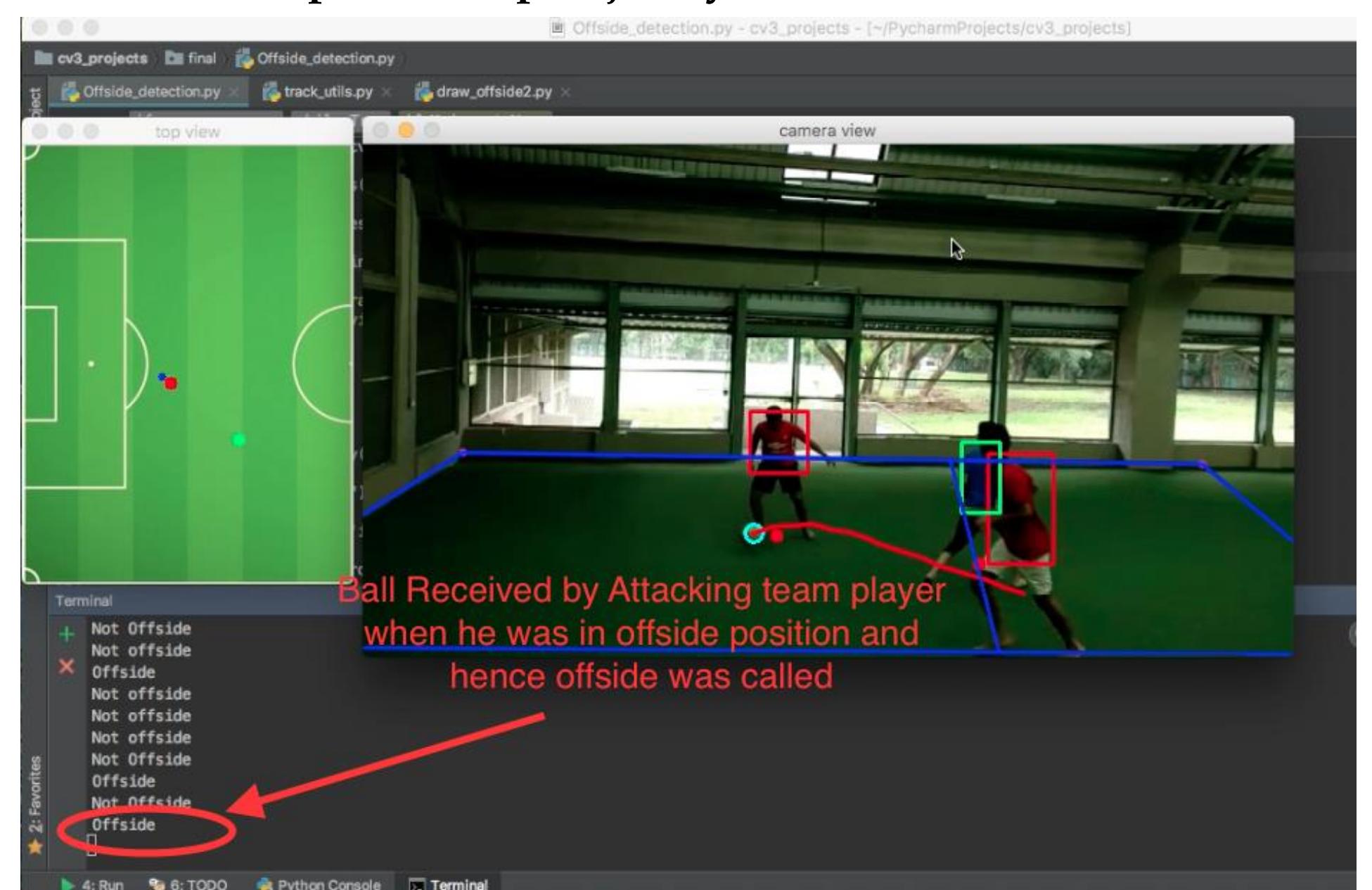


Figure 3: Finally it all comes together and delivers the expected results

#### ACKNOWLEDGEMENTS

- Our Mentors **Arunabh Ghosh** and **Ritik Madan** were of great help and support. Web Resources and online tutorials were essential towards completing our project. Some specific Web resources were
  - <https://tinyurl.com/techtitans1> <For Tutorials>
  - <https://tinyurl.com/academicPaper> <Earlier attempt at this problem>
  - <https://tinyurl.com/IOTbased>