

# Player Re-identification Assignment Report

## Project Overview

This project is aimed at building a robust Player Re-Identification system using a single video feed. The goal is to track and assign consistent IDs to players even when they leave and re-enter the frame. The system uses:

- YOLOv8 for player detection
- Norfair for tracking across frames
- OpenCV for visualization and video processing

All code and resources are shared via GitHub: [PlayerTracking-YOLOv8](#)

## Approach and Methodology

### 1. Player Detection

- Used YOLOv8n pretrained model for lightweight, real-time performance.
- Detected players in each frame and extracted bounding boxes.

### 2. Tracking with Norfair

- Transformed each bounding box into a center point (x, y).
- Fed these points into Norfair's tracker with a custom distance function (Euclidean).
- Assigned unique IDs to each detected player.

### 3. Visual Annotations

- Drew bounding boxes and center dots for each player.
- Overlaid consistent colors for each unique ID.
- Displayed an in-frame match timer (MM:SS).

### 4. Output Generation

- Created annotated video using OpenCV.
- Saved tracking log ([tracking\\_log.csv](#)) with `frame`, `id`, `cx`, `cy`.

## Techniques I Tried & Outcomes

| Technique Tried                     | Outcome  |
|-------------------------------------|--|
| Using YOLOv8n for detection         | Fast inference, good enough for small-scale test videos. |
| Norfair tracking with center points | IDs remained consistent even after partial occlusion.    |
| FPS Tuning and smoothing            | Helped sync detection frame rate and video playback.     |
| Random color mapping for each ID    | Visually helped identify players clearly.                |

## Challenges Encountered

1. Re-identification After Re-entry:
  - Difficult to maintain the same ID if the player re-enters after long occlusion.
  - Norfair may create a new ID due to the large motion gap.
2. Low Confidence Detections:
  - YOLO occasionally missed players in motion blur.
  - Could be improved with better model or frame pre-processing.
3. Tracking Accuracy:
  - False positives in background objects during certain frames.
  - Limited by model size (used YOLOv8n due to Colab constraints).
4. Google Colab Storage/Memory:
  - Frame-by-frame processing needed optimization to avoid timeouts.

## If I Had More Time / Resources...

- Use a Stronger YOLO Variant (YOLOv8m/l) for better accuracy.
- Train a Re-ID embedding model to track players based on appearance.
- Add smoothing filters for better bounding box stability.
- Use DeepSort instead of Norfair for feature-based tracking.
- Implement Multi-Camera Setup for cross-angle identification.

## Final Output

- Input Video: [15sec\\_input\\_720p.mp4](#)
- Code Notebook: [Players\\_TrackingCode.ipynb](#)
- Output Video: [player\\_tracking\\_output.mp4](#)
- Log File: [tracking\\_log.csv](#)

Repo: <https://github.com/SasiRekhaSadanala/PlayerTracking-YOLOv8>