

# Project Name: Person Re-identification

Course Name: CSE623 Machine Learning Theory and Practice

**Professor Name:** Prof Mehul Raval **University:** Ahmedabad University

#### Team Members:

- 1. Shlok Shelat AU2240025
- 2. Purvansh Desai AU2240036
- 3. Shrey Salvi AU2240033
- 4. Rushi Moliya AU2240020
- 5. Dhananjay Kanjariya AU2240023



## Summary of Work Done This Week:

This week, our team focused on understanding the fundamentals of **Person Re-identification**, including key challenges and solutions from state-of-the-art research. We reviewed existing datasets, particularly **PRW** (**Person Re-identification in the Wild**), which provides large-scale annotated pedestrian images captured from multiple cameras. Our main tasks included:

#### 1. Understanding the Re-ID Pipeline:

- Explored the two key components: **pedestrian detection** and **person recognition**.
- Studied the impact of different detection and recognition models on reidentification accuracy.

## 2. Dataset and Benchmarking:

- Analyzed the PRW dataset, which contains 932 identities and 11,816 frames.
- Reviewed benchmarking results for various CNN architectures such as AlexNet,
   VGGNet, and ResNet.

# 3. Methods for Performance Improvement:

- Examined **cascaded fine-tuning**, where a detection model is trained first, followed by a classification model.
- Understood Confidence Weighted Similarity (CWS), which integrates detection scores into similarity measurements for better accuracy.

#### 4. Challenges Identified:

- The need for **high-precision pedestrian detection** to improve re-ID accuracy.
- The trade-off between **detection recall and false positives**, affecting the quality of gallery images.
- The importance of **feature extraction techniques** for better representation learning.

#### Plan for Next Week:

Based on our findings, we will move on to Feature Extraction and Optimization Techniques for re-identification. Our key objectives will be:

#### 1. Feature Extraction Methods:

- Implement **CNN-based embeddings** for better identity discrimination.
- Compare hand-crafted features vs deep-learning features in terms of accuracy.

# 2. Fine-tuning Detection Models:

- Train pedestrian detectors on **custom datasets** to improve performance.
- Evaluate different CNN architectures for feature extraction.

#### 3. Metric Learning Techniques:

• Experiment with distance metrics (Euclidean, cosine similarity, and learned metrics) to improve person matching.

# 4. Preliminary Implementation:

• Start implementing bounding box annotation and object detection using deep-learning frameworks.



# Conclusion:

Our team has gained a strong theoretical understanding of **Person Re-identification** and its challenges. Moving forward, we will focus on **feature extraction** and optimizing detection models to improve accuracy.