Project Title: Hybrid Deep Learning Pipelines for Gender Classification & Face Verification

Problem Statement

Develop two systems:

- Task A: Classify Genders by training through the given images
- Task B: Verify whether two face images belong to the same person, even under distortion.

Task A - Gender Classification

Approach: Binary classification using deep feature fusion + traditional ML.

Pipeline:

- Preprocessing: Resize (224×224), normalize, label from folders.
- Feature Extraction:
 - o ResNet-50 (2048D), EfficientNet-B3 (1536D), DeepFace (VGG-Face) (2622D).
- **Feature Fusion:** Concatenated features → normalized via StandardScaler.
- Classification Models:
 - Logistic Regression (scikit-learn)
 Support Vector Machine (SVM with linear kernel)
- **Evaluation:** Accuracy, Precision, Recall, F1-score (weighted).

Task B - Face Verification

Approach: Face verification using a Siamese neural network + hard negative mining.

Pipeline:

- Face Detection: DeepFace.extract_faces() (OpenCV backend), normalized & cached.
- Embeddings: DeepFace.represent() using ArcFace (512D).
- · Pair Generation:
 - Positive: same identity;
 - Hard Negative: cosine similarity with other identities.
- Model:
 - Siamese Net: [Embedding A + B] \rightarrow FC(1024 \rightarrow 256 \rightarrow 1) \rightarrow Sigmoid.
- Training: Binary Cross Entropy loss, Adam optimizer.
- **Prediction:** Match if score > threshold; else predict "unknown".

• **Evaluation:** Accuracy, Precision, Recall, F1-score.

Key Innovations

- **Hybrid Feature Fusion:** Combines CNNs + DeepFace for rich representations.
- Hard Negative Mining: Ensures challenging training pairs.
- Open-Set Support: Threshold-based unknown face detection.
- Efficient Caching: For embeddings, faces, and labels.
- **Concatenating Model Outputs:** For gender classification, led to reduction in overfitting and better performance.

Tools & Technologies

- Python, PyTorch, DeepFace, scikit-learn, OpenCV
- Pretrained models: ResNet50, EfficientNet-B3, ArcFace, VGG-Face