

# Face Recognition using PCA + ANN – Interview Cheat Sheet

## Workflow:

1. Dataset: 450 images (100×100 pixels, 9 persons)
2. Preprocessing: Resize → Grayscale → Flatten
3. PCA:
  - Compute mean face
  - Extract eigenfaces (principal components)
  - Project images into k-dim feature space (k=50)
4. ANN:
  - Input: PCA features (50)
  - Hidden Layers: Dense + ReLU
  - Output: Softmax (9 classes)
  - Loss: Categorical Cross-Entropy
  - Optimizer: Adam
5. Results:
  - Accuracy: ~50%
  - Saved model (face\_ann\_model.h5)
  - Visualizations: mean face, eigenfaces
6. Prediction: New image → PCA → ANN → Predicted Person

## Key Concepts:

- PCA: Reduces dimensionality, keeps max variance → eigenfaces
- Eigenfaces: Principal components representing max variance in faces
- ANN vs CNN: ANN after PCA, CNN learns features directly
- Why PCA? Avoids training on 10,000 pixels → reduced to 50 features
- Why ANN? Learns non-linear boundaries on PCA features

Q: Why PCA?	A: To reduce dimensionality, remove noise, and avoid overfitting.
Q: Why ANN (not CNN)?	A: Simplicity – PCA extracts features. CNN would perform better.
Q: Accuracy is only ~50%. Why?	A: Small dataset, ANN overfitting. CNN or transfer learning would help.
Q: Challenges?	A: Dataset structuring, imports, preventing overfitting, ensuring PCA→ANN compat
Q: Improvements?	A: Use CNN (e.g., VGGFace), data augmentation, larger dataset, real-time deploym

## Workflow Diagram:

Image (100x100) → Flatten → PCA (50 comps)  
↓  
PCA Features  
↓  
ANN Classifier  
↓  
Predicted Person (Softmax)