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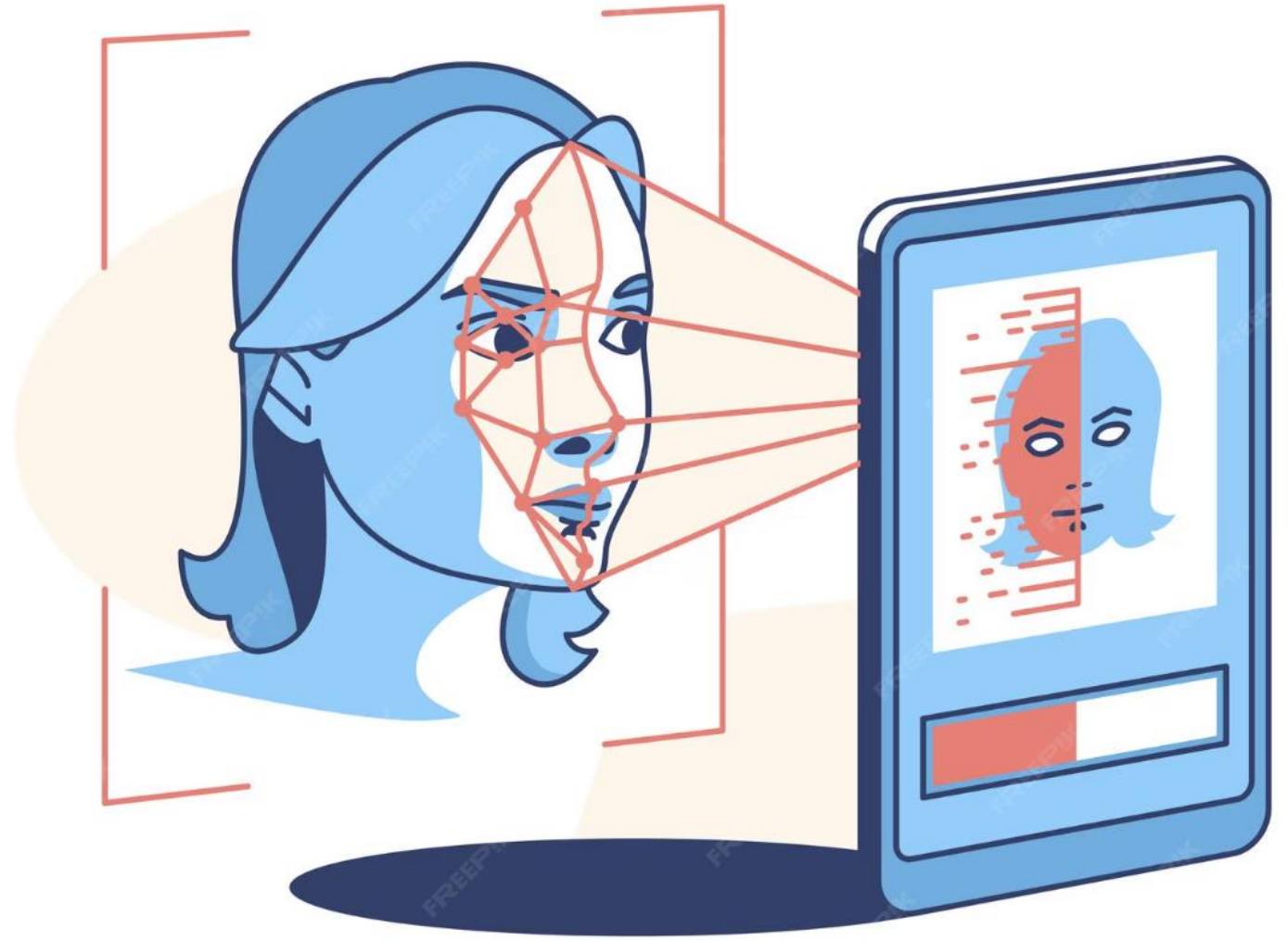
# DEEP LEARNING FOR FACE RECOGNITION: RESEARCH & MINI- PROJECT SUMMARY

# INTRODUCTION TO FACE RECOGNITION

1. Widely used in security, authentication, surveillance
2. Traditional methods struggled: lighting, pose, occlusion
3. Deep learning transformed the field

# PROJECT MOTIVATION

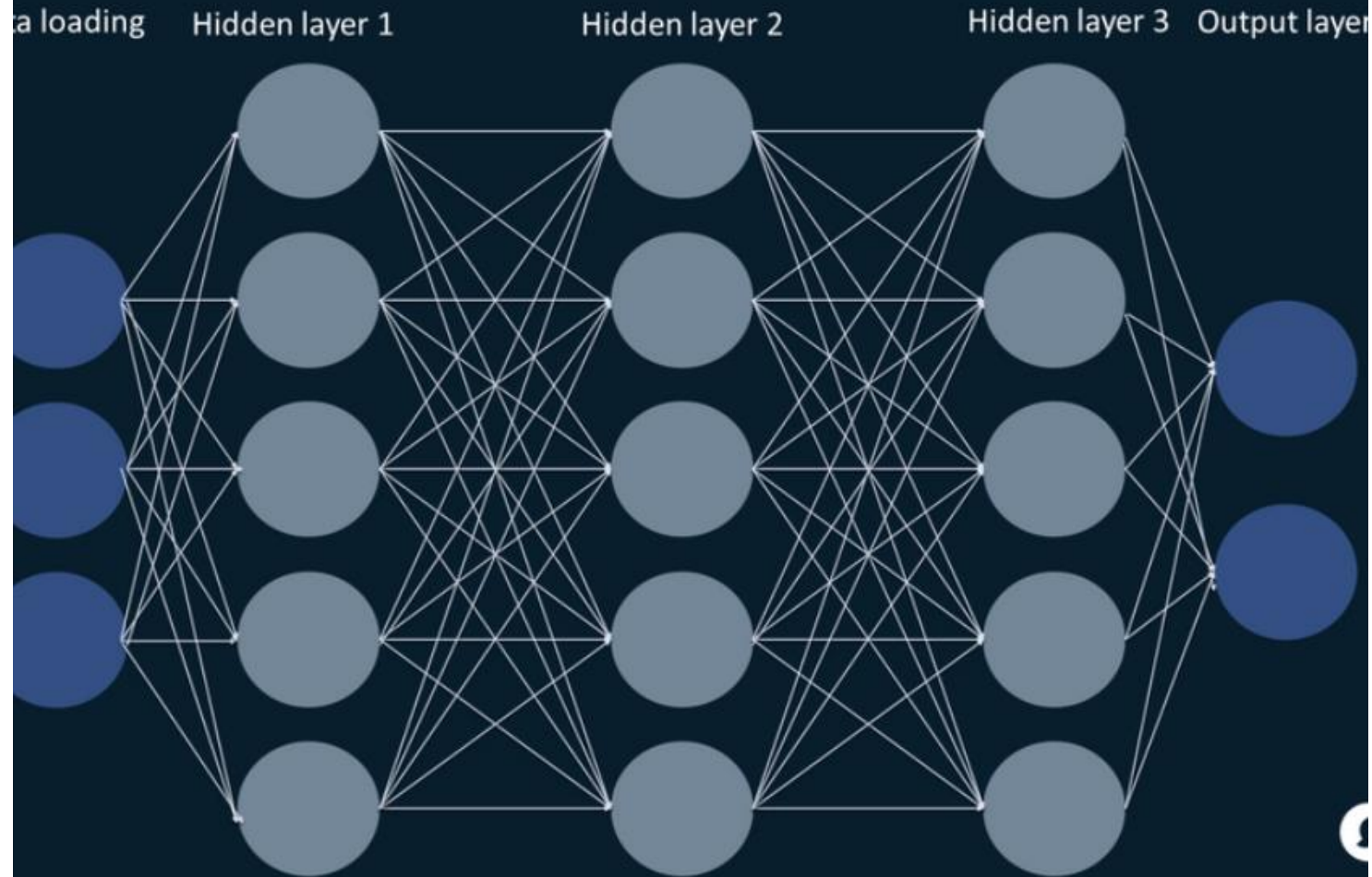
- Mini-project complements research paper
- Build end-to-end face recognition pipeline
- Evaluate classification + verification approaches



- 3,000+ images, 62 identities
- Real-world variations: pose, illumination, occlusion
- Pre-processed to RGB, resized to 160×160

DATASET (LFW SUBSET)

# DEEP NEURAL NETWORKS ARCHITECTURE



## EMBEDDING EXTRACTION

- Used MobileNetV2 pretrained model
- Generates 1280-dimensional embeddings
- Basis for all further models





# CLASSIFICATION MODELS

## **Logistic Regression**

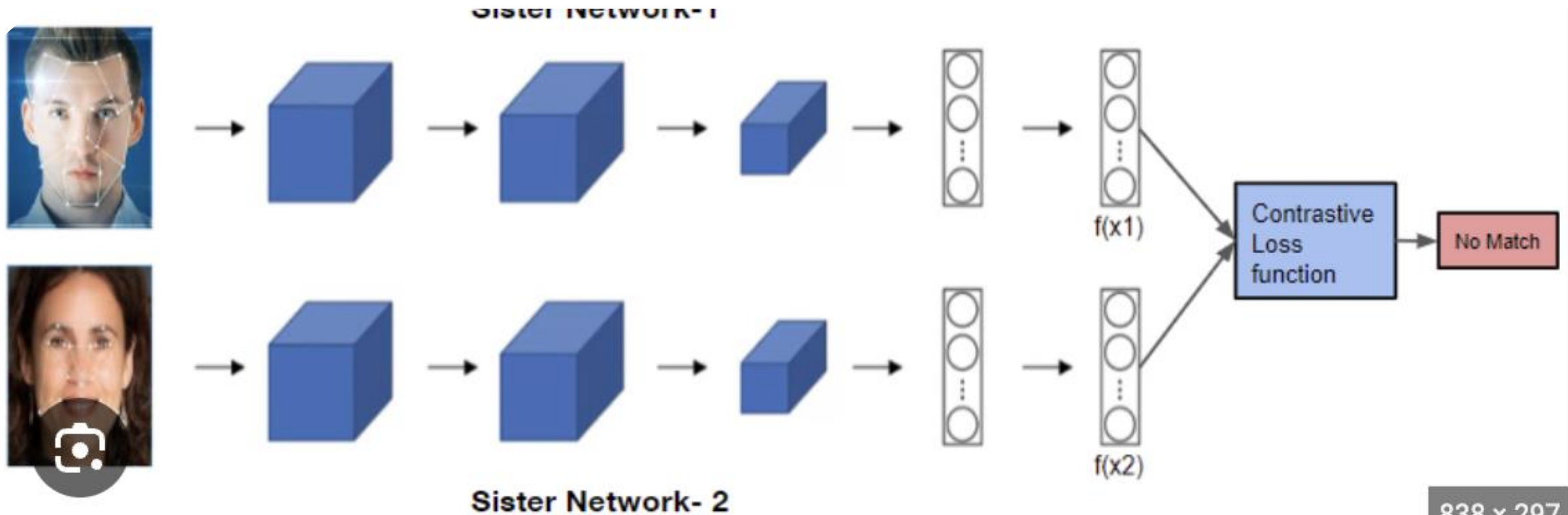
- Linear baseline

## **SVM (RBF)**

- Non-linear separation
- Often performs better on deep embeddings

# SIAMESE NETWORK FOR VERIFICATION

- Twin networks with shared weights
- Contrastive loss
- Determines if two faces belong to same identity



- **LOGISTIC REGRESSION: 69.3%**
- **SVM: 65.7%**
- **SIAMESE NETWORK: 52.3%**

## RESULTS SUMMARY



# VISUALIZATIONS:

t-SNE
Shows partial clustering of identities
Overlap indicates limited separation

Confusion Matrix
• Correct predictions along diagonal
• Misclassifications due to class imbalance

# KEY INSIGHTS

- K Deep embeddings useful even without CNN training
- Class imbalance affects accuracy
- Verification difficult without face-specific embeddings

# LIMITATIONS AND FUTURE DIRECTIONS



LIST THE  
LIMITATIONS

- Embeddings not face-trained
  - Siamese network trained on embeddings, not full images
  - No adversarial or fairness evaluation
- 
- Fairness and bias reduction
  - Privacy-preserving methods (Federated Learning)
  - Lightweight on-device models

# **Research Insights: Modern Advances**

## **Content:**

- FaceNet, ArcFace, Transformers
- GANs for augmentation
- Anti-spoofing and multimodal biometrics

# CONCLUSION

- Built full pipeline: embeddings → classification → verification
- Results match real-world challenges
- Practical experience deepened understanding



# THANK YOU

AML FINAL PROJECT PRESENTATION