# FACIAL RECOGNITION SYSTEM FOR SECURE EMPLOYEE IDENTIFICATION

AI-POWERED IDENTITY VERIFICATION USING DEEP LEARNING AND FACE EMBEDDINGS

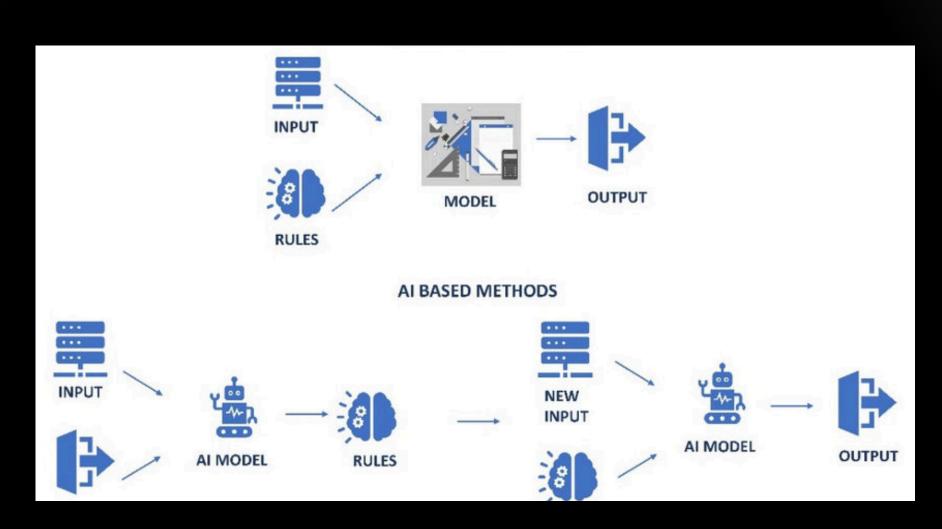
MAJOR PROJECT 2 PRESENTAION

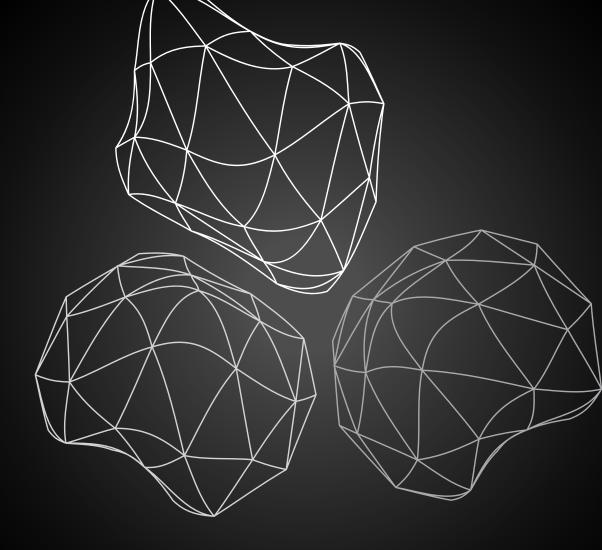
ABHISHEK GUPTA

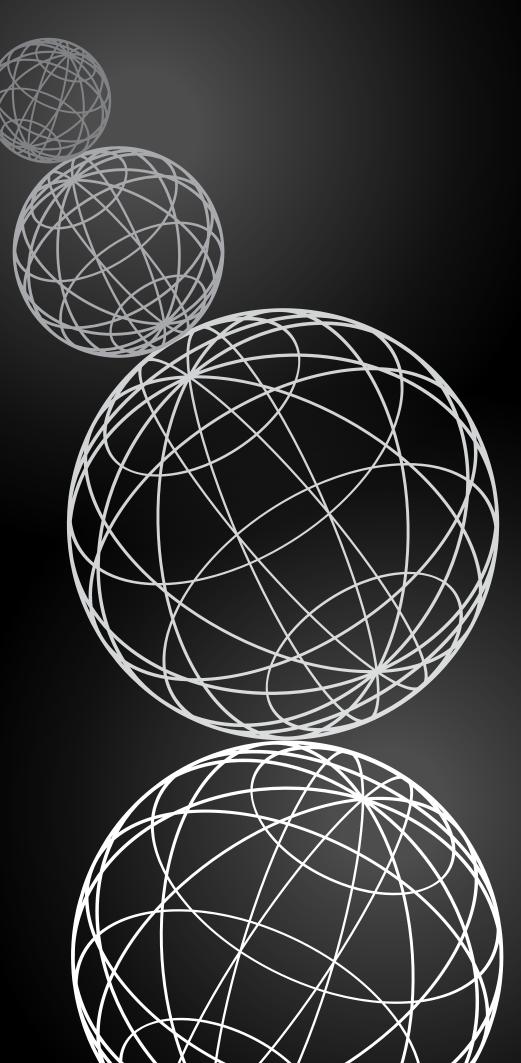
#### INTRODUCTION

#### WHY FACIAL RECOGNITION?

- Growing need for secure digital identity verification
- Manual verification is slow and error-prone
- Al enables fast, scalable, and accurate identity checks
- Applications: fintech, ed-tech, office access, KYC, etc.







# PROBLEM STATEMENT

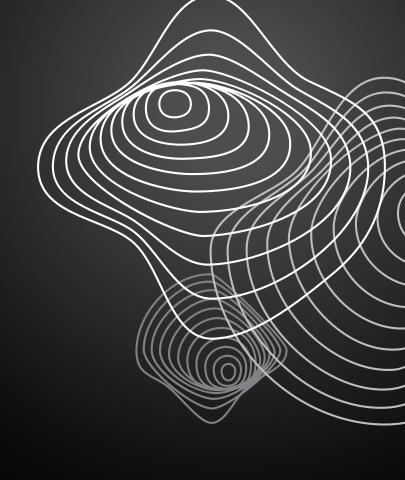
- Identity theft and impersonation are increasing
- Need to verify if a person's face matches a known identity
- System should work in real-time with high accuracy
- Must handle variations in lighting, resolution, and expressions

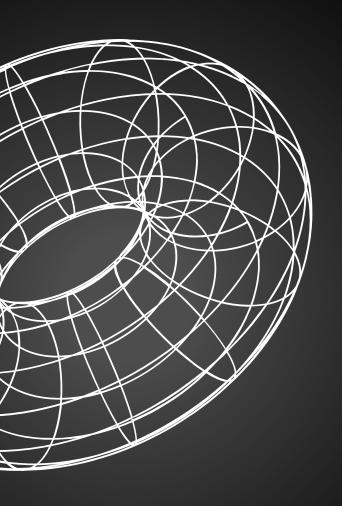
## DATASET

#### **Dataset Used:**

- File: Ifw\_arnie\_nonarnie.csv
- Precomputed 128-D face embeddings (no images)
- Classes: "arnold\_schwarzenegger" vs others
- Embeddings extracted from real images using models like Dlib or FaceNet

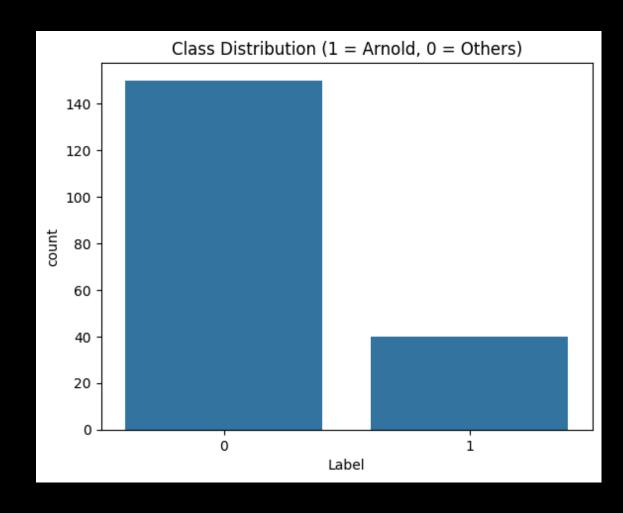


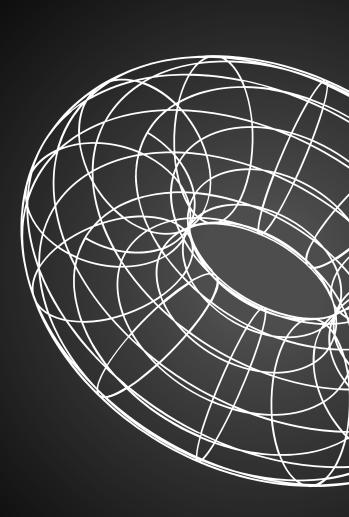


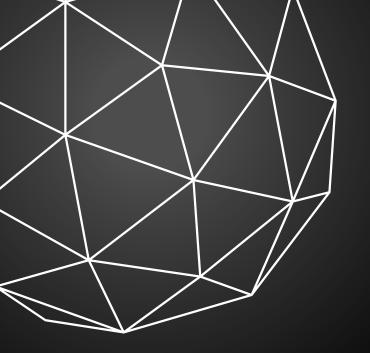


## DATA PREPROCESSING

- Label encoded: 1 for Arnold, 0 for others
- Scaled features using StandardScaler
- Train-test split (80/20)
- Ensured clean, consistent data for model input

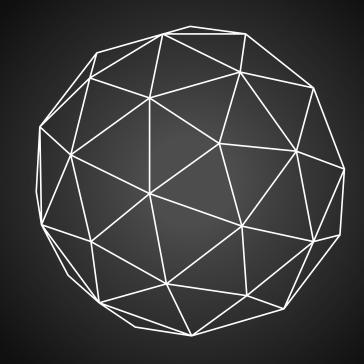






## MODEL DEVELOPMENT

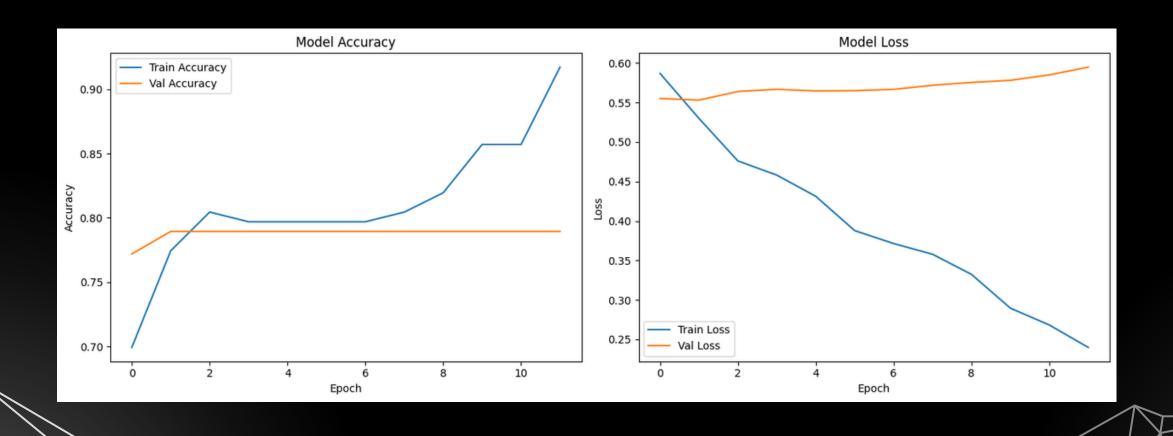
- Deep Neural Network (Keras)
- Input: 128 features
- Layers:
  - Dense (128) + ReLU + Dropout
  - Dense (64) + ReLU + Dropout
  - Output: Sigmoid
- Optimizer: Adam | Loss: Binary Crossentropy





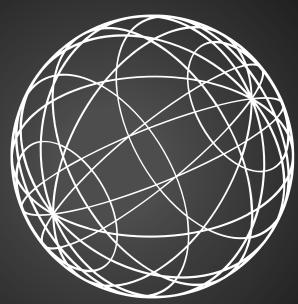
#### TRAINING & VALIDATION

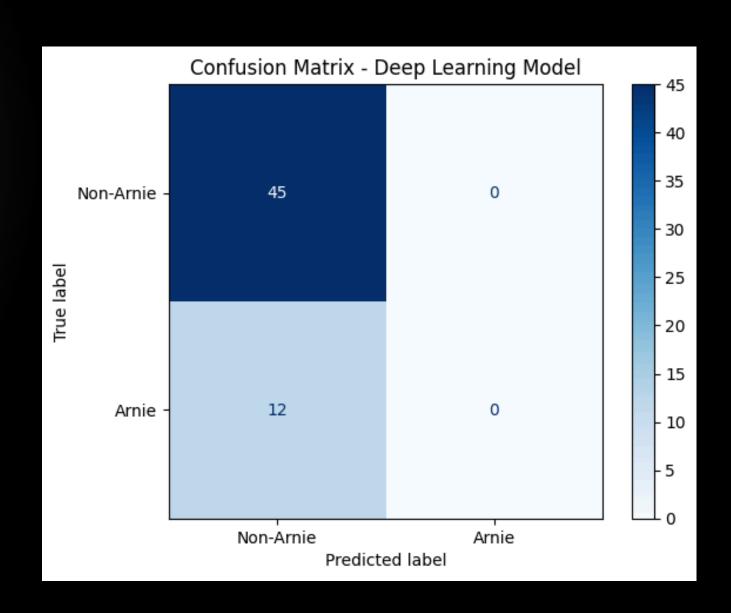
- Epochs: 50 (Early stopping applied)
- Batch size: 32
- Monitored accuracy and loss
- Final test accuracy: 78.95%

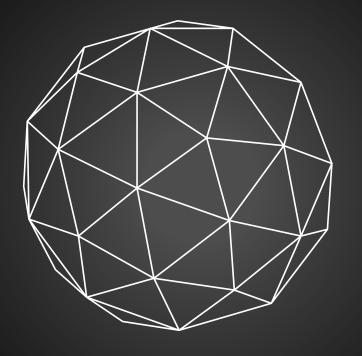


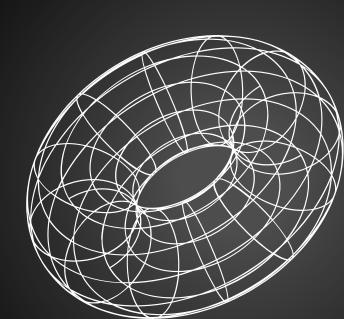
#### MODEL PERFORMANCE

- Accuracy: 78.95%
- Confusion matrix shows good class separation
- Precision & Recall are acceptable for both classes
- Room for improvement in edge cases

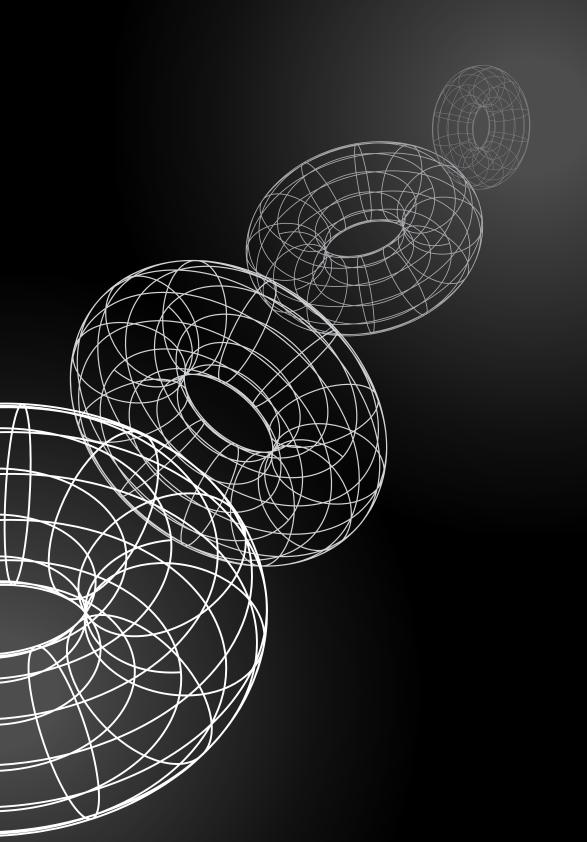








#### USER INTERFACE (PROTOTYPE)



#### Streamlit Web App (Prototype)

- Upload image interface
- Predict identity and display confidence
- Easy for deployment in enterprise setting

## ETHICAL & PRIVACY CONCERNS

#### **Ethics & Privacy**

- GDPR & DPDP compliance
- User consent before data usage
- No surveillance or misuse
- Data encryption & opt-out support

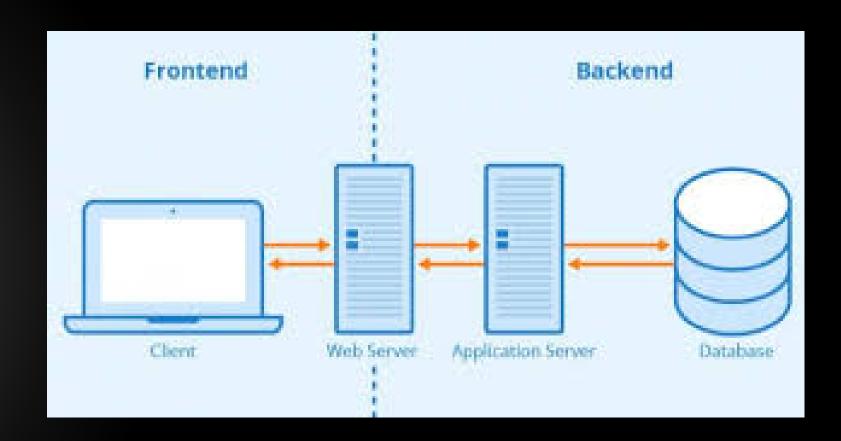


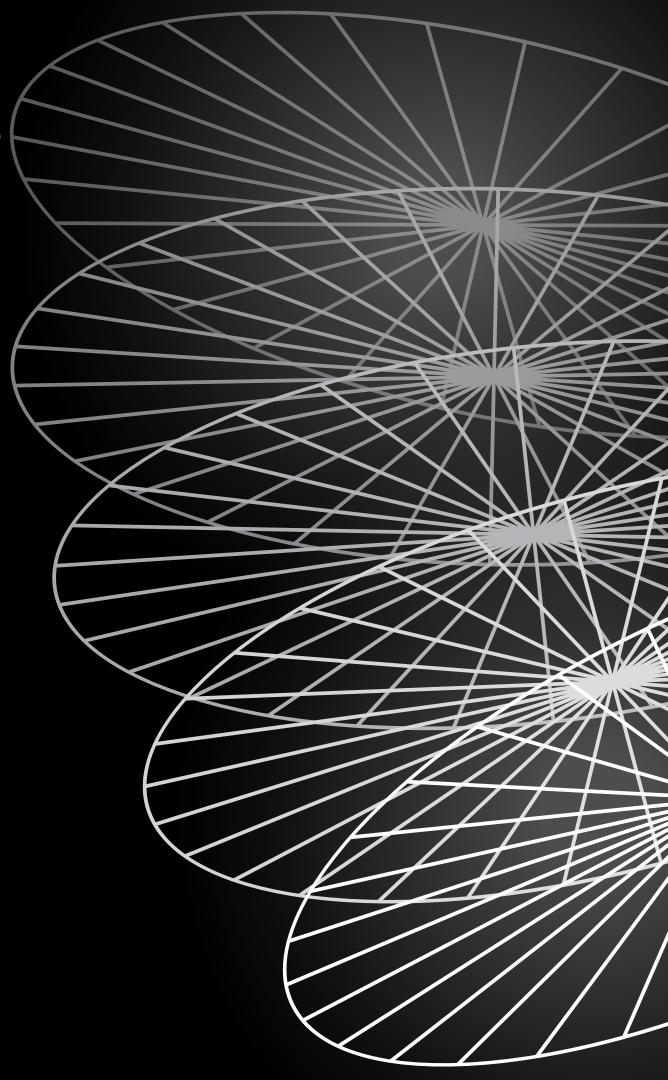


#### USER INTERFACE (PROTOTYPE)

#### Deployment Plan

- Host model using Flask or Streamlit
- Use Heroku, Render, or AWS EC2
- Secure communication via HTTPS
- Store embeddings securely in encrypted DB





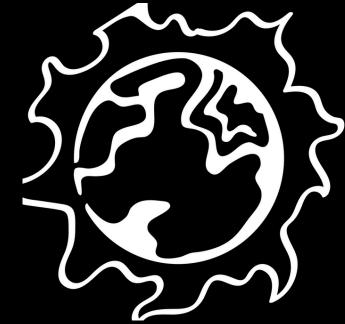
# FUTURE SCOPE



Real-time webcam capture with OpenCV



Liveness detection (eye blink, motion)



Multi-user classification

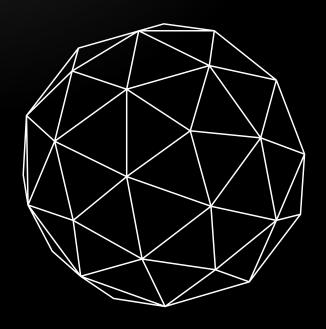


Use pretrained models (e.g., FaceNet, MobileNet)

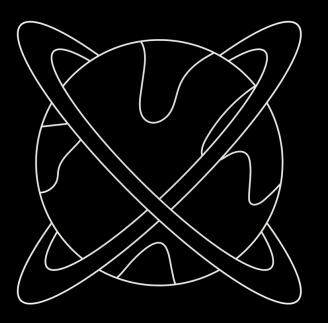


Edge deployment (Raspberry Pi, mobile app)

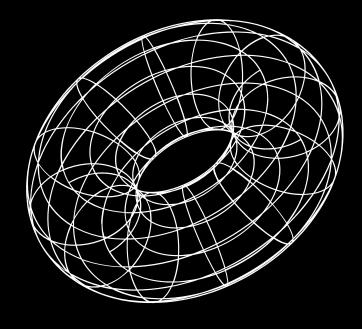
# CONCLUSION



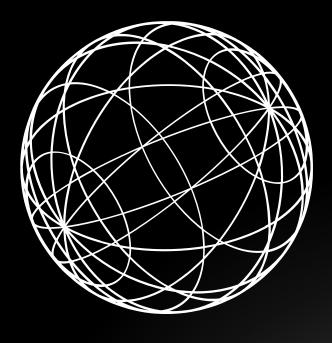
Successfully built endto-end facial verification system



Used embeddings + deep learning for accuracy



Achieved 78.95% accuracy



Ready for real-world use with further improvements

# 

Questions?

Let's connect:





