

Biometric systems 2025/2026

Evaluation of Vision Transformers for face verification under attacks and appearance variations

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Problem Statement

Introduction

Face Verification



Face verification is the process of confirming a person's identity by comparing two facial images and deciding if they belong to the same individual.

Real-world conditions can significantly affect verification performance. In addition, **spoofing attacks** represent a serious threat to system reliability.



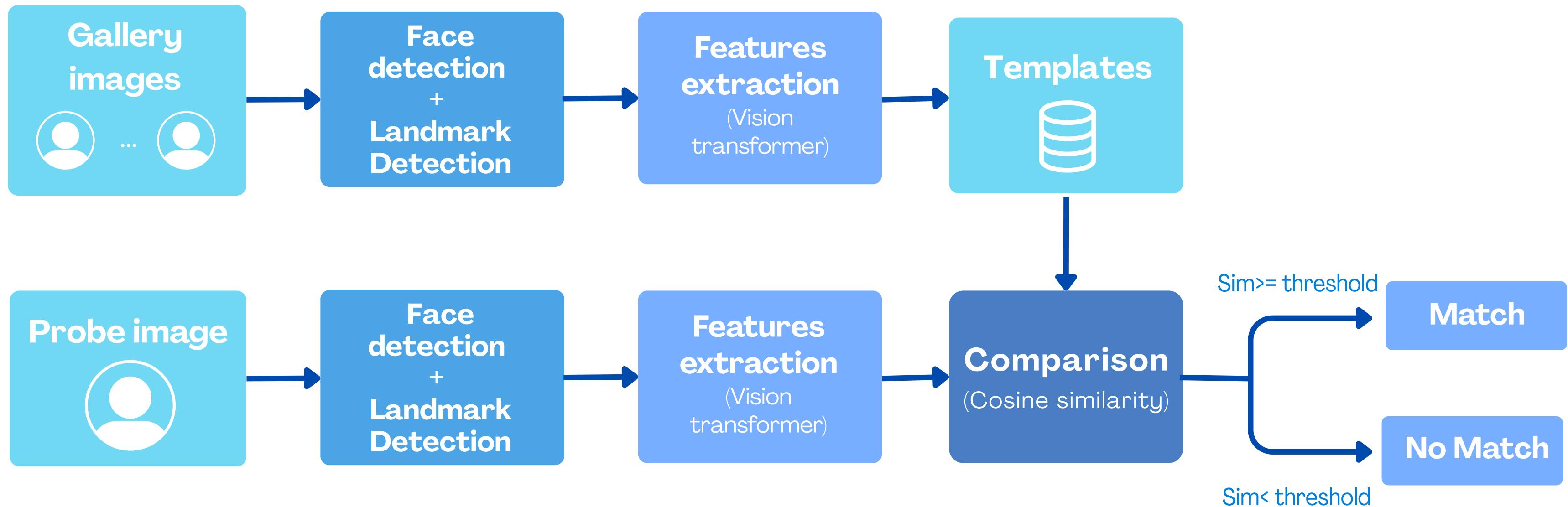
Objectives



- **Analyze** a ViT-based verification model in real-world scenarios
- **Evaluate** performance on a custom dataset
- **Assess** resistance to attacks

Project overview

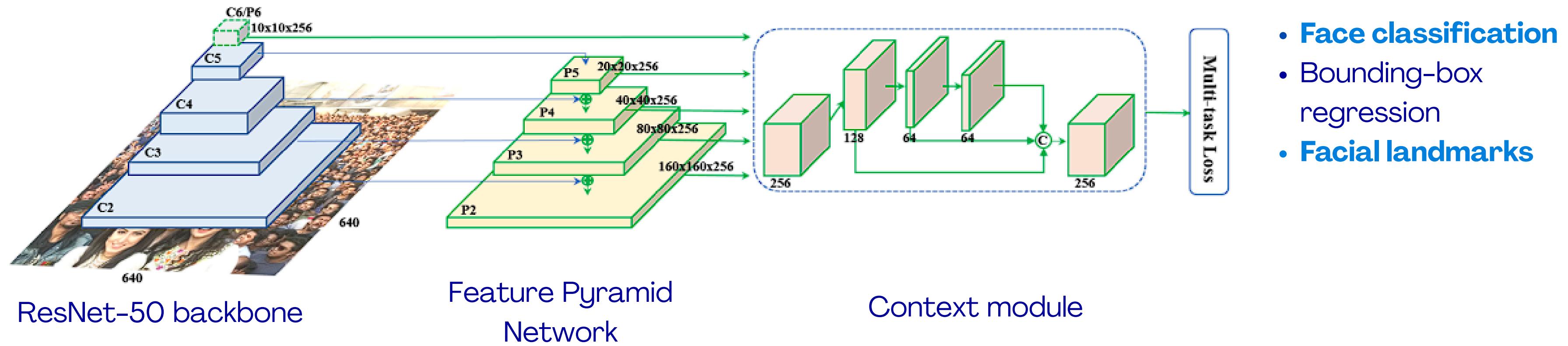
Full pipeline



Model Architectures (1)

RetinaFace

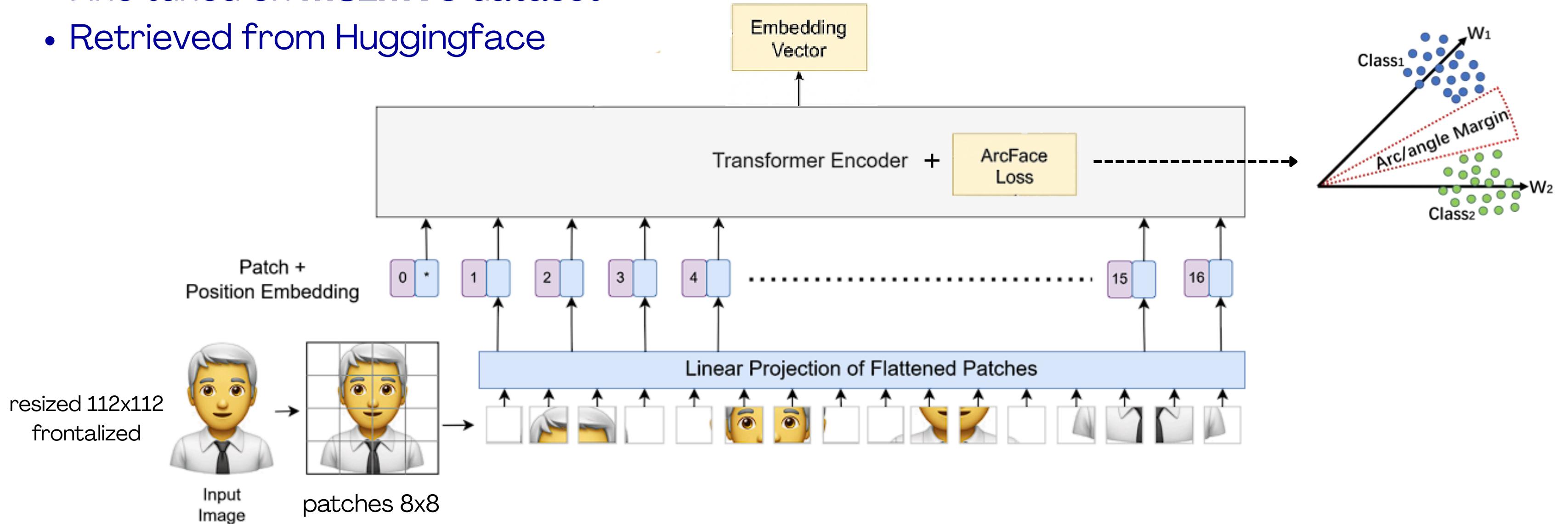
- Trained on **WIDER FACE** dataset
- **Anchor-based** training strategy



Model Architectures (2)

ViT

- Fine tuned on **MS1MV3** dataset
- Retrieved from Huggingface



- **Arcface loss** for discriminative embeddings:

$$\mathcal{L} = \frac{-1}{N} \sum_{i=1}^N \log \frac{e^{s \cdot \cos(\theta_{y_i} + m)}}{e^{s \cdot \cos(\theta_{y_i} + m)} + \sum_{j \neq y_i} e^{s \cdot \cos(\theta_j)}}$$

Datasets (1)

User variations

- Gallery: **212 images** (3 for each identity)
- Each category contains **12 images** (27 for interpersonal)
- Overlap categories



Make-up



Age



Plastic surgery

Datasets (2)

Potential attacks



Print attacks



Camouflage



Deepfake



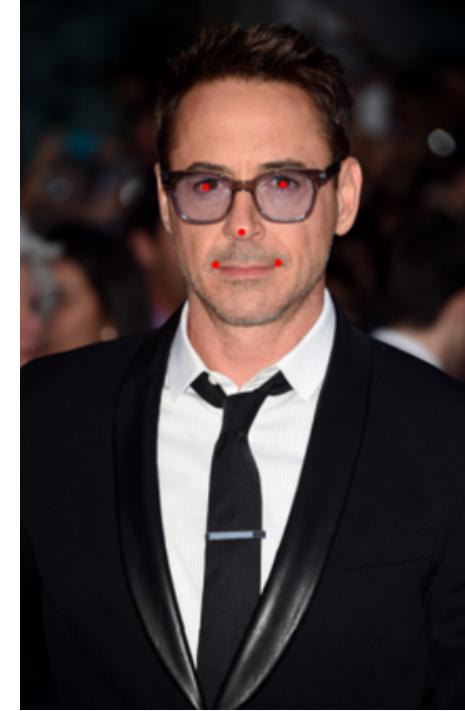
Interpersonal

Proposed method (1)

Step 1: Landmarks detection and Face alignment

- **Frameworks:** OpenCV, insightface library

1. **Face localization and landmarks extraction** (buffalo_l model, RetinaFace-10GF)
2. **Matching and warping** with estimateAffinePartial2D and warpAffine
3. **Result:** normalised 112×112 face

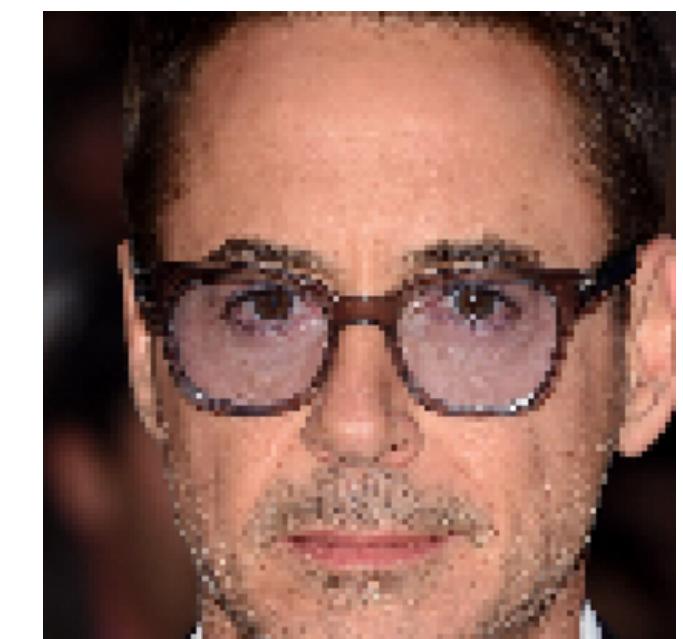


LANDMARK	X	Y
Left eye	38.2946	51.6963
Right eye	73.5318	51.5014
Nose tip	56.0252	71.7366
Left mouth	41.5493	92.3655
Right mouth	70.7299	92.2041

Standard ArcFace 5-point template

Original image

Image with landmarks



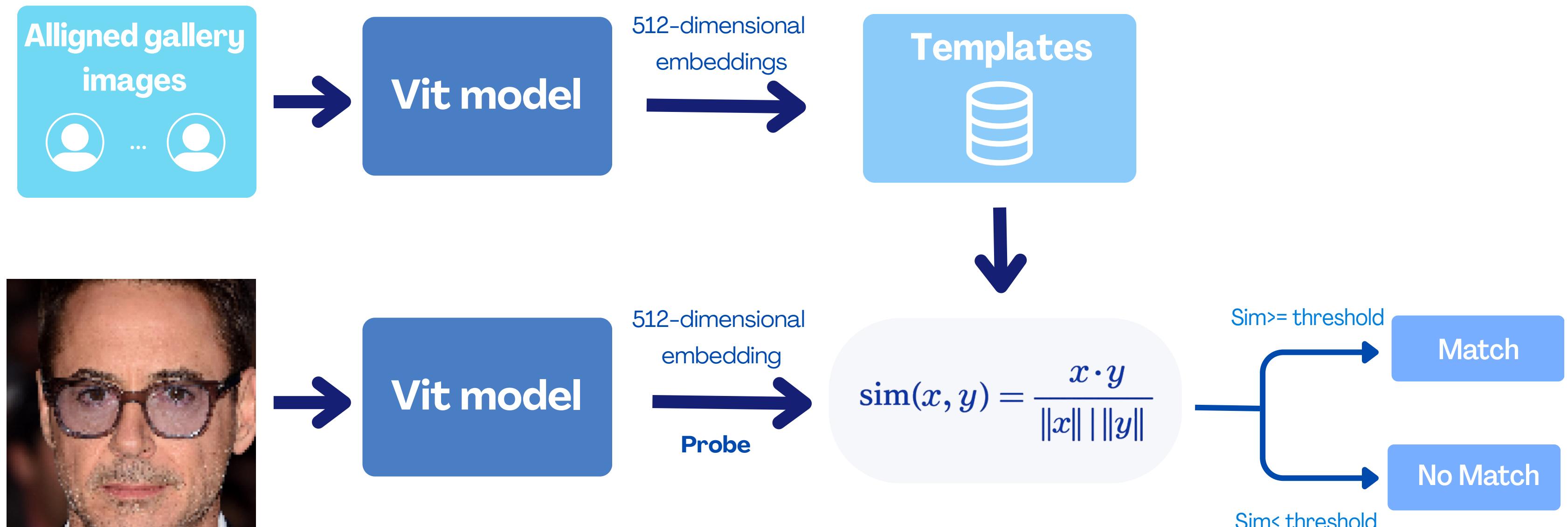
Aligned image (112×112)

*same processing for gallery images

Proposed method (2)

Step 2: Features Extraction and Similarity Computation

- Thresholds from 0 to 1 (step of 0.5)



*Probe vs Gallery comparisons are used to generate genuine and impostor verification trials

Proposed method (3)

Step 3: Evaluation

Scores

- Genuine Acceptance
- False Acceptance
- Genuine Rejection
- False Rejection

Graphics

- FAR
- FRR
- GAR
- GRR
- ERR
- ROC
- DET
- Margin

Results (1)

Detection and alignment

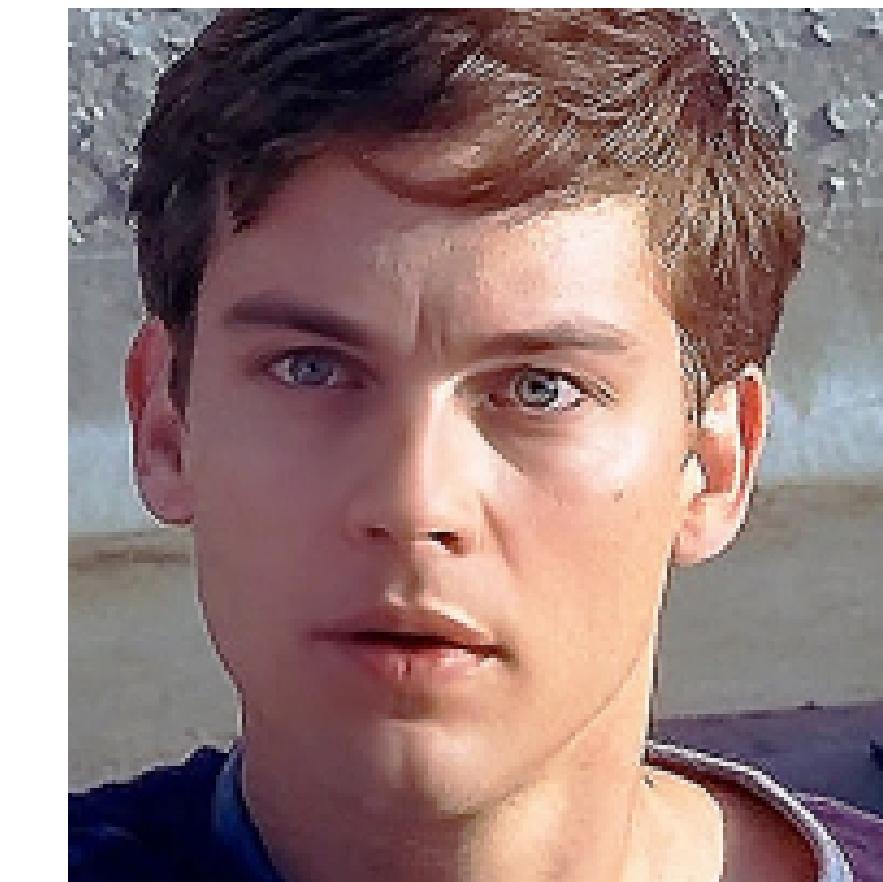
- Images where the face detector did not produce any bounding box:



(a)



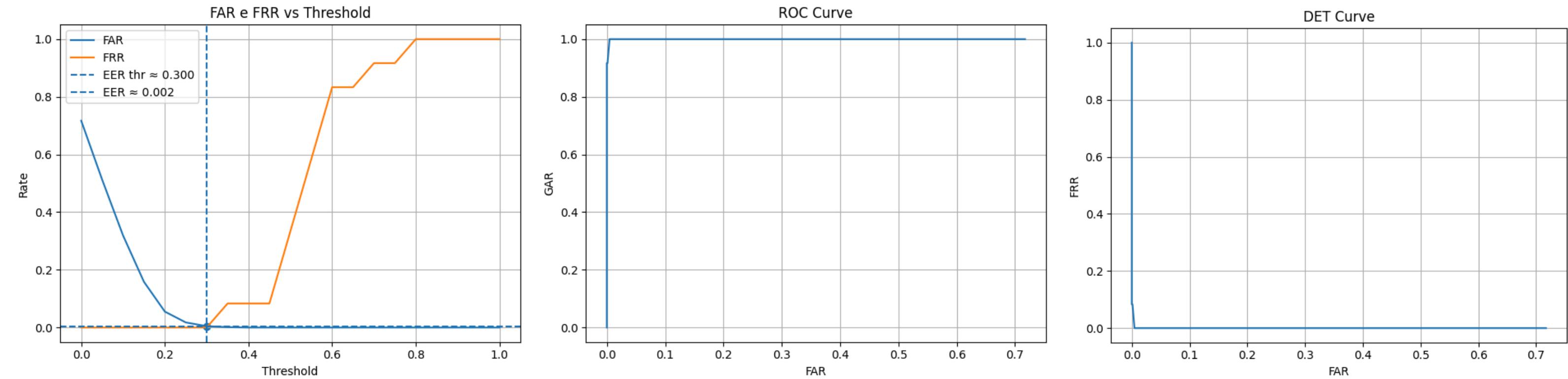
(b)



(c)

Results (2)

Age



Genuine acceptances:



Best Match: Zac Efron

Similarity score: 0.7523

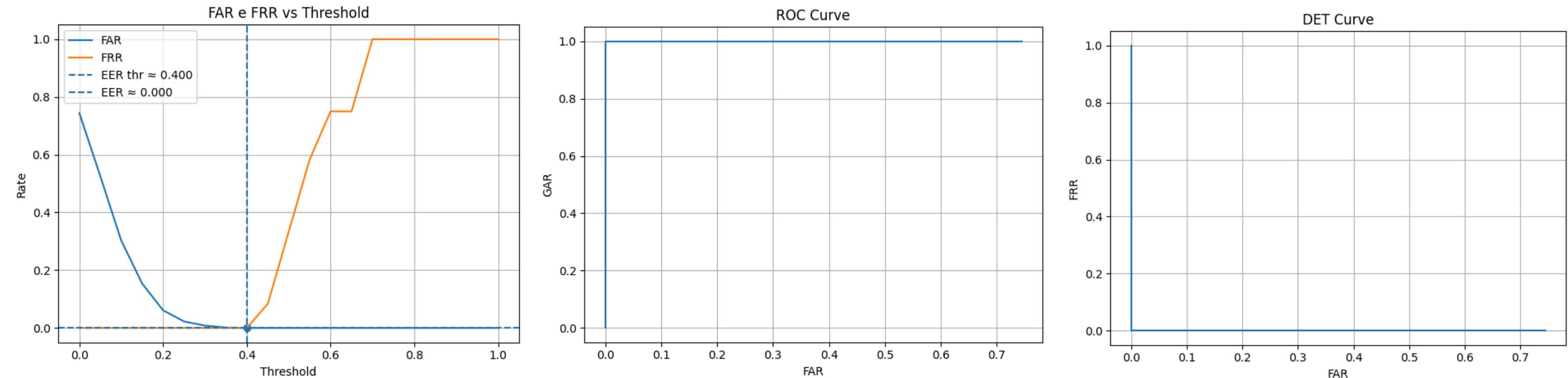


Worst Match: Queen Elizabeth

Similarity score: 0.2571

Results (3)

Camouflage



Genuine acceptances:



Best Match: Harry Styles

Similarity score: 0.6961

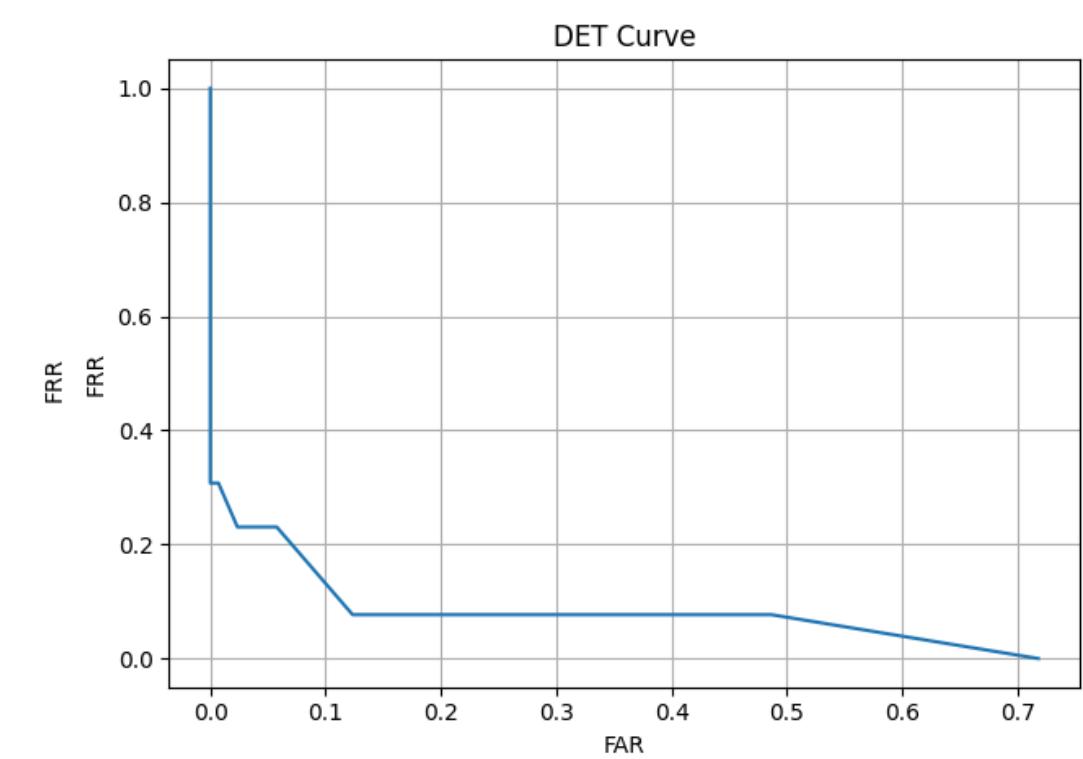
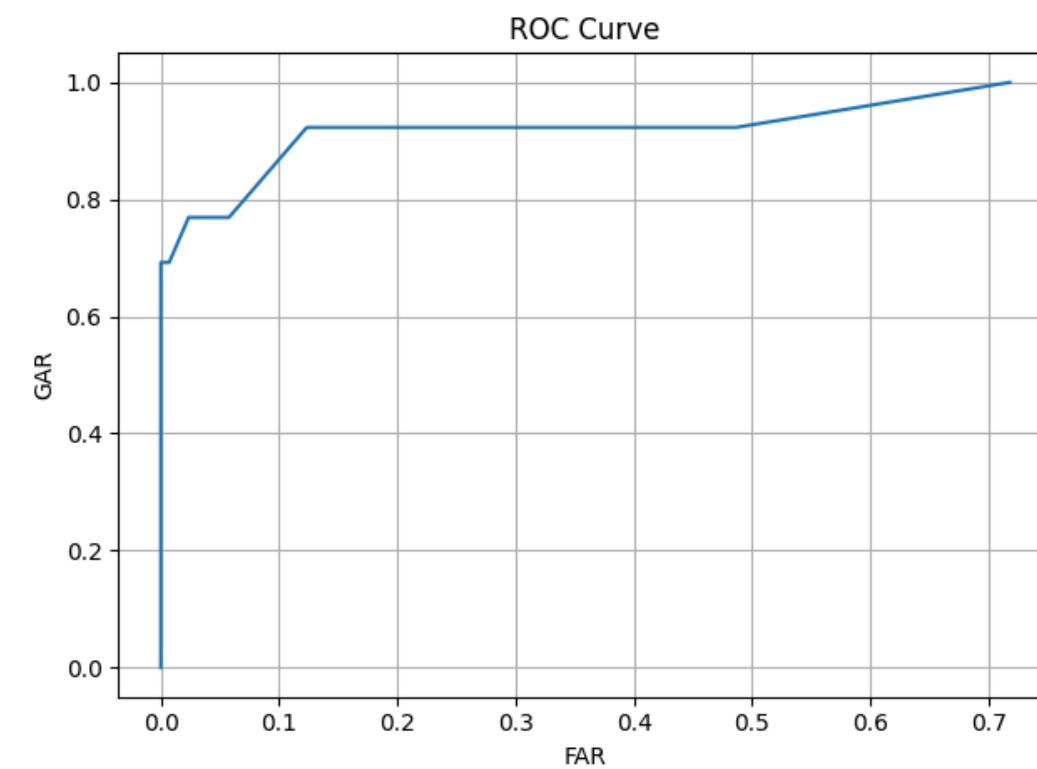
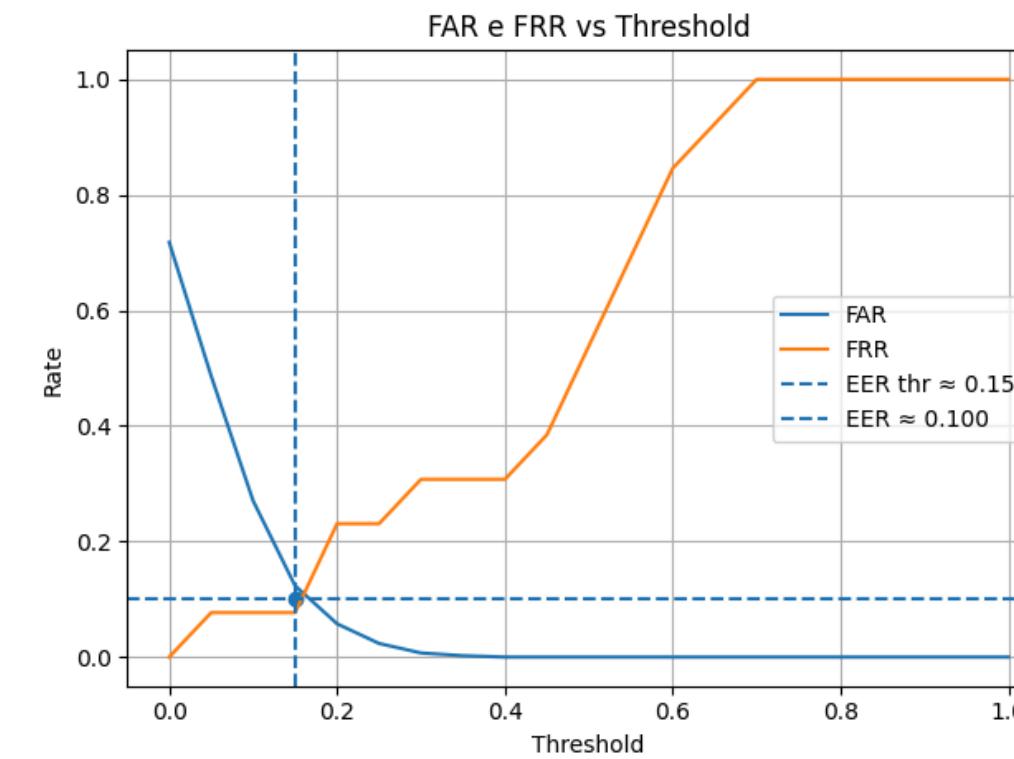


Worst Match: Chris Evans

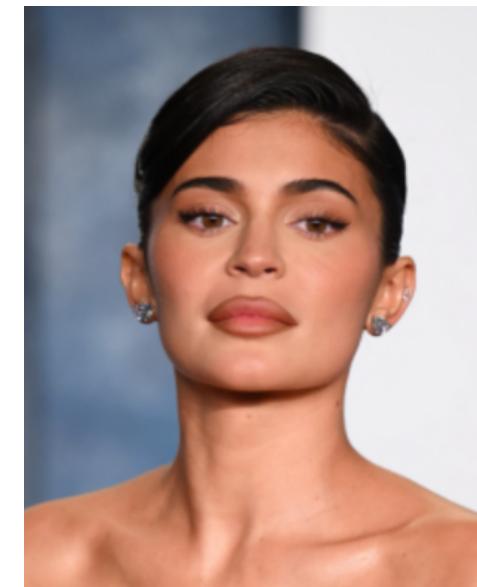
Similarity score: 0.3594

Results (4)

Make up



Genuine acceptances:



Best Match: Kylie Jenner

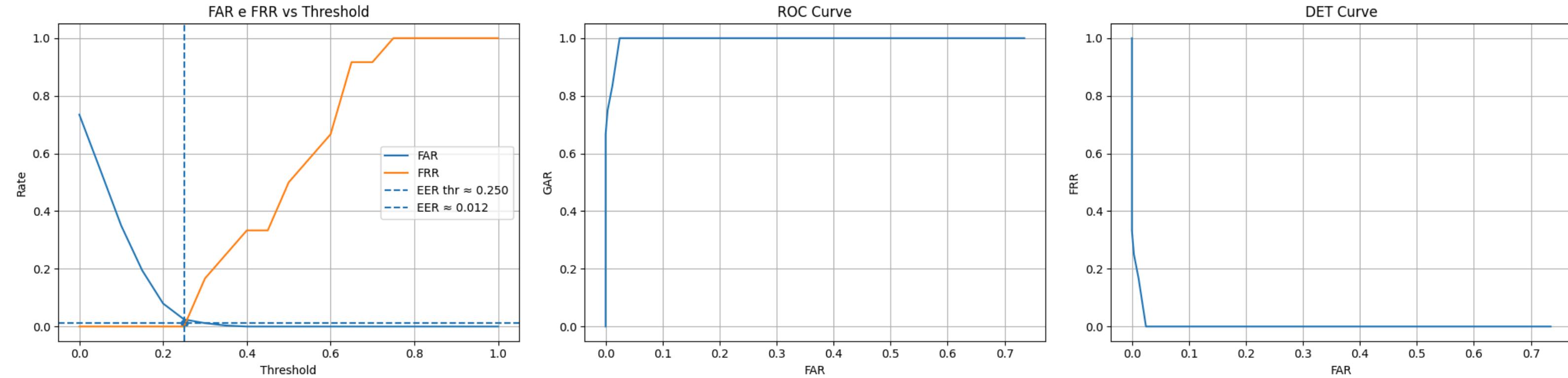
Similarity score: 0.6662

Worst Match: Robin Williams

Similarity score: 0.0232

Results (5)

Plastic Surgery



Genuine acceptances:



Best Match: Zac Efron

Similarity score: 0.7205

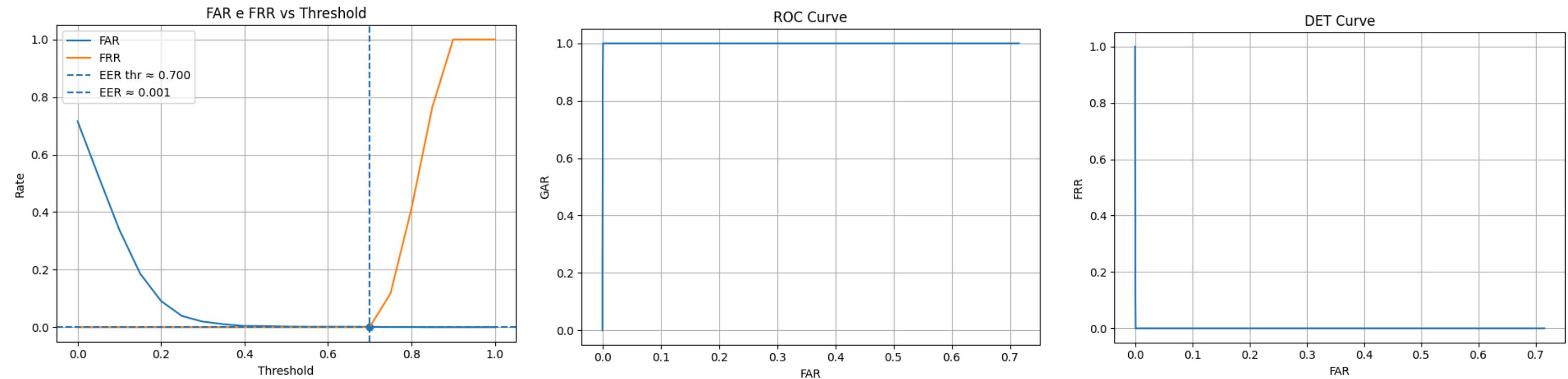


Worst Match: Lindsay Lohan

Similarity score: 0.1889

Results (6)

Inter-personal



Genuine
acceptances
and false
acceptances:



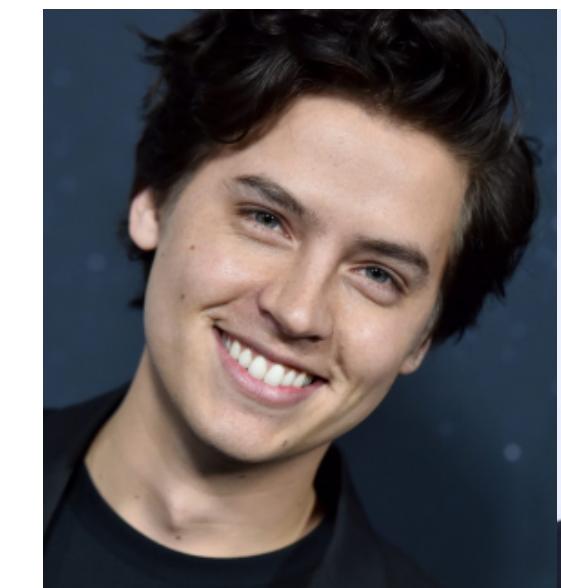
Probe
(In gallery)



Genuine match
Sim score: 0.8281



Impostor match
Sim score: 0.4920



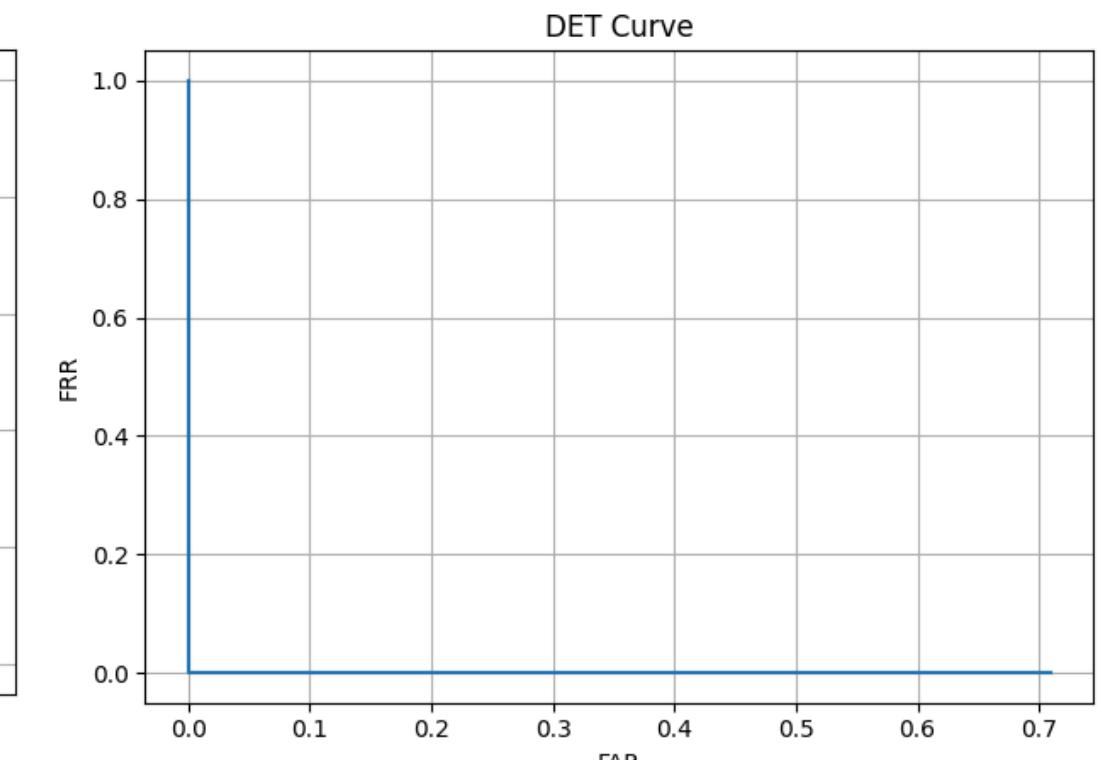
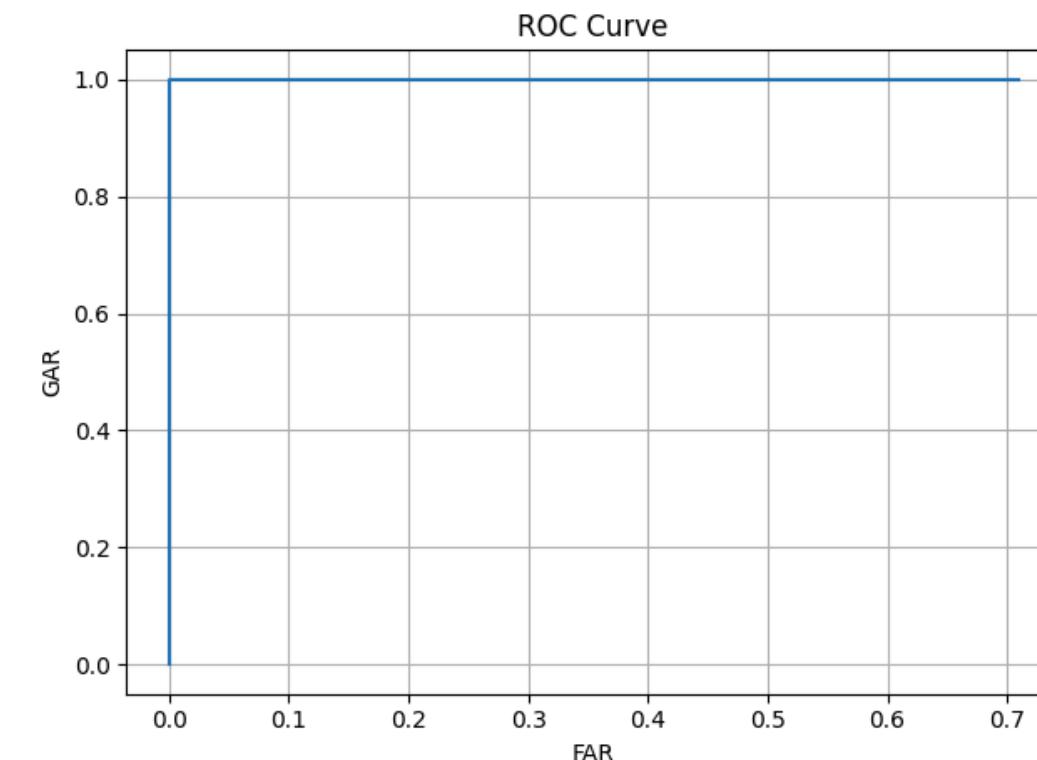
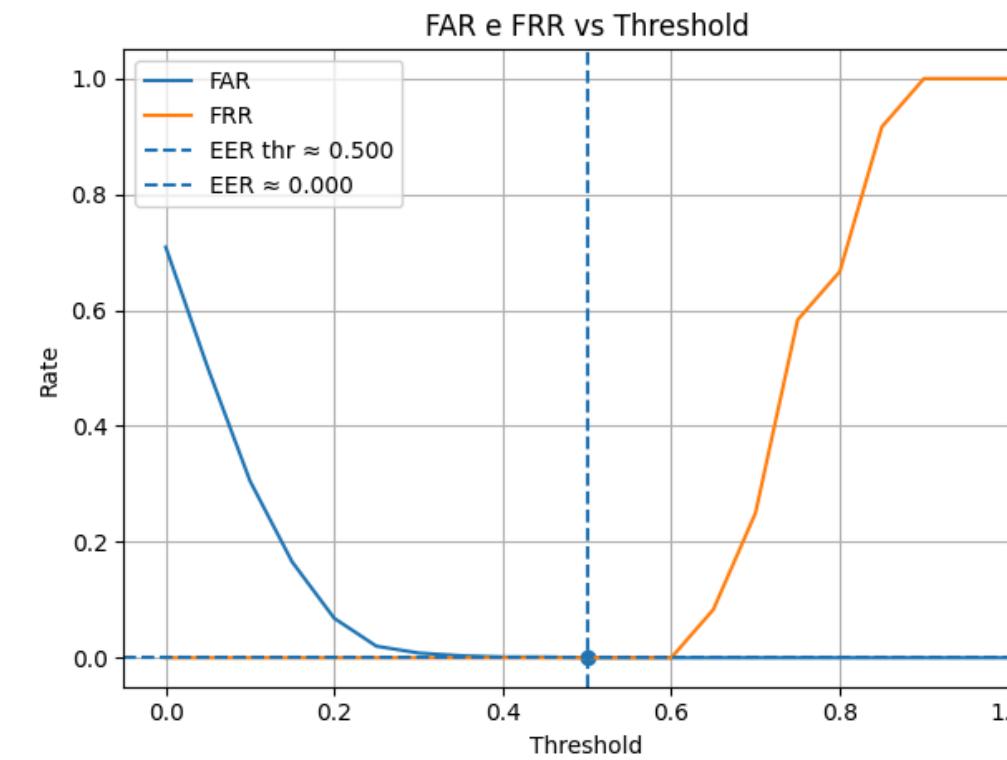
Probe
(Not in gallery)



Impostor match
Sim score: 0.6659

Results (7)

Print attack



Genuine acceptances:



Best Match: Leonardo Di Caprio

Similarity score: 0.8629

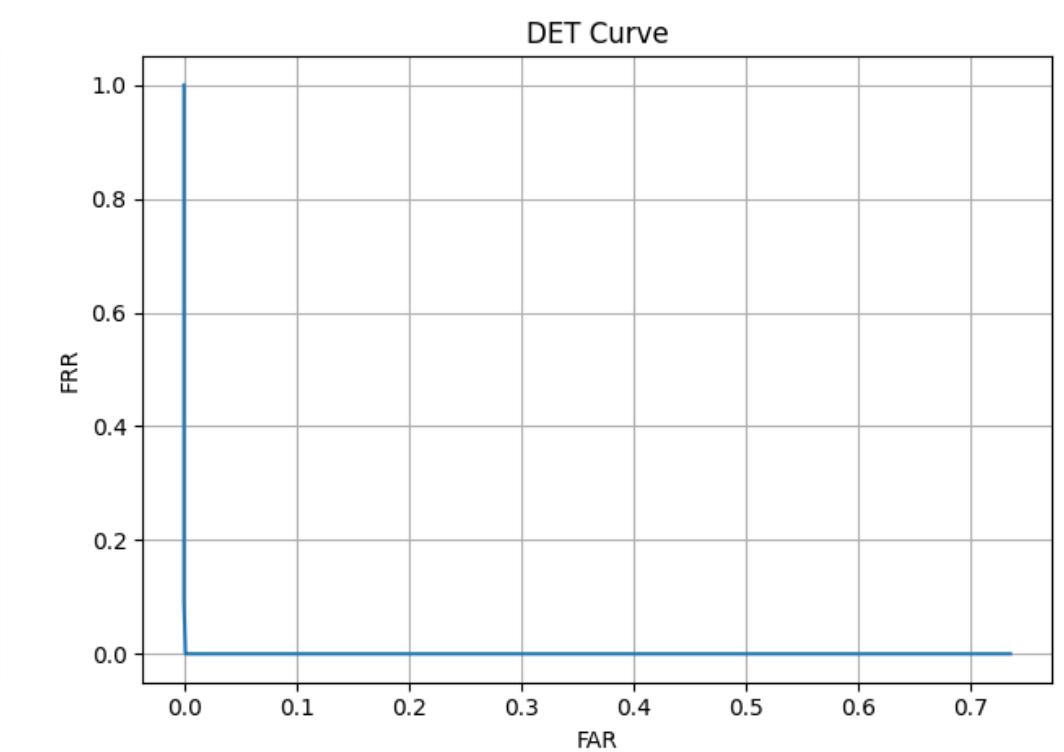
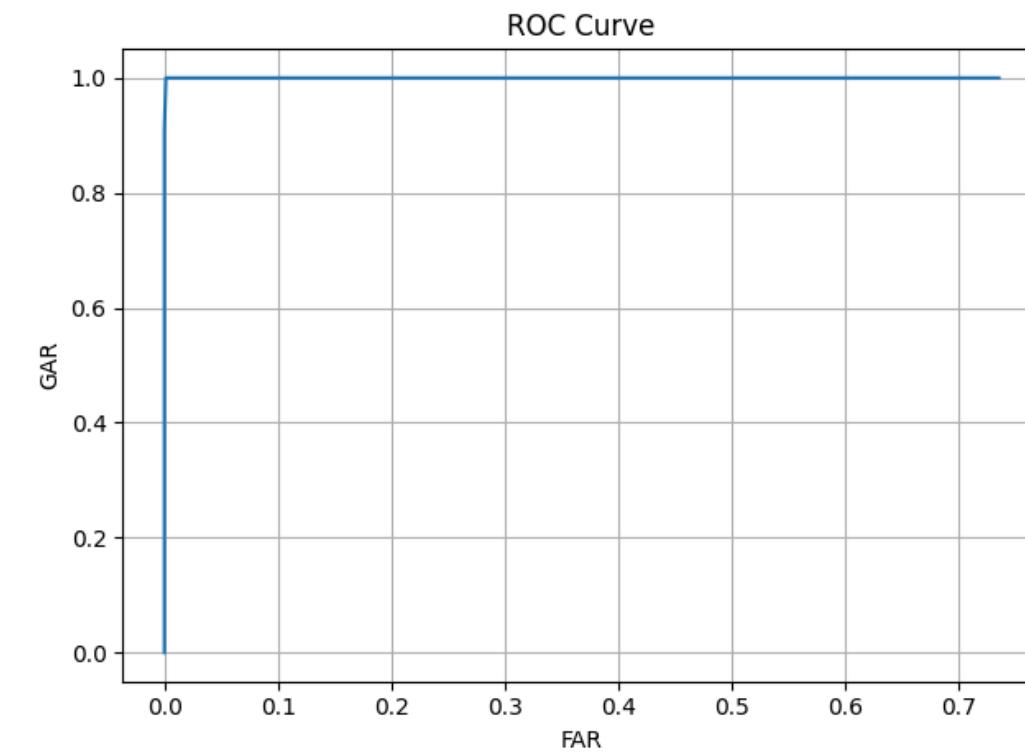
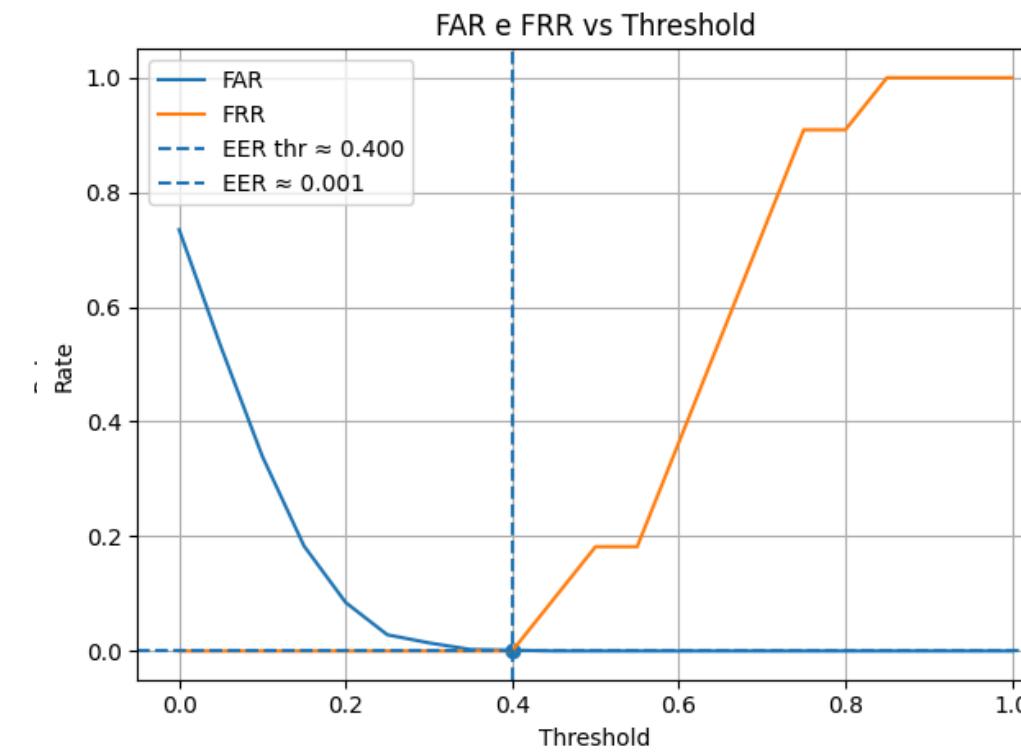


Worst Match: Johnny Depp

Similarity score: 0.4295

Results (8)

Deep fake



Genuine acceptances:



Best Match: Jennifer Lawrence

Similarity score: 0.8629

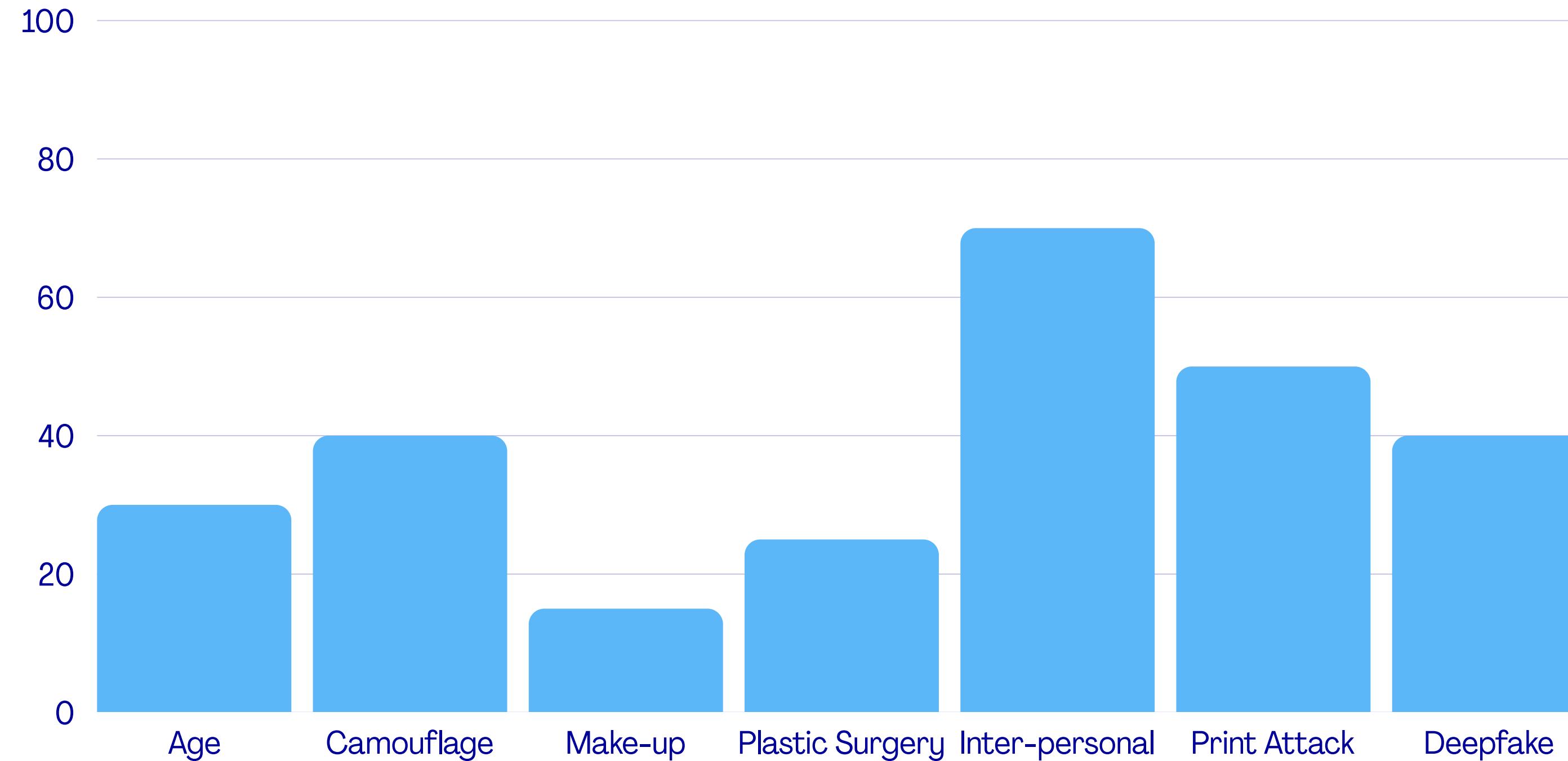


Worst Match: Johnny Depp

Similarity score: 0.4295

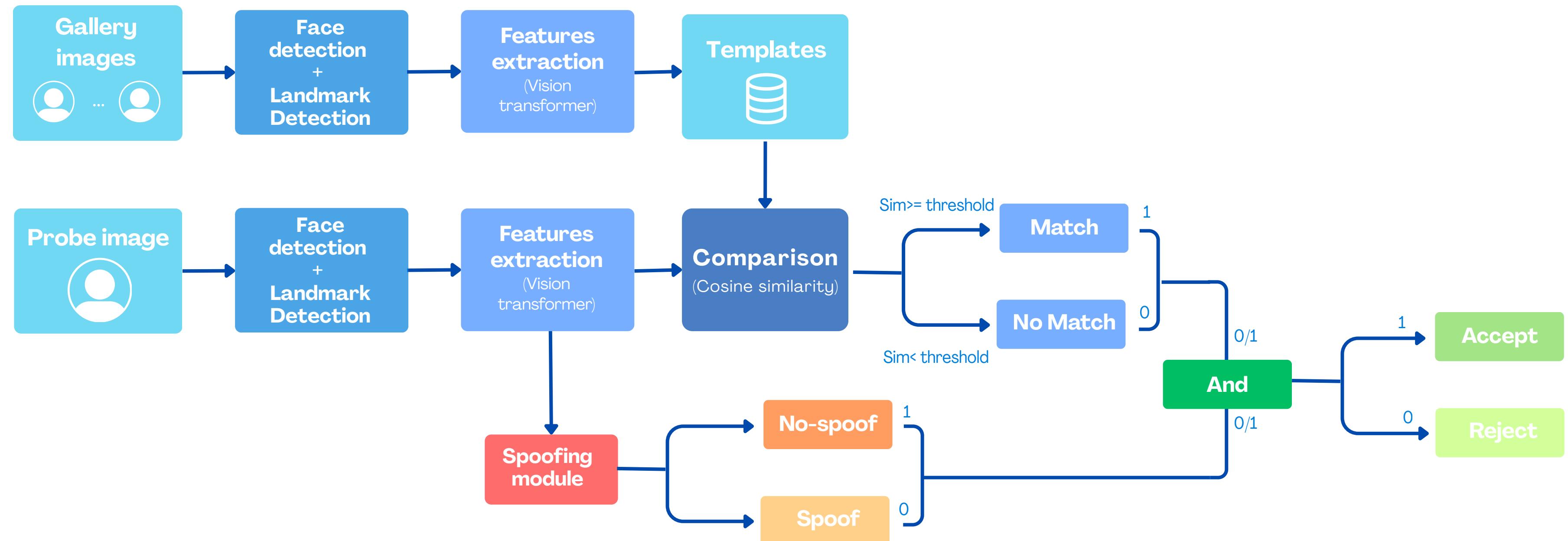
Future works

Adaptive Thresholds



Future works

What's next?



Conclusions

- ViT + ArcFace achieves **high face verification accuracy**
- Strong performance under natural variations
- **Performance decreases** under **extreme changes**
- **Spoofing attacks** are **not rejected** → require dedicated modules

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**Thank you
for your attention!**

