

Altered Biometric Data

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Biometrics

- Automated **recognition** of individuals based on their **biological** and **behavioral** characteristics
- Traits from which **distinguishing, repeatable** features can be extracted

C. G. Brown

Height	1m 71.6	Head l'gth	19.8	L. Foot	27.1	Circle	Leh	Age	22	Born in	
Eng. H'ght	5-10 3/4	Head width	16.3	L. Mid. F.	11.9	Periph Z		Apparent Age			
Outs. A	1m 75.5	Cheek width	14.4	L. Lt. F.	8.7	Color of Left Eye	Leh - Mel	Nativity	<i>Livingstone, Ky.</i>		
Trunk	94.9	R. Ear	6.8	L. Fore A.	46.6	Pecul		Occupation	<i>Johnson</i>		

Remarks Incident to Measurement

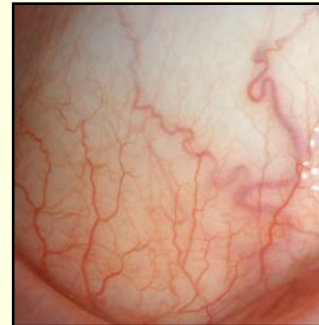
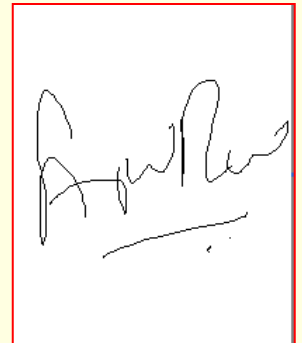
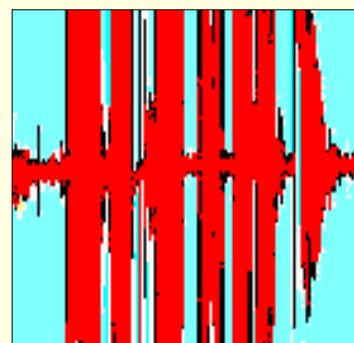
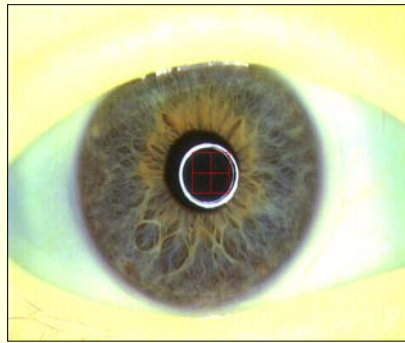
DESCRIPTIVE

Incl.	<i>Reddy</i>	Ridge	<i>None</i>	R. Ear		Beard	<i>Shaved</i>
Height	<i>M</i>	Base	<i>(Ear)</i>	Teeth	<i>Upper front</i>	Hair	<i>Black</i>
Width	<i>Brn</i>	DIMENSIONS			Complexion	<i>M. Dark</i>	
Pecul		Length	<i>Er</i>	Projection	<i>Er</i>	Weight	<i>165</i>
		Breadth	<i>M</i>	Chin	<i>M. Prom</i>	Build	<i>M. Slim</i>

BUREAU OF IDENTIFICATION
Department of Police,
Tulane Ave. and Saratoga St.
New Orleans, La.

Measured *Feb 1 1912*
By *J. G. Stone*

Biometric Traits



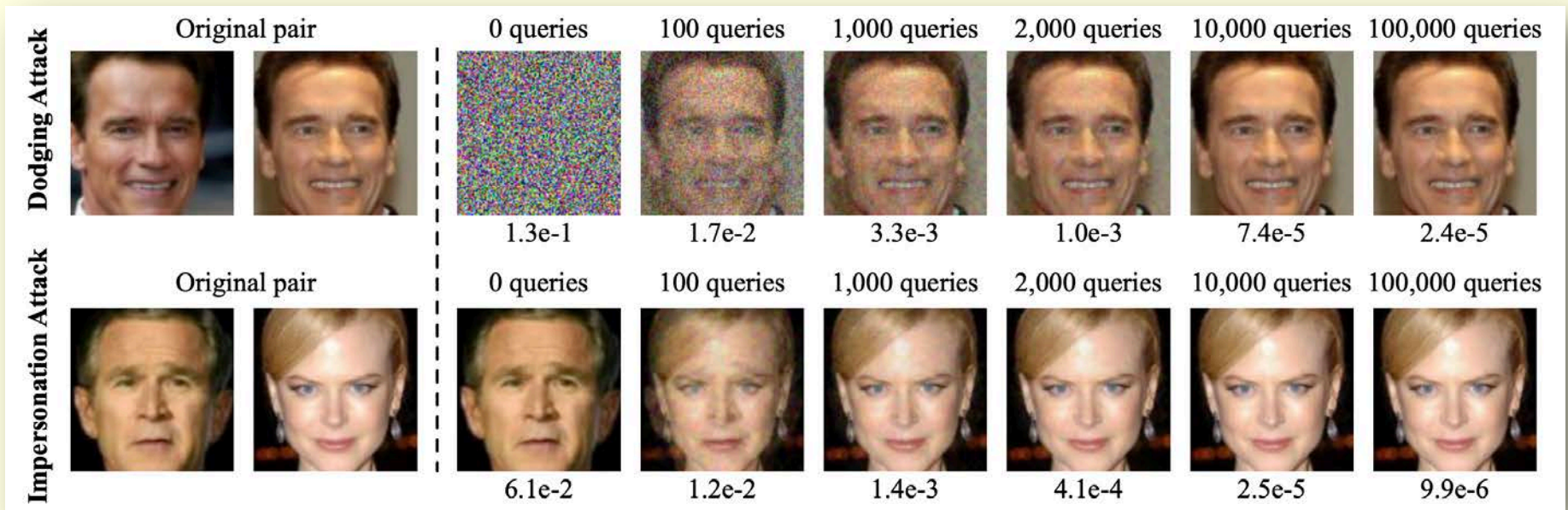
Biometric Matching/Comparison

- Given **two biometric samples**, estimate two numbers:
 - the likelihood that they are of the **same person**
 - the likelihood that they are of **different people**



Altered Data: Blackbox Attacks

■ Black-box Attacks



Dong et al, "Efficient decision-based black-box adversarial attacks on face recognition", CVPR 2019

Altered Data: Physical Attacks

- Presentation attacks: face masks
- 3D printed glasses: for dodging and impersonating others
 - Sharif et al., "Accessorize to a crime: Real and stealthy attacks on state-of-the-art face recognition, 2016
- Adversarial patches printed on T-Shirts to evade detectors
 - Thys et al., "Fooling automated surveillance cameras: Adversarial patches to attack person detection," CVPRW 2019



Real-world Challenges

Motivation – Why is the focus on biometric images?

- Widespread use of **Photoshop** and **Snapchat** filter on face images
- **Deep learning**-based manipulations are increasingly prevalent (attribute modifications, makeup transfer)

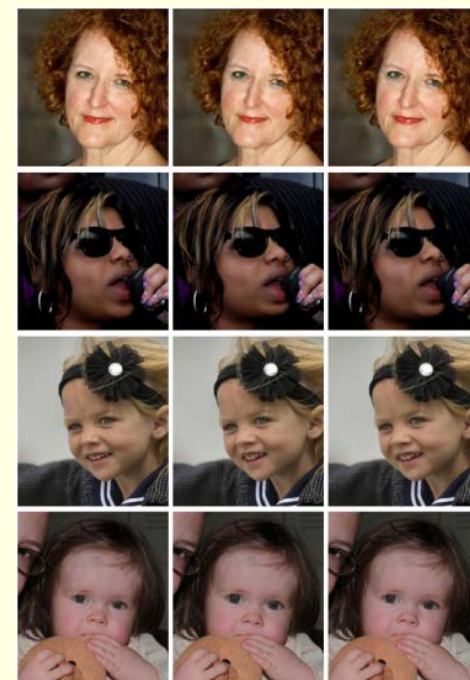
Media forensics



<https://www.hindawi.com/journals/tswj/2013/795408/>



<https://arxiv.org/pdf/1711.10678.pdf>



<https://neurohive.io/en/news/adobe-trained-a-neural-network-that-detects-photoshopped-faces/>

Image Forensics

- **Origin:** Which sensor produced this image?
- **Altered:** Is this an altered image?
- **Relationship:** How are these images related?

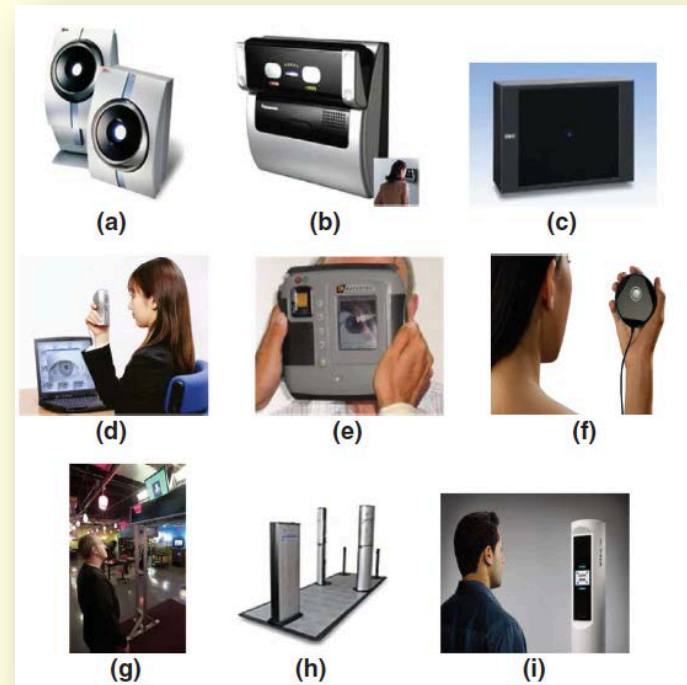


From Image to Sensor

IMAGE



SENSORS



Unit Level versus Brand Level

Sensor “Noise”

- Pixel Non-Uniformity:
 - Shot noise (random component)
 - Fixed Pattern noise (deterministic component)
- Fixed Pattern noise:
 - Dark-signal non-uniformity (DSNU)
 - Photo-response non-uniformity (PRNU)

General Approach

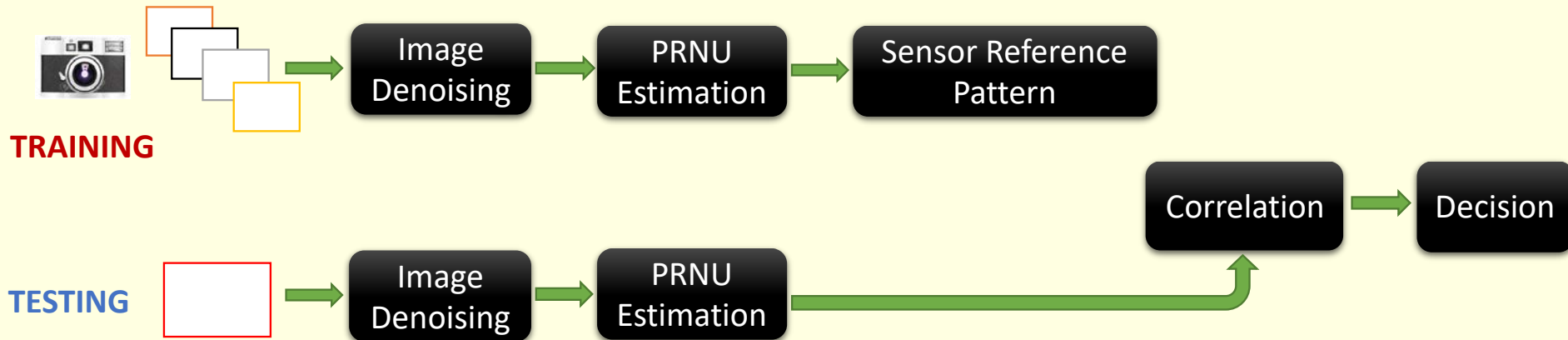


Image Denoising : Extract the PRNU, w_i from an image I_i using a **denoising** filter, $F(\cdot)$ to suppress scene influences

$$w_i = I_i - F(I_i)$$

$F(\cdot)$ can be wavelet-based filter

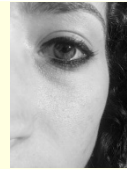
Sensor De-identification

Original Image

PRNU Spoofed Image



IP5 1 FRONT



Galaxy S4 FRONT



Galaxy S4 REAR



IP5 2 REAR



IP5 2 FRONT



IP5 1 FRONT



Galaxy S4 FRONT



IP5 1 FRONT

Source Sensor

Target Sensor



Digital Data: DeepFakes

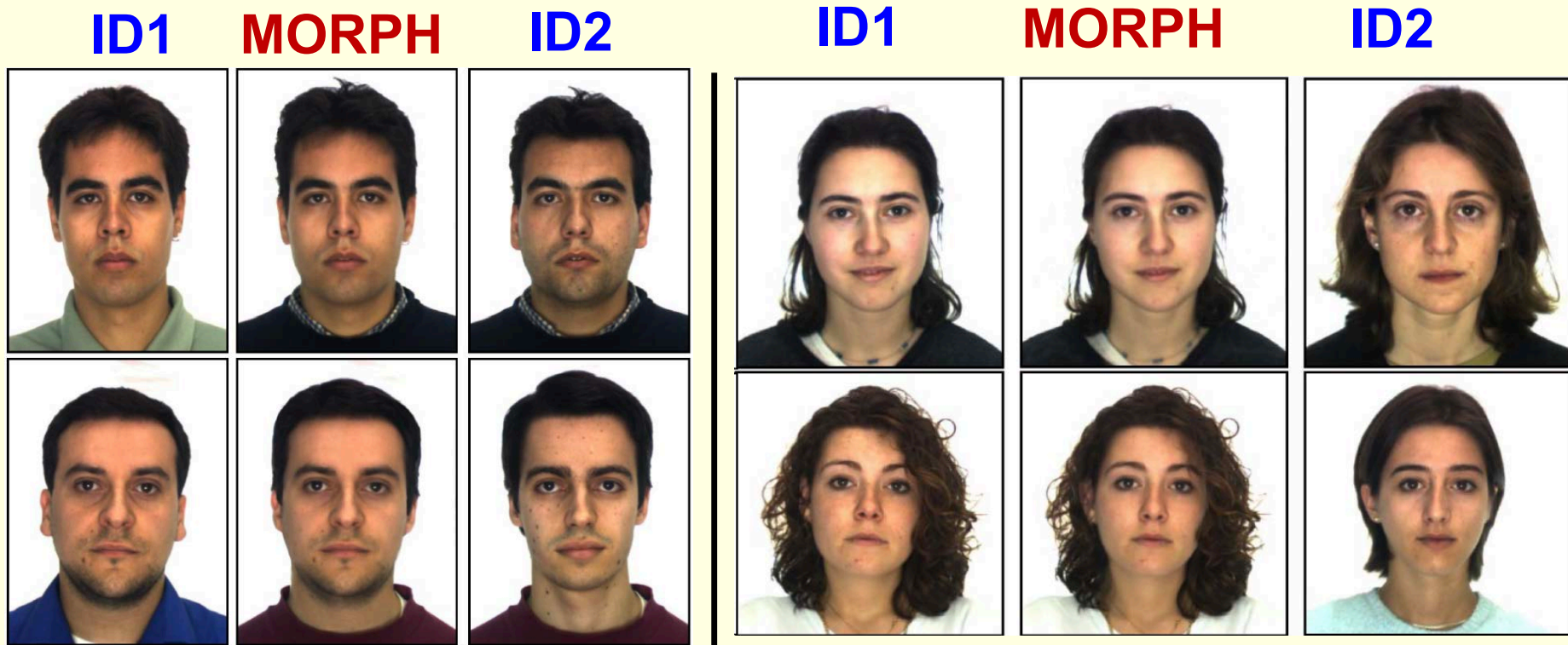
- **DeepFakes:** Synthetically Generated Images



<https://thispersondoesnotexist.com/>

Digital Data: Morphed Faces

- **Morphed Faces:** Combining two face images



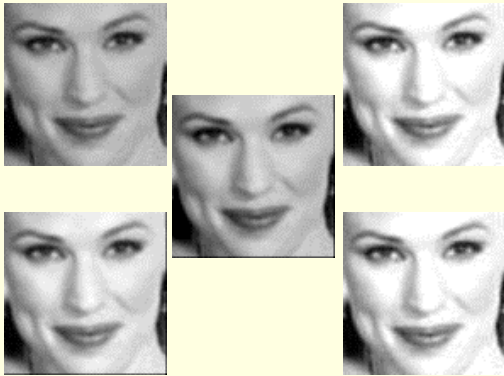
Ferrara et al, "The Magic Passport," IJCB 2014

Also see, Othman and Ross,
"Privacy of Facial Soft Biometrics: Suppressing Gender But Retaining Identity," ECCVW 2014

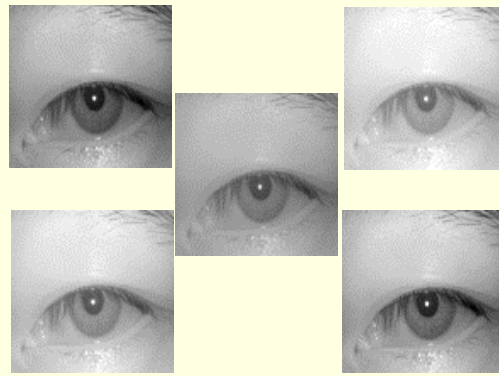
Digital Data: Near Duplicates

- **Near Duplicates:** Subtly Modified Images

Brightness adjustment



Gamma transformation

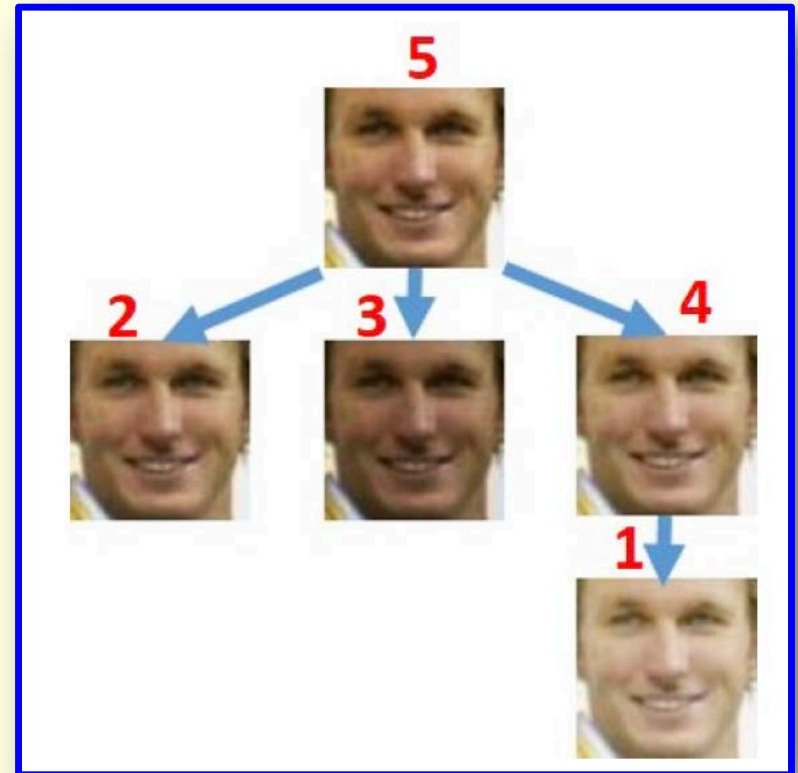
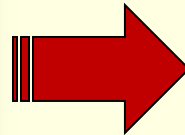
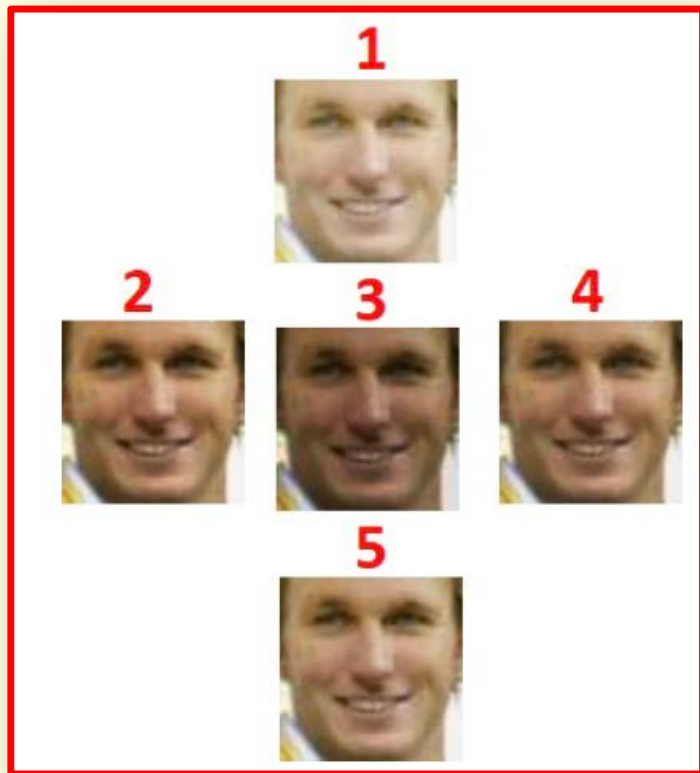


Rotation



Relationship Between Images

- **Phylogeny Tree:** Relationship between near duplicate images



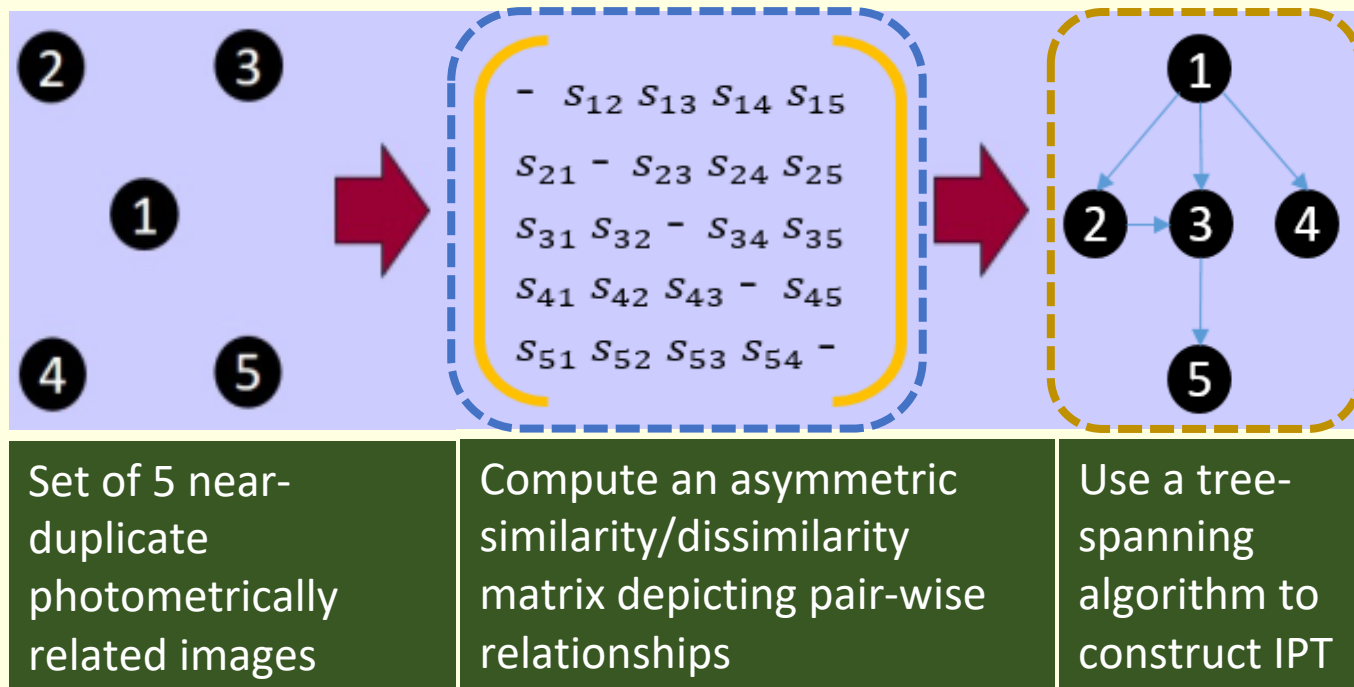
Importance of Problem

- Deduce whether a set of photometrically transformed images originated from a single source image or multiple sources
- Detection of image tampering hinted by significant photometric variation between two images
- Determination of transformation parameters relating two images

Forensics + Data Analysis

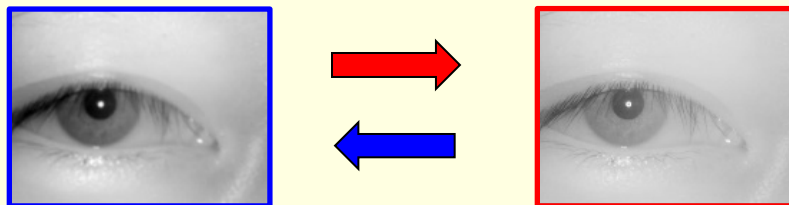
Image Phylogeny Tree (IPT)

- IPT construction is a 2-step process:
 - STEP I : Computing pairwise asymmetric measure
 - STEP II: Using a tree-spanning algorithm



What are the Challenges?

- Photometric Transformations ▶ Large number
 - E.g., Brightness, Contrast, CLAHE, Gamma, Median, Gaussian
- ↓
- Each Transformation ▶ multiple parameters
 - E.g., Gaussian: window size and variance
- ↓
- Each Parameter ▶ multiple values
 - E.g., Window size: 3x3, 5x5, 9x9, 13x13,
- Need to distinguish between $A \rightarrow B$ and $B \rightarrow A$

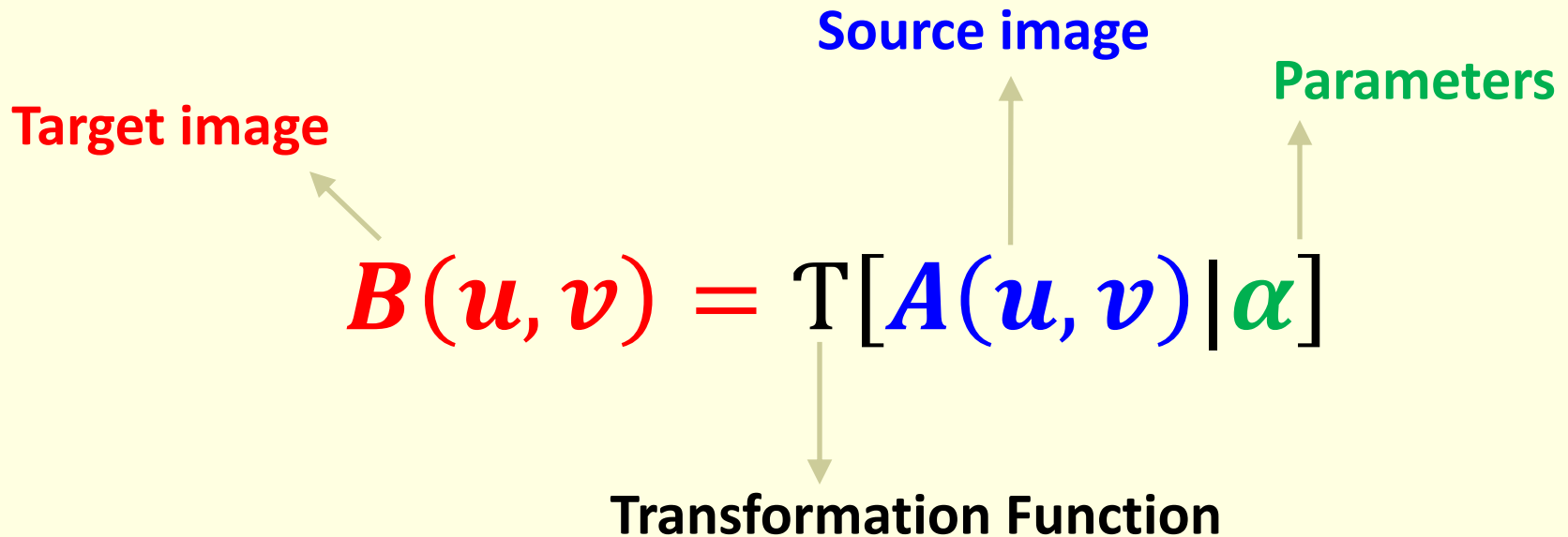


Our Approach

- Use a **generic parametric transformation function** to model the relationship between any two images
- Given two images, **A** and **B**, **estimate the parameters** of the function in both directions
- Use the **likelihood of the parameters** to determine which of the two cases is more likely, i.e., **A → B** or **B → A**

- Banerjee and Ross, "Face Phylogeny Tree: Deducing Relationships Between Near-Duplicate Face Images Using Legendre Polynomials and Radial Basis Functions," BTAS 2019
- Banerjee and Ross, "Face Phylogeny Tree Using Basis Functions," IEEE TBIOM 2020

Transformation Function



Transformation Function

- Model transformation from $A \rightarrow B$ such that the pairwise photometric error (PE) is **minimized** for all pixels p

$$\min_{\alpha} PE(A, B) = \min_{\alpha} \sum_{p=1}^N \|B(p) - \tau(A(p); \alpha)\|_2^2$$

- We approximate transformations using a set of **basis** functions

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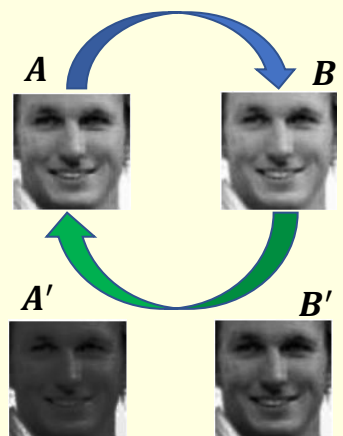
Basis Functions

Basis Functions		Utility	Formulation
Polynomials	Legendre	Used for image template matching and image reconstruction	$L_n(p) = 2^n \sum_{k=0}^n p^k \binom{n}{k} \left(\frac{n+k-1}{2} \right)$
	Chebyshev	Used for approximating complex functions (spectral convolutions)	$C_n(p) = p^n \sum_{k=0}^{\lfloor \frac{n}{2} \rfloor} \binom{n}{2k} (1 - x^{-2})^k$
Wavelets	Gabor	Used as texture descriptors, acts as bandpass filters	$\varphi(p, \theta, \lambda) = g(p, \lambda) \cdot w(p, \theta)$ $\lambda = \{2, 3, 4, 5\}; \theta = \{0^\circ, 45^\circ, 90^\circ, 135^\circ\}$
Radial Basis Functions	Gaussian	Used for interpolation	$K(p) = \exp\ p - \mu\ ^2$
	Bump	Used as smooth cutoff functions	$K(p) = \exp\left(-\frac{1}{1 - p^2}\right)$

p: Pixel intensity value ; *n*: Polynomial order ; μ : Mean pixel intensity value ; λ : Scale ; θ : Orientation

Banerjee and Ross, "Face Phylogeny Tree: Deducing Relationships Between Near-Duplicate Face Images Using Legendre Polynomials and Radial Basis Functions," BTAS 2019

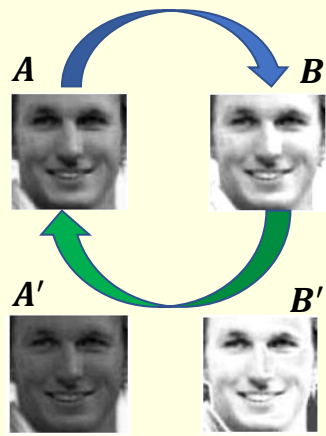
Basis Functions



Output of modeling
 $B \rightarrow A$

Output of modeling
 $A \rightarrow B$

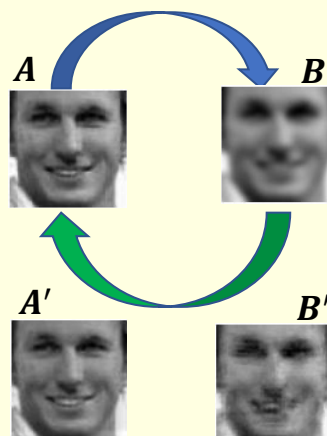
LEGENDRE



Output of modeling
 $B \rightarrow A$

Output of modeling
 $A \rightarrow B$

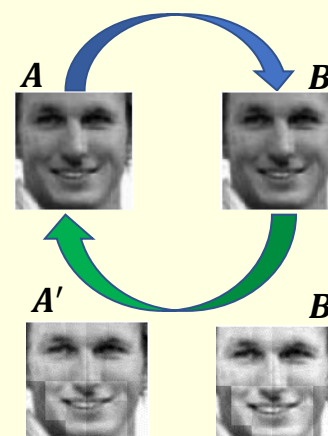
CHEBYSHEV



Output of modeling
 $B \rightarrow A$

Output of modeling
 $A \rightarrow B$

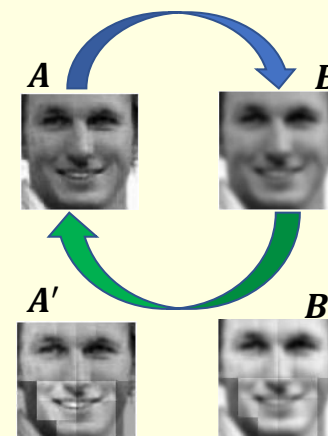
GABOR



Output of modeling
 $B \rightarrow A$

Output of modeling
 $A \rightarrow B$

GAUSSIAN RBF



Output of modeling
 $B \rightarrow A$

Output of modeling
 $A \rightarrow B$

BUMP RBF

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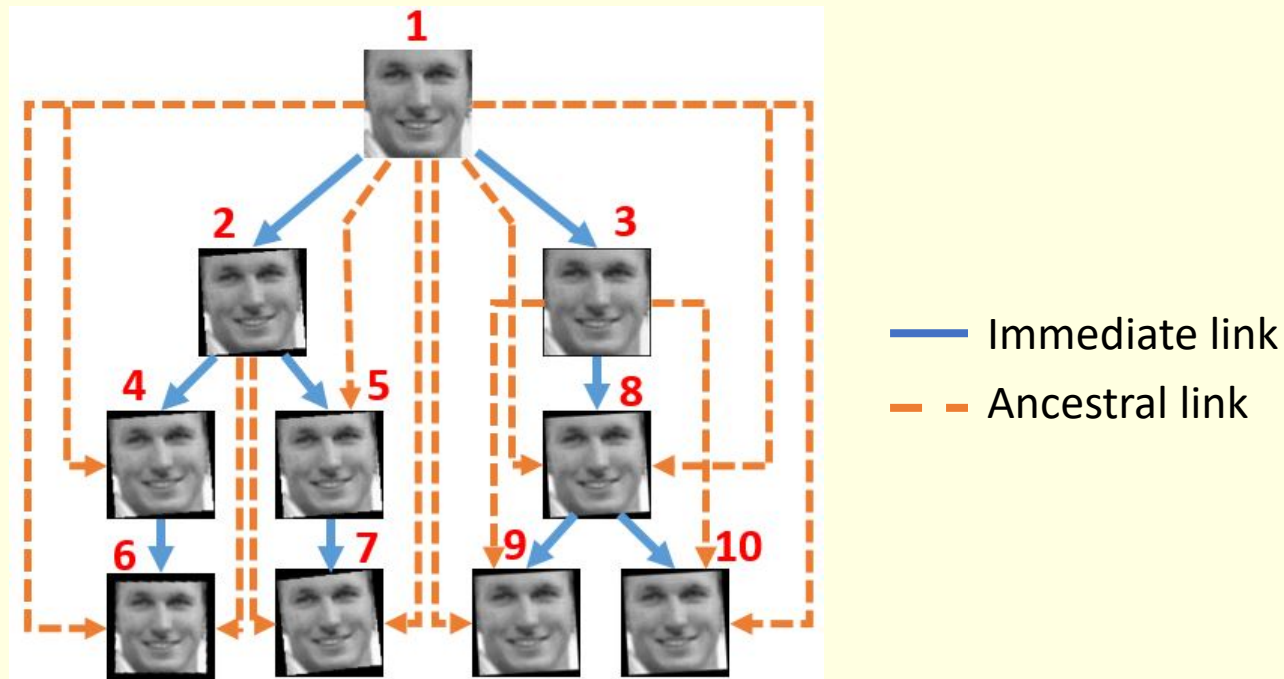
Asymmetric Measure

- Modeling the transformation in both directions results in **two estimated** parameter vectors (α, β)
- Compute the **likelihood ratio** $\left(\Lambda_{\alpha} = \frac{p_f(\alpha)}{p_r(\alpha)}, \Lambda_{\beta} = \frac{p_f(\beta)}{p_r(\beta)}\right)$ of the estimated parameters to obtain asymmetric measure
- Use **depth first search** to construct IPT

- Banerjee and Ross, "Face Phylogeny Tree: Deducing Relationships Between Near-Duplicate Face Images Using Legendre Polynomials and Radial Basis Functions," BTAS 2019
 - Banerjee and Ross, "Face Phylogeny Tree Using Basis Functions," IEEE TBIOM 2020

Experiments

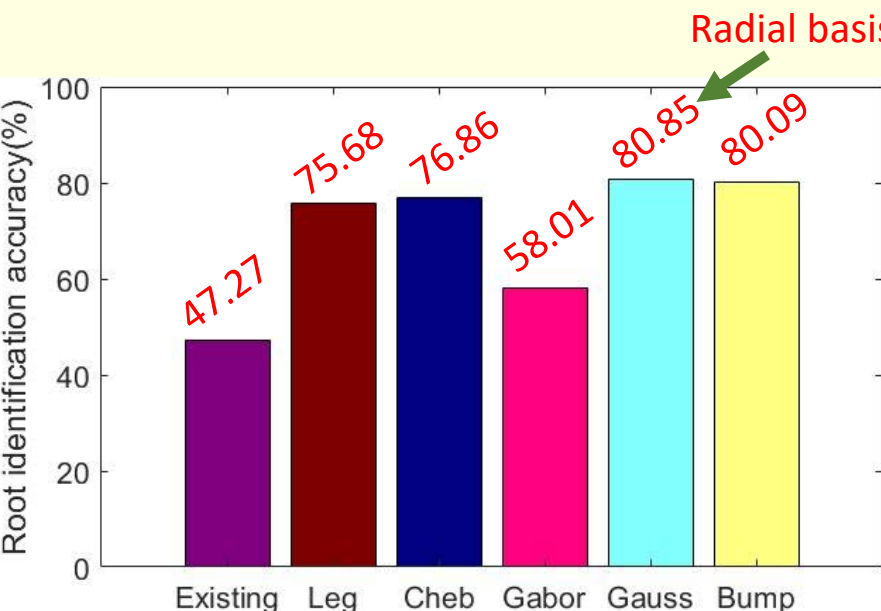
- We generated 2,727 IPTs by subjecting face images to random sequence of 4 transformations resulting in 27,270 images



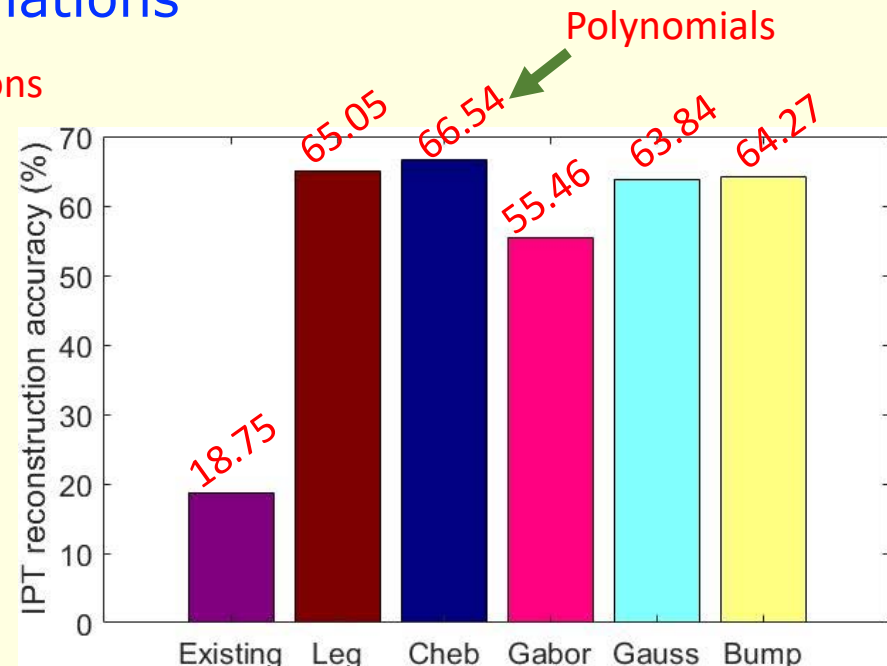
- Banerjee and Ross, "Face Phylogeny Tree: Deducing Relationships Between Near-Duplicate Face Images Using Legendre Polynomials and Radial Basis Functions," BTAS 2019
 - Banerjee and Ross, "Face Phylogeny Tree Using Basis Functions," IEEE TBIOM 2020

Performance

- We compared the performance with an existing method
- The problem is hard:
 - face images vs natural scenes
 - subtle photometric transformations



ROOT IDENTIFICATION



IPT RECONSTRUCTION

Generalizability

- **Unseen modalities:** 7,260 near-duplicate iris images from CASIA Iris V2 Device 2 dataset
- **Unseen transformations:** 175 near-duplicates using Photoshop and 1,080 near-duplicates using deep learning-based transformations

Experimental Settings		Root identification accuracy (%)	IPT reconstruction accuracy (%)
Unseen modality	Iris	95	68
Unseen transformations	Photoshop	90	100
	Deep learning-based	83	65

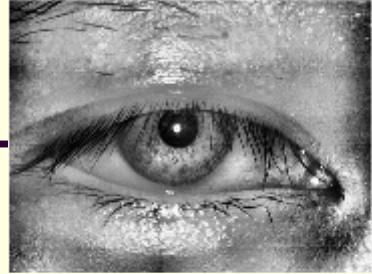
- Banerjee and Ross, "Face Phylogeny Tree: Deducing Relationships Between Near-Duplicate Face Images Using Legendre Polynomials and Radial Basis Functions," BTAS 2019
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Summary

- Information can be derived from a single biometric image
 - Demographic attributes
 - Environmental characteristics
 - Sensor properties; etc.
- Digital Image Forensics:
 - Which sensor did this image come from?
 - Has this image been digitally tampered?
 - What is the relationship between a set of near-duplicate images?
- Privacy:
 - Semi-adversarial Networks (SAN): Differential Privacy

Digital Image Forensics

- ❑ El-Naggar, Ross, "**Which Dataset is this Iris Image From?**," WIFS 2015
- ❑ Kalka, Bartlow, Cukic, Ross, "**A Preliminary Study on Identifying Sensors from Iris Images**," CVPRW 2015
- ❑ Banerjee, Ross, "**From Image to Sensor: Comparative Evaluation of Multiple PRNU Estimation Schemes for Identifying Sensors from NIR Iris Images**," IWBF 2017
- ❑ Banerjee, Ross, "**Computing an Image Phylogeny Tree from Photometrically Modified Iris Images**," IJCB 2017
- ❑ Banerjee, Ross, "**Impact of Photometric Transformations on PRNU Estimation Schemes: A Case Study Using Near Infrared Ocular Images**," IWBF 2018
- ❑ Banerjee, Mirjalili, Ross, "**Spoofing PRNU Patterns of Iris Sensors while Preserving Iris Recognition**," ISBA 2019
- ❑ Banerjee, Ross, "**Smartphone Camera De-identification while Preserving Biometric Utility**," BTAS 2019
- ❑ Banerjee, Ross, "**Face Phylogeny Tree: Deducing Relationships Between Near-Duplicate Face Images Using Legendre Polynomials and Radial Basis Functions**," BTAS 2019
- ❑ Banerjee, Ross, "**Face Phylogeny Tree Using Basis Functions**," IEEE TBIOM 2020



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