Transfer Learning based Evolutionary Algorithm for Face Sketch Recognition

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Pose, Illumination, Expression



Heterogeneity: Resolution, Spectrum, Sketch

















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Skelch to Digital Photo Matching



 Sketches are important in Law enforcement applications

Different types of Sketches



Viewed vs Semi-Forensics Matching Scenario



IIITD Composite Face Sketch Dataset

- a 150 subjects
- 150 composite sketches and 300
 digital face images
- One image used to create sketch
 image, the other used for matching
- Sketches are created using FACES
 software
- Dataset will be made publicly available for research







Digital Images



Sketches











Experiment	Source Domain	Pairs in Training	Target Domain
1	Digital Image	250	Composite Sketch Images (Proposed IIITD Composite Face Sketch Database)
2	Viewed Hand Drawn Sketch	482	
3	Semi-Forensic Hand-Drawn Sketch	106	
4	Forensic Hand-Drawn Sketch	190	
5	Composite Sketch	25	

Proposed IIITD Composite Face Sketch CMU Multi-PIE (Gross et al.) IIIT-Delhi Sketch (Bhatt et al.)





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Best performing Results: when source domain is semi-forensic sketches

Key Observalions

- Rank-10 identification accuracy improves
 by 5-10% after applying the transfer
 learning technique
- Training with forensic sketches yields
 Lower accuracy than semi-forensic sketches
- However, Rank-1 (even Rank-10) accuracy
 is very low

Next (Fulture) Step

 Try representation-learning with domain adaptation (transfer learning) for very small sample size

S. Nagpal, M. Singh, R. Singh, M. Vatsa, A. Noore, and A. Majumdar, Face Sketch Matching via Coupled Deep Transform Learning, International Conference on Computer Vision, 2017

Thank Youll