

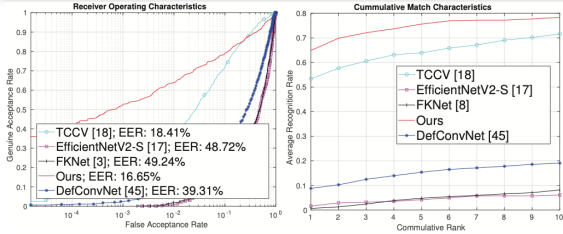
Towards Explainable and Unprecedented Accuracy in Matching Challenging Finger Crease Patterns

Zhenyu Zhou, Chengdong Dong, Ajay Kumar

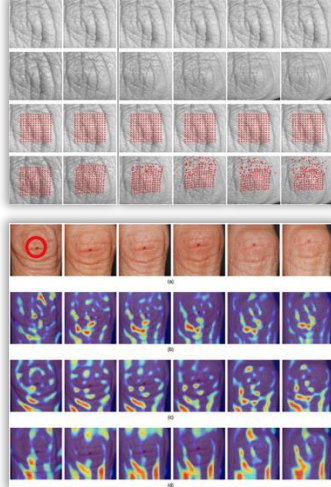
Department of Data Science and Artificial Intelligence, The Hong Kong Polytechnic University

Problems:

1). Poor Cross-Pose Performance

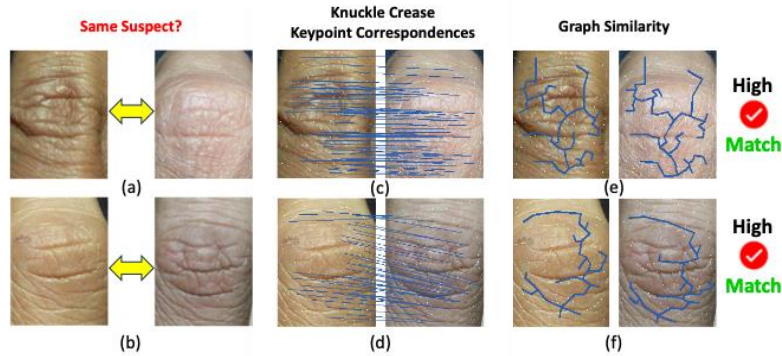


2). Consolidate Match Evidence



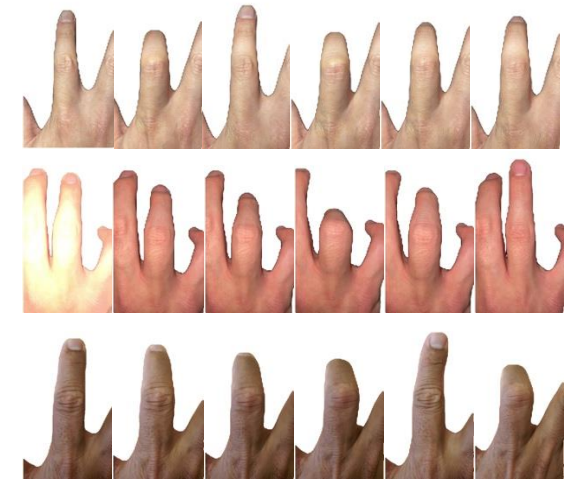
Solution:

Explainable and Unprecedented Accuracy

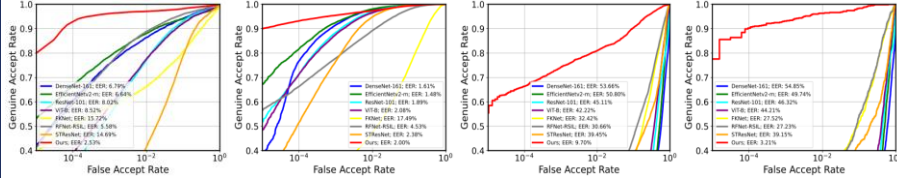


Contribution:

1). A Multi-Pose Finger Knuckle Dataset



Within-Database and Cross-Database Experiment



Estimation of Finger Knuckle Uniqueness Experiment

Δ_k	Δ_f	(m, n, w, α)	$p(T^P, T^S)$	$\bar{\lambda}_\alpha$	FRC_α
6.750	0.726	(69,69,15,0.05)	2.980×10^{-4}	6.167	6.780×10^{-4}
6.750	0.726	(69,69,17,0.05)	2.980×10^{-4}	6.613	1.860×10^{-4}
6.750	0.726	(69,69,19,0.05)	2.980×10^{-4}	6.996	4.408×10^{-5}
6.750	0.726	(69,69,21,0.05)	2.980×10^{-4}	7.325	9.055×10^{-6}
3.750	0.726	(69,69,17,0.05)	1.031×10^{-4}	2.962	2.948×10^{-9}
5.250	0.726	(69,69,17,0.05)	1.914×10^{-4}	4.950	4.731×10^{-6}
8.250	0.726	(69,69,17,0.05)	4.119×10^{-3}	8.061	1.729×10^{-3}
9.750	0.726	(69,69,17,0.05)	5.273×10^{-4}	9.124	6.080×10^{-3}
6.750	0.626	(69,69,17,0.05)	6.210×10^{-6}	N/A	N/A
6.750	0.676	(69,69,17,0.05)	5.309×10^{-5}	1.367	1.546×10^{-14}
6.750	0.776	(69,69,17,0.05)	1.194×10^{-3}	12.476	8.305×10^{-2}
6.750	0.826	(69,69,17,0.05)	3.620×10^{-3}	14.893	2.421×10^{-1}

3). Uniqueness of 2D Finger Knuckle Patterns

- Joint Mixture Gaussian Distribution

$$f(k|f|\Omega) = \sum_{i=1}^N \epsilon_n f_n^K(k|\mu_n^{(1)}, \Sigma_n^{(1)}) \cdot f_n^F(f|\mu_n^{(2)}, \Sigma_n^{(2)})$$

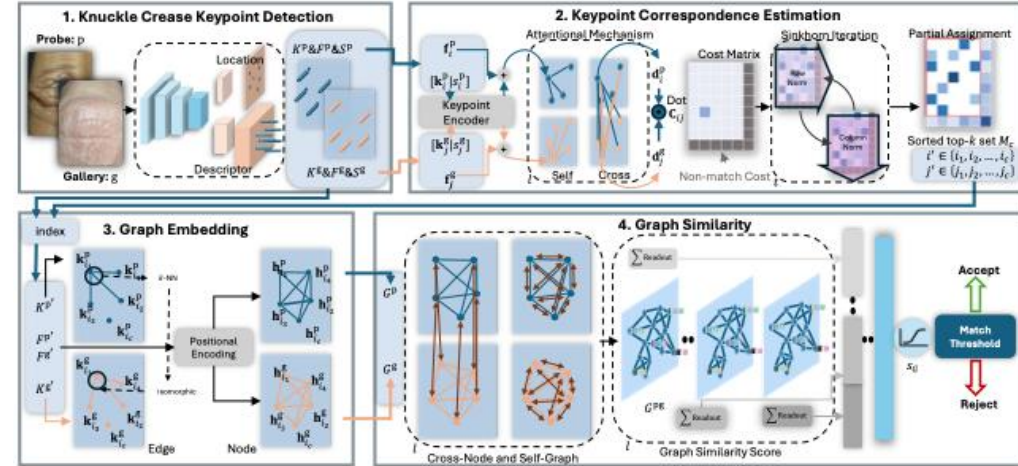
- Calculate of \overline{FRC} Index

$$p(w; T^P, T^S) = \sum_w \frac{e^{-\lambda(T^P, T^S)} \lambda(T^P, T^S)^w}{w!}$$

$$p(T^P, T^S) = \int_{(k^P, f^P) \in B((k^S, f^S))} f_{T^P}((k^P, f^P)) f_{T^S}((k^S, f^S)) dk^P df^P dk^S df^S$$

$$\overline{FRC} = \frac{2}{z(z-1)} \sum_{T^P \neq T^S} p(w; T^P, T^S)$$

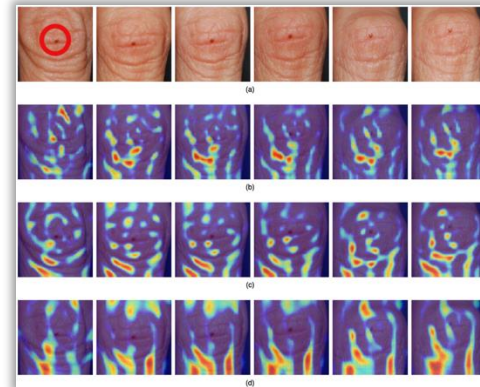
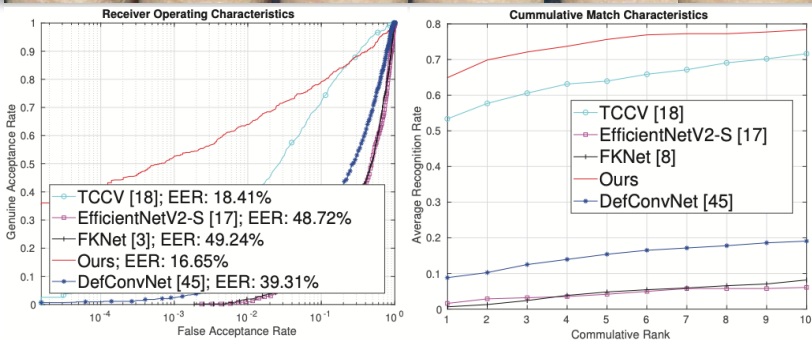
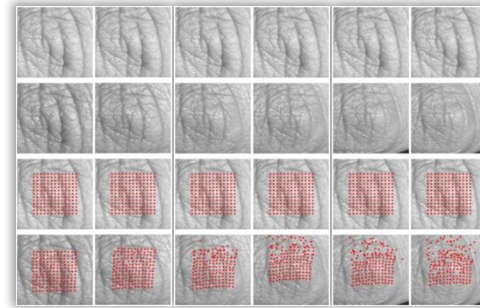
2). Our Framework to Match Cross-Pose Images



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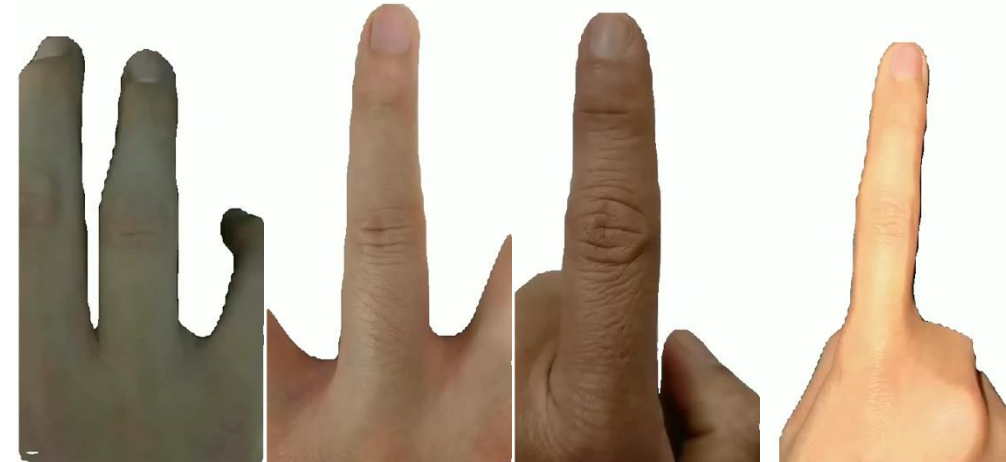
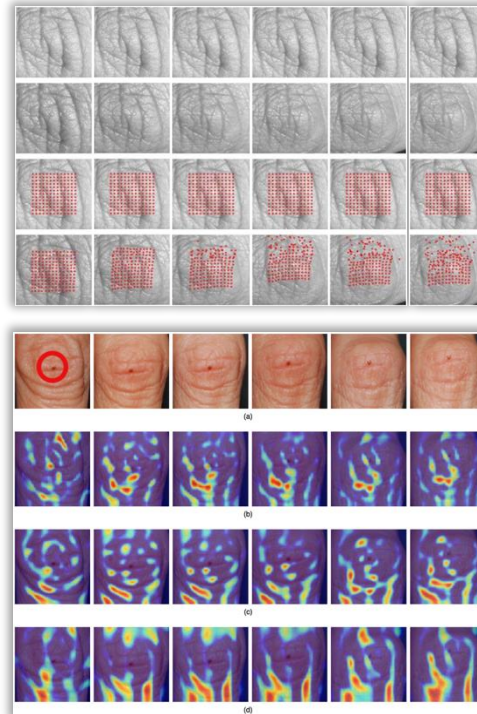
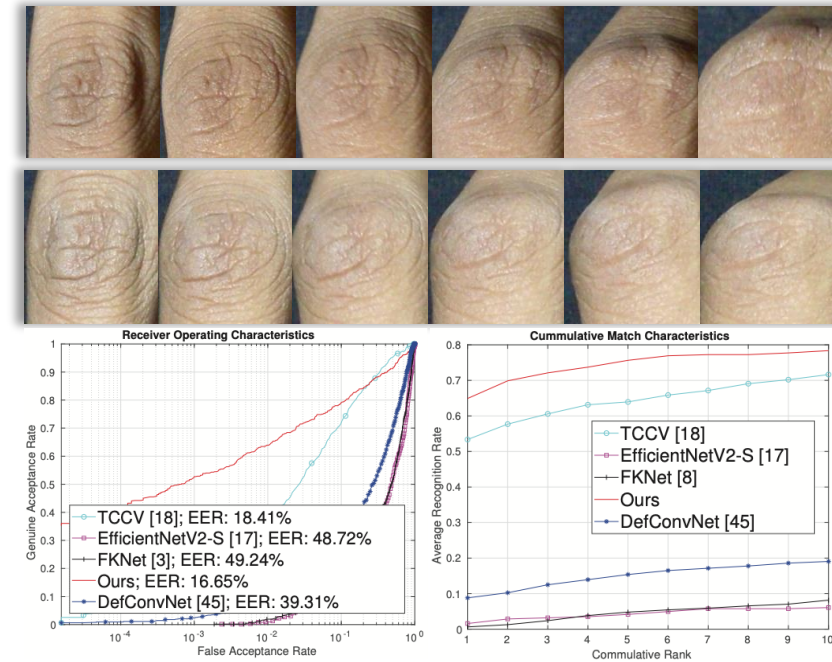
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Problem Definition:

- How to **accurately** match the cross-pose finger crease pattern?
- How to **efficiently** give the match evidences?



Multi-Pose Dataset

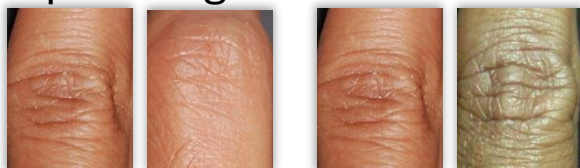
Samples from our collected the large-scale multi-pose finger knuckle dataset.

Matching Framework: Explainable and Unprecedented Accuracy

Genuine

Imposter

0). Input Image Pairs



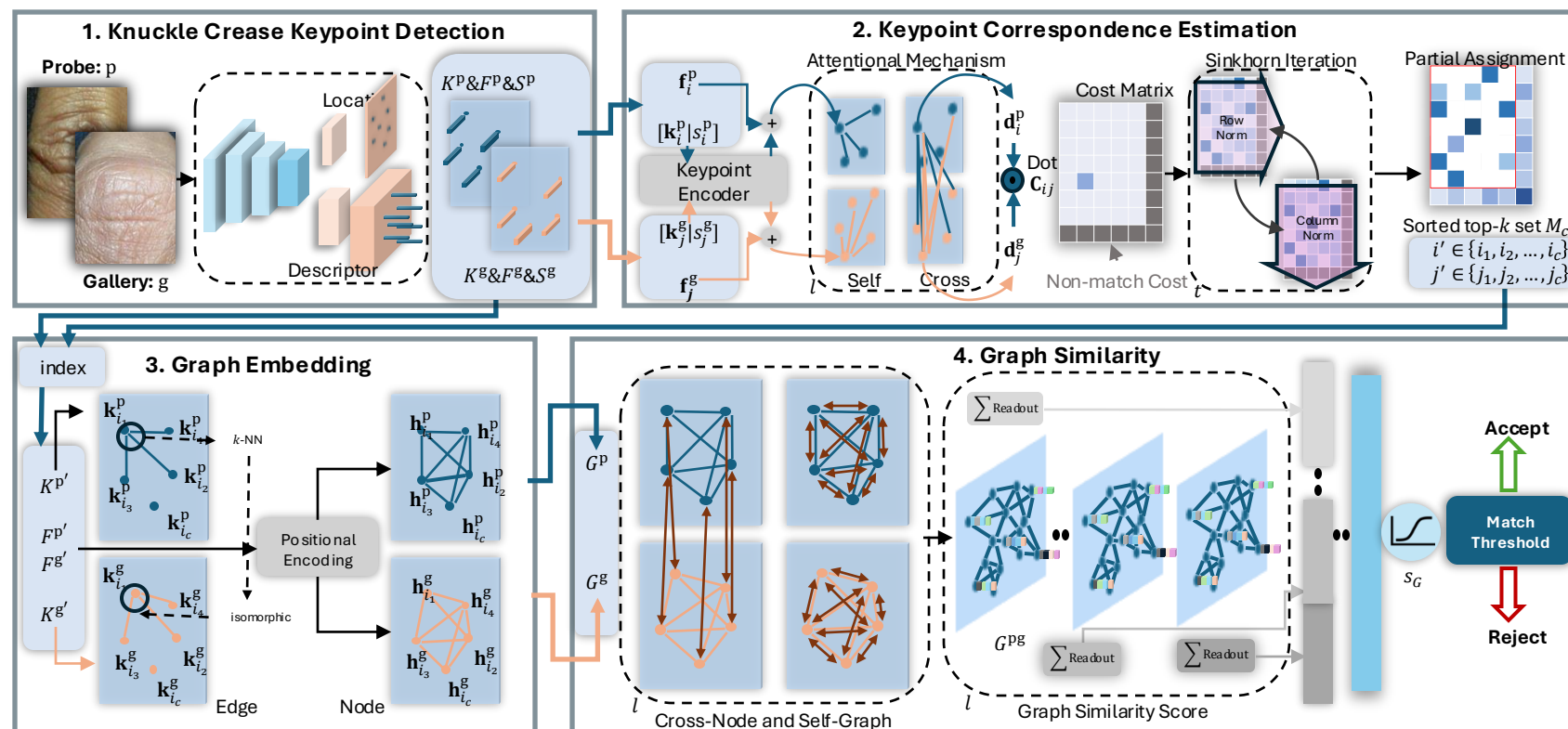
1). Keypoint Detection



2). Correspondence Estimation



3). Graph Embedding & Matching

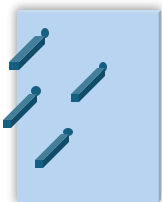


Theoretical Explainability: Uniqueness of Contactless 2D Finger Knuckle Patterns

1). Detected a pair of template T^p and T^g :

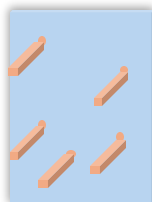
$$F^p = \{f_i^p \mid i \in \{1, 2, \dots, n\}\}$$

$$K^p = \{k_i^p \mid i \in \{1, 2, \dots, n\}\}$$



$$F^g = \{f_i^g \mid i \in \{1, 2, \dots, m\}\}$$

$$K^g = \{k_i^g \mid i \in \{1, 2, \dots, m\}\}$$



where, $k = (x, y) \in K$ and $f = (f_1, f_2, \dots, f_d) \in F$

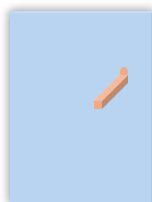
2). Joint Mixture Gaussian Distribution

$$f(k, f \mid \Omega) = \sum_{n=1}^N \epsilon_n f_n^K(k \mid \mu_n^{(1)}, \Sigma_n^{(1)}) \cdot f_n^F(f \mid \mu_n^{(2)}, \Sigma_n^{(2)})$$

4). Calculate of \overline{FRC} Index

$$p(T^p, T^g) = \int_{(k^p, f^p) \in B(k^g, f^g)} f_{T^p}((k^p, f^p)) f_{T^g}((k^g, f^g)) dk^p df^p dk^g df^g$$

3). $p(T^p, T^g)$:



$$\lambda(T^p, T^g) = m * n * p(T^p, T^g)$$

$$p(w; T^p, T^g) = \sum_w \frac{e^{-\lambda(T^p, T^g)} \lambda(T^p, T^g)^w}{w!}$$

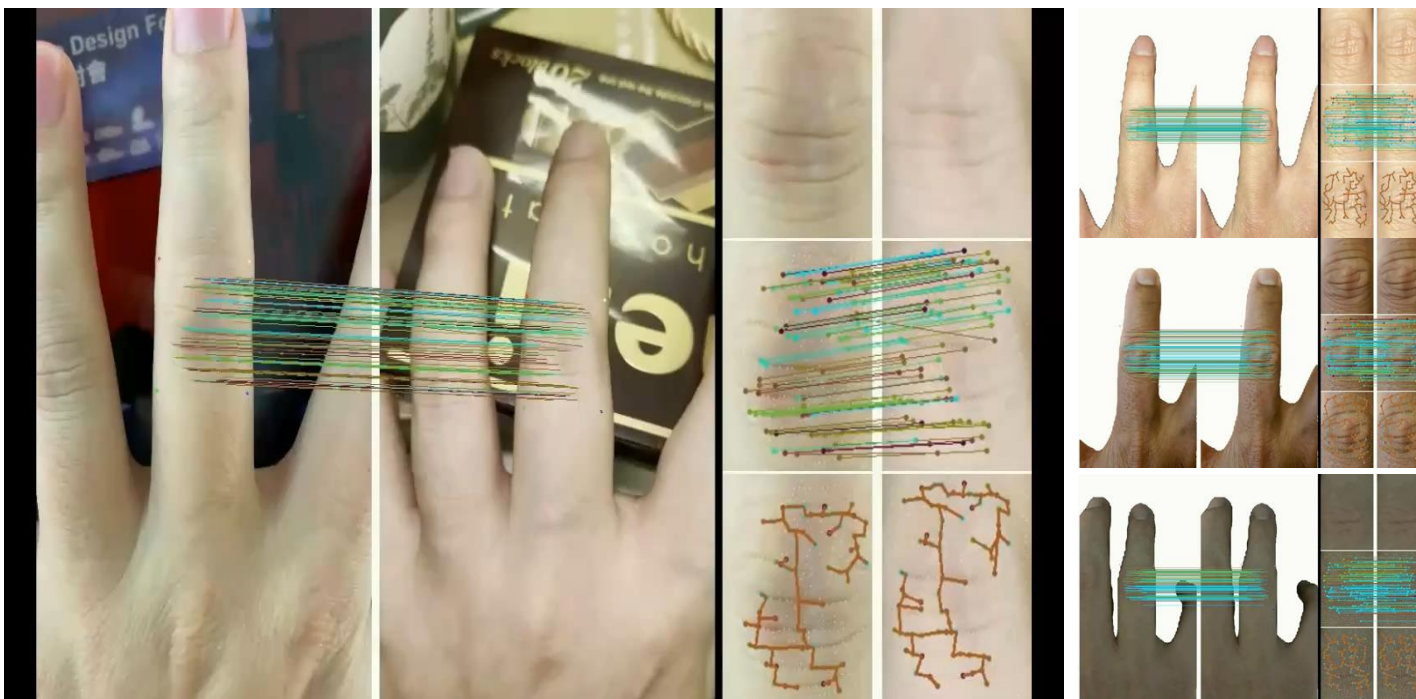
$$\overline{FRC} = \frac{2}{z(z-1)} \sum_{T^p \neq T^g} p(w; T^p, T^g)$$

Towards Explainable and Unprecedented Accuracy in Matching Challenging Finger Crease Patterns

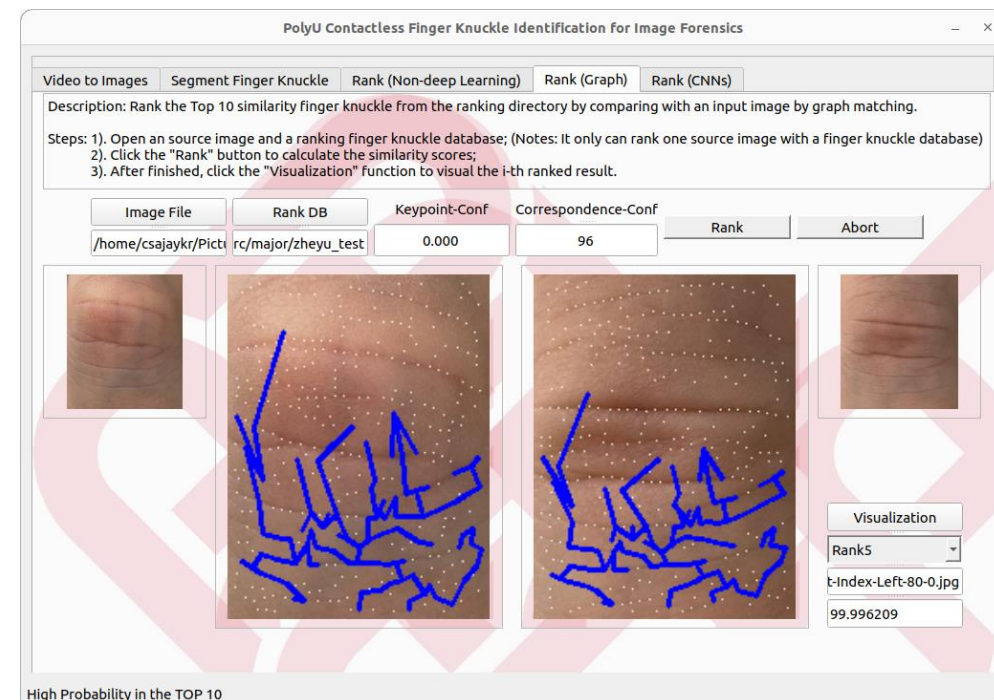
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Experiments: Explainable Visual Graphs



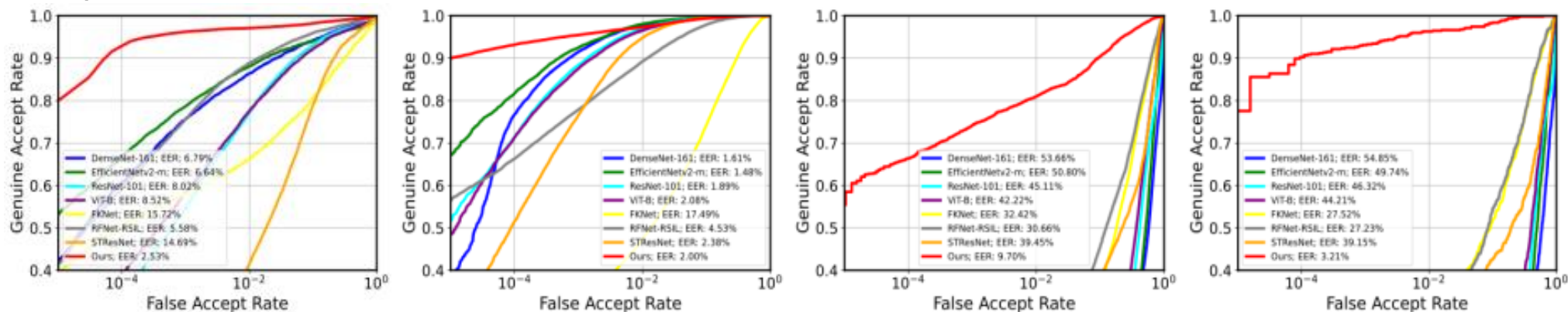
Knuckle Crease Keypoint Correspondence Estimation



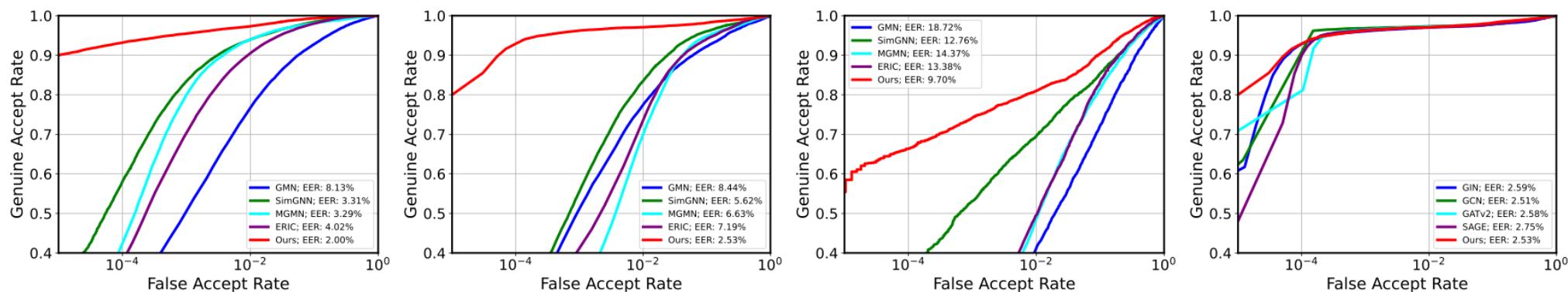
High Security and Law Enforcement Online System

Experiments: Within-Database and Cross-Database

- Compare with CNN or attention-based methods



- Compare with graph-based methods



Towards Explainable and Unprecedented Accuracy in Matching Challenging Finger Crease Patterns

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Experiments: Estimation of 2D Finger Knuckle Uniqueness

Estimated Resolution for This Dataset: ~ 200 dpi

Δ_k	Δ_f	(m, n, w, α)	$\overline{p}(T^p, T^g)$	$\bar{\lambda}_\alpha$	\overline{FRC}_α
6.750	0.726	(69,69,15,0.05)	2.980×10^{-4}	6.167	6.780×10^{-4}
6.750	0.726	(69,69,17,0.05)	2.980×10^{-4}	6.613	1.860×10^{-4}
6.750	0.726	(69,69,19,0.05)	2.980×10^{-4}	6.996	4.408×10^{-5}
6.750	0.726	(69,69,21,0.05)	2.980×10^{-4}	7.325	9.055×10^{-6}
3.750	0.726	(69,69,17,0.05)	1.031×10^{-4}	2.962	2.948×10^{-9}
5.250	0.726	(69,69,17,0.05)	1.914×10^{-4}	4.950	4.731×10^{-6}
8.250	0.726	(69,69,17,0.05)	4.119×10^{-4}	8.061	1.729×10^{-3}
9.750	0.726	(69,69,17,0.05)	5.273×10^{-4}	9.124	6.080×10^{-3}
6.750	0.626	(69,69,17,0.05)	6.210×10^{-6}	N/A	N/A
6.750	0.676	(69,69,17,0.05)	5.309×10^{-5}	1.367	1.546×10^{-14}
6.750	0.776	(69,69,17,0.05)	1.194×10^{-3}	12.476	8.305×10^{-2}
6.750	0.826	(69,69,17,0.05)	3.620×10^{-3}	14.893	2.421×10^{-1}

Estimated Resolution for This Dataset: ~ 760 dpi

Δ_k	Δ_f	(m, n, w, α)	$\overline{p}(T^p, T^g)$	$\bar{\lambda}_\alpha$	\overline{FRC}_α
9.258	0.844	(69,69,2,0.05)	5.387×10^{-5}	0.316	4.161×10^{-3}
9.258	0.844	(69,69,4,0.05)	5.387×10^{-5}	0.478	1.400×10^{-4}
9.258	0.844	(69,69,6,0.05)	5.387×10^{-5}	0.601	3.328×10^{-6}
9.258	0.844	(69,69,8,0.05)	5.387×10^{-5}	0.694	5.497×10^{-8}
9.258	0.844	(69,69,17,0.05)	5.387×10^{-5}	0.880	1.515×10^{-17}
6.258	0.844	(69,69,4,0.05)	2.459×10^{-5}	0.230	4.385×10^{-6}
7.758	0.844	(69,69,4,0.05)	3.765×10^{-5}	0.341	2.915×10^{-5}
10.758	0.844	(69,69,4,0.05)	7.306×10^{-5}	0.624	4.709×10^{-4}
12.258	0.844	(69,69,4,0.05)	9.494×10^{-5}	0.784	1.290×10^{-3}
9.258	0.744	(69,69,4,0.05)	5.932×10^{-6}	N/A	8.678×10^{-9}
9.258	0.794	(69,69,4,0.05)	1.954×10^{-5}	0.194	1.968×10^{-6}
9.258	0.894	(69,69,4,0.05)	1.247×10^{-4}	0.954	2.999×10^{-3}
9.258	0.944	(69,69,4,0.05)	2.510×10^{-4}	1.551	2.106×10^{-2}

Experiments: Estimation Uniqueness using Knuckle Crease Bifurcations and Endings

**Knuckle crease
bifurcations & *endings*:**



Dataset	Δ_k	Δ_f	(m, n, w, α)		$p(T^p, T^g)$		\overline{FRC}_α	
			Bifurcation	Ending	Bifurcation	Ending	Bifurcation	Ending
Hand Dorsal [30]	6.750	0.726	(55,55,11,0.05)	(14,14,2,0.05)	3.929×10^{-4}	4.300×10^{-4}	1.233×10^{-3}	5.843×10^{-4}
	6.750	0.726	(55,55,13,0.05)	(14,14,4,0.05)	3.929×10^{-4}	4.300×10^{-4}	2.807×10^{-4}	6.523×10^{-6}
	6.750	0.726	(55,55,15,0.05)	(14,14,6,0.05)	3.929×10^{-4}	4.300×10^{-4}	5.292×10^{-5}	4.640×10^{-8}
	5.250	0.726	(55,55,11,0.05)	(14,14,2,0.05)	2.541×10^{-4}	2.661×10^{-4}	7.947×10^{-5}	1.454×10^{-4}
	8.250	0.726	(55,55,11,0.05)	(14,14,2,0.05)	5.445×10^{-4}	6.267×10^{-4}	6.028×10^{-3}	1.642×10^{-3}
	6.750	0.676	(55,55,11,0.05)	(14,14,2,0.05)	7.376×10^{-5}	1.357×10^{-4}	5.252×10^{-10}	1.903×10^{-5}
	6.750	0.776	(55,55,11,0.05)	(14,14,2,0.05)	1.508×10^{-3}	1.106×10^{-3}	1.021×10^{-1}	6.641×10^{-3}
Ours	9.258	0.844	(61,61,3,0.05)	(8,8,1,0.05)	5.646×10^{-5}	2.068×10^{-4}	3.220×10^{-4}	9.609×10^{-5}
	9.258	0.844	(61,61,5,0.05)	(8,8,3,0.05)	5.646×10^{-5}	2.068×10^{-4}	5.419×10^{-6}	2.174×10^{-8}
	9.258	0.844	(61,61,7,0.05)	(8,8,5,0.05)	5.646×10^{-5}	2.068×10^{-4}	6.383×10^{-8}	3.146×10^{-12}
	7.758	0.844	(61,61,3,0.05)	(8,8,1,0.05)	4.006×10^{-5}	1.467×10^{-4}	9.134×10^{-5}	4.670×10^{-5}
	10.758	0.844	(61,61,3,0.05)	(8,8,1,0.05)	7.548×10^{-5}	2.781×10^{-4}	9.101×10^{-4}	1.779×10^{-4}
	9.258	0.794	(61,61,3,0.05)	(8,8,1,0.05)	2.103×10^{-5}	9.755×10^{-5}	9.293×10^{-6}	2.078×10^{-5}
	9.258	0.894	(61,61,3,0.05)	(8,8,1,0.05)	1.303×10^{-4}	3.858×10^{-4}	4.425×10^{-3}	3.373×10^{-4}