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SPOT-Face: Forensic Face Identification using Attention Guided Optimal Transport

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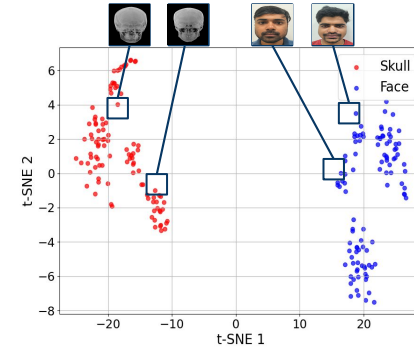
Content

1. Introduction
2. Related Work
3. Dataset
4. Methodology
5. Experimental Results
6. Conclusion and Discussion

3

Problems:

- Very high domain-gap.
- Skull don't have visible attribute like face.
- Limited availability of benchmark skull-to-face dataset.



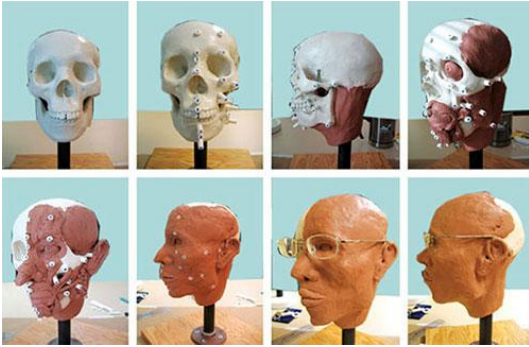
Objectives:

- To create craniofacial dataset.
- To develop novel method for craniofacial identification (either by matching or by reconstruction) from partial human remains in forensic investigations.



Existing Methods in Craniofacial Identification

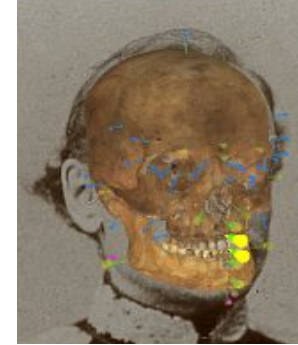
Traditional way



Clay based method^[2]

Limitations:

1. Based on tissue depth mark information.
2. Time consuming.
3. Need expert.



Superimposition based method^[3]

Limitations:

1. Based on tissue depth mark information.
2. Manual process of overlapping skull onto face.

[2] Taylor, Karen T. *Forensic art and illustration*. CRC press, 2000.

[3] Damas, S. et al. Forensic identification by computer-aided craniofacial superimposition: a survey. *ACM Comput. Surv.(CSUR)* 43, 1–27 (2011)

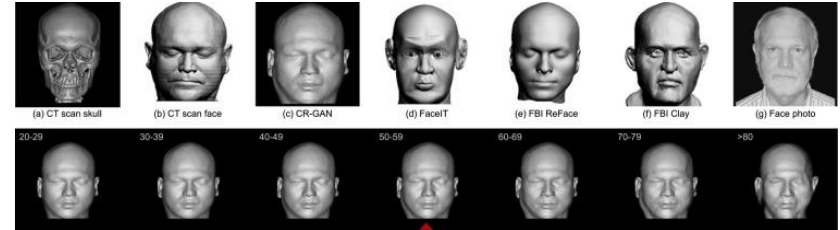
Modern deep learning way



Direct skull to face matching method^[4]

Limitations:

1. Data is collected from various sources.
2. HOG features are used for matching
3. Low accuracy



Direct skull to face generation method^[5]

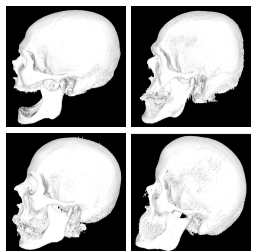
Limitations:

1. Data is collected from 3D CT scans.
2. Biased to chinese population.
3. Training instability.

[4] Singh, Maneet, et al. "Learning a shared transform model for skull to digital face image matching." *2018 IEEE 9th International Conference on Biometrics Theory, Applications and Systems (BTAS)*. IEEE, 2018.

[5] Li, Yuan, et al. "CR-GAN: Automatic craniofacial reconstruction for personal identification." *Pattern Recognition* 124 (2022): 108400.

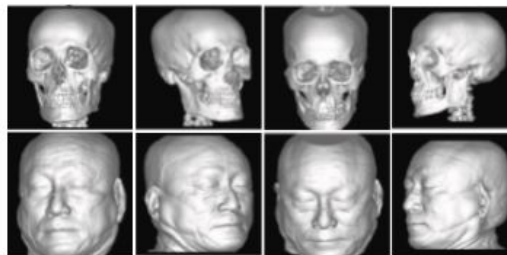
Craniofacial Datasets



MUG500+^[6]



IdentifyMe^[4]



Chinese CT dataset^[5]



IITMandi_S2F (Ours)

Dataset	# Subjects	# Paired Samples	Skull Modality	Face Modality	Pose/View	Public
MUG500+	500	×	CT-derived Skull	×	Multi-view	✓
IdentifyMe	464	464	Skull Photograph	Face Photograph	Frontal	✓
ICCR-Diff	4802	72016	CT-derived Skull	CT-derived Face	Multi-view	×
IITMandi_S2F (Ours)	51	5100	Skull X-ray	RGB Face Photograph	Frontal + Lateral	✓

[6] Li, Jianning, et al. "MUG500+: Database of 500 high-resolution healthy human skulls and 29 craniotomy skulls and implants." *Data in Brief* 39 (2021): 107524.

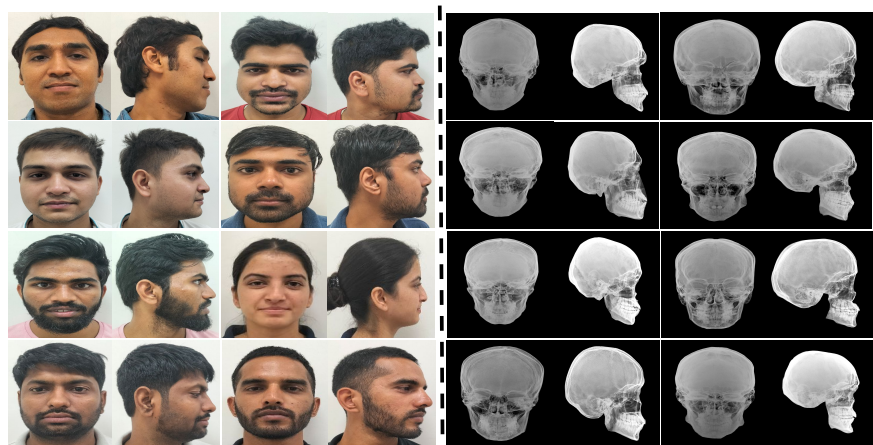
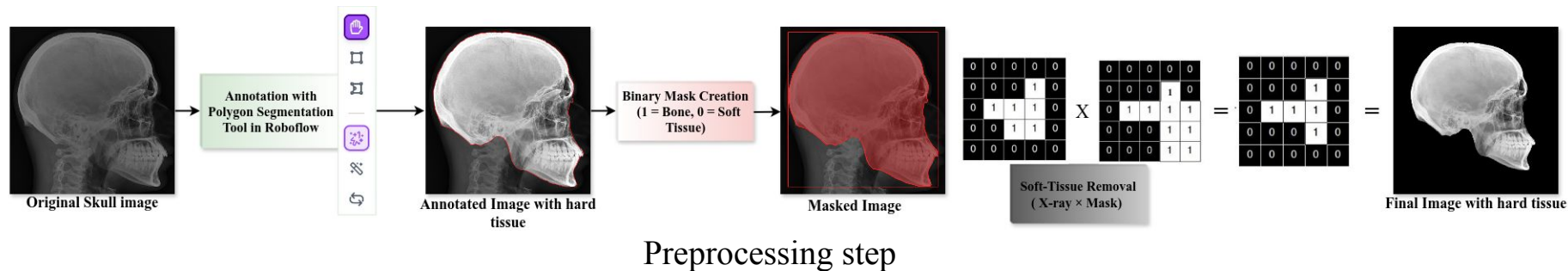
Methodology



Internationally recognized forensic artist at National center for Missing & Exploited Children, USA

Quote

“Some say, ‘That’s such an old case, why do it?’ I say, ‘Why not? They still deserve their names back.’”
- Joe Mullins, National Center for Missing and Exploited Children.



IITMandi_S2F

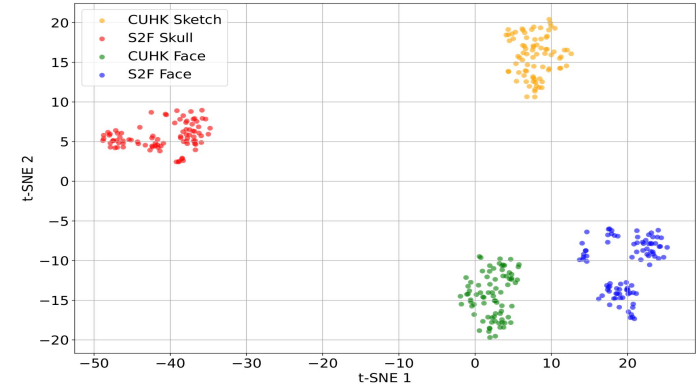
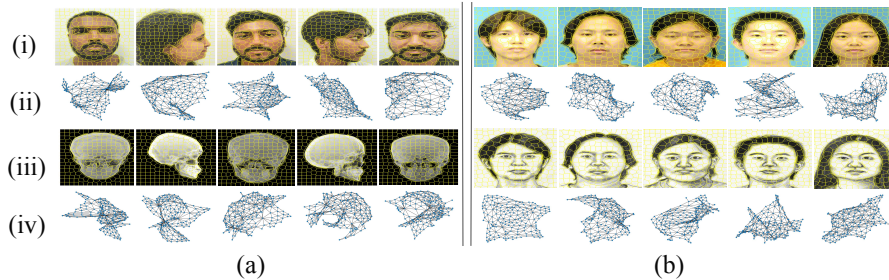
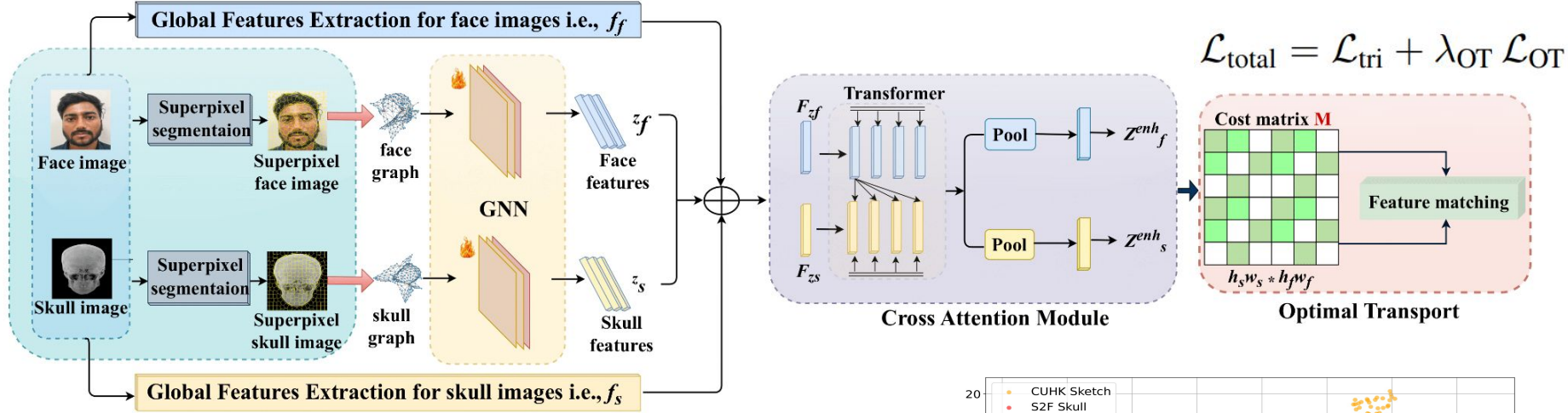
Contains X-ray image of face and their corresponding optical images from frontal and lateral views.

Total Subjects: 51 (Age: 21-30)

Male: 40

Female: 11

Title: *SPOT-Face: Forensic face identification using attention-guided optimal transport*^[7]



2: Quantitative Results^[7]

Comparison of different GNN backbones on Skull-Face (IIT_Mandi_S2F) and Sketch-Face (CUFS) datasets. Here, Recall@K and mAP@K (%) are reported and the best results are shown in bold.

Skull-Face (IIT_Mandi_S2F)				
Backbone	R@1	R@5	mAP@1	mAP@5
GCN	50.0	71.0	50.0	63.9
GAT	49.5	68.1	49.5	62.0
GraphSAGE	50.0	70.1	50.0	63.4
GraphTransformer	50.0	72.0	50.0	65.4

Sketch-Face (CUFS)				
Backbone	R@1	R@5	mAP@1	mAP@5
GCN (Quickshift) [8]	82.4	–	–	–
GCN (SLIC) [8]	87.7	–	–	–
GCN	88.4	88.9	88.4	88.8
GAT	88.4	88.6	88.4	88.6
GraphSAGE	88.4	88.8	88.4	88.6
GraphTransformer	88.4	89.0	88.4	88.9

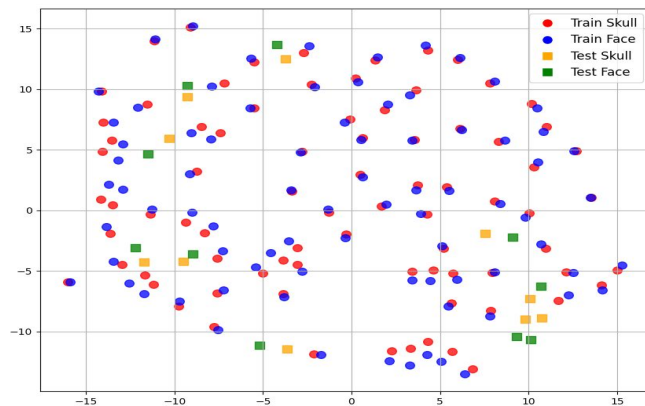
Impact of graph KNN hyperparameter (k) on Skull-Face (IIT_Mandi_S2F) and Sketch-Face (CUFS) datasets. Here, Recall@K and mAP@K (%) are reported and the best results are shown in bold.

k	Skull-Face (S2F)				Sketch-Face (CUFS)			
	R@1	R@5	mAP@1	mAP@5	R@1	R@5	mAP@1	mAP@5
4	50.0	71.0	50.0	64.5	88.4	89.0	88.4	88.7
6	50.0	72.0	50.0	65.4	88.4	88.9	88.4	88.7
8	49.0	66.1	49.0	60.6	88.4	89.0	88.4	88.8
12	50.0	72.0	50.0	64.1	88.4	89.0	88.4	88.9

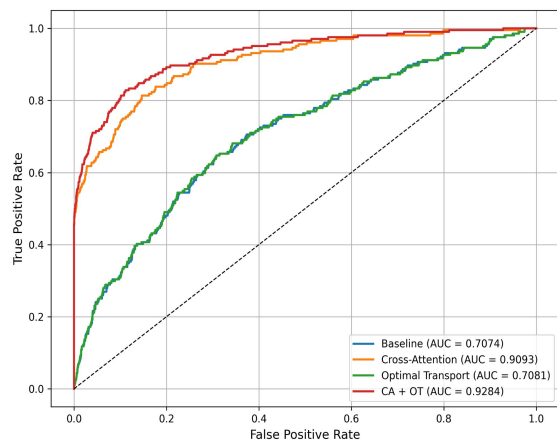
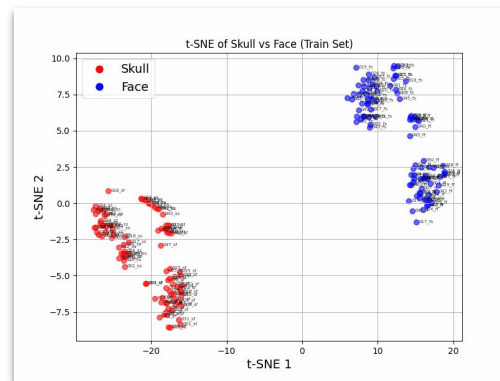
Ablation of Cross-Attention (CA) and Optimal-Transport (OT) modules on Skull-Face (S2F) and Sketch-Face (CUFS) matching. Recall@K and mAP@K (%) are reported. Best results are shown in bold.

CA	OT	Skull-Face (S2F)				Sketch-Face (CUFS)			
		R@1	R@5	mAP@1	mAP@5	R@1	R@5	mAP@1	mAP@5
✗	✗	31.3	46.0	31.3	38.1	83.4	88.7	83.4	85.4
✓	✗	48.5	60.7	48.5	56.8	88.4	88.8	88.4	88.7
✗	✓	31.3	45.5	31.3	37.7	83.4	88.8	83.4	85.8
✓	✓	50.0	72.0	50.0	65.4	88.4	89.0	88.4	88.9

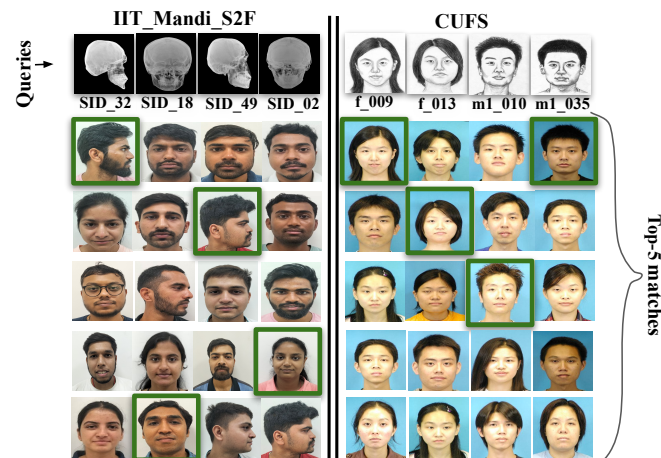
2: Qualitative Results^[7]



2D t-SNE visualization of skull and face embedding after training.



ROC-AUC curve



Top-5 retrieval results

Conclusion and Discussion

- We introduced a framework SPOT-Face for craniofacial identification with matching using 2D images of skulls and faces, including both frontal and lateral views.
- Our proposed method illustrates improved performance over previous methods, providing an alternative to manual annotation.
- Our results indicate that our proposed method performs significantly better in sketch-to-face matching, demonstrating its general applicability in various matching scenarios.
- Future work will explore how graph-based methods can be utilized for craniofacial reconstruction.

Thank You



Corresponding author



Co-author



Google scholar



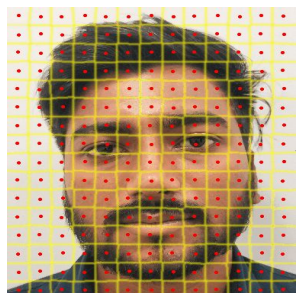
ResearchGate



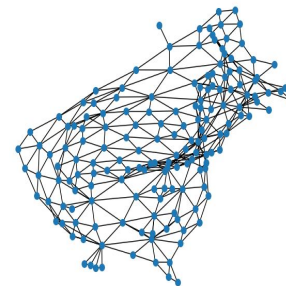
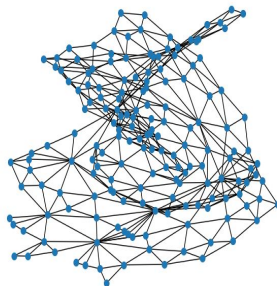
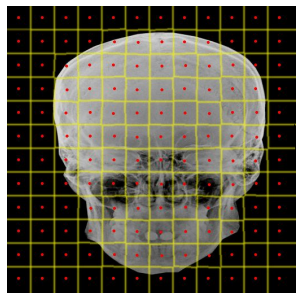
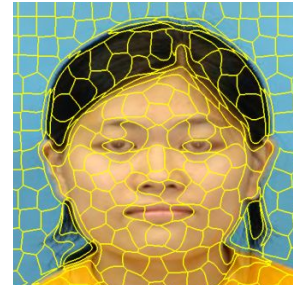
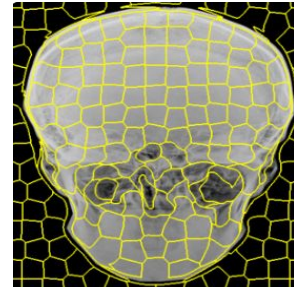
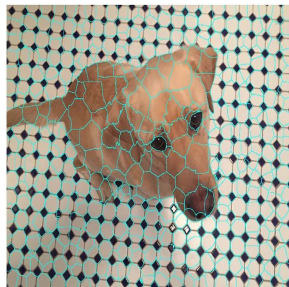
LinkedIn

Supplementary slides

How to capture more structural information? SLIC^[9] superpixel segmentation to Graph



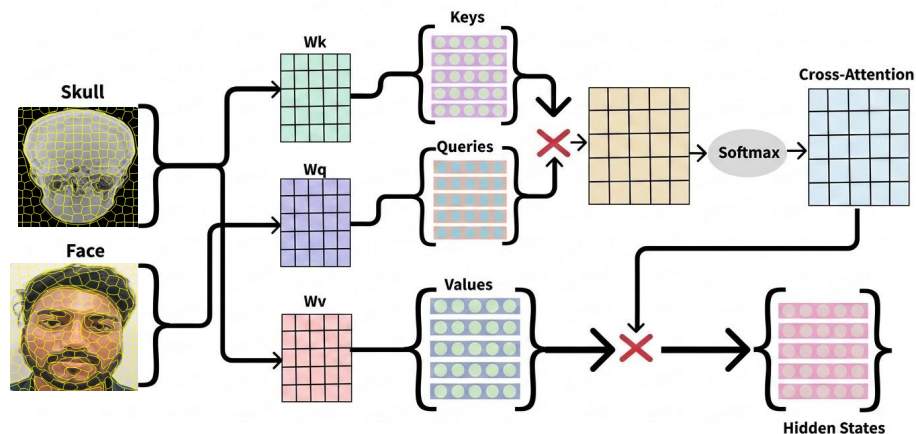
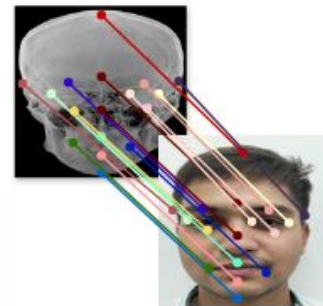
SLIC: Simple Linear Iterative Clustering



Node: each superpixel (R, G, B, x, y)

Edge: connection between each neighboring superpixel (KNN neighbour)

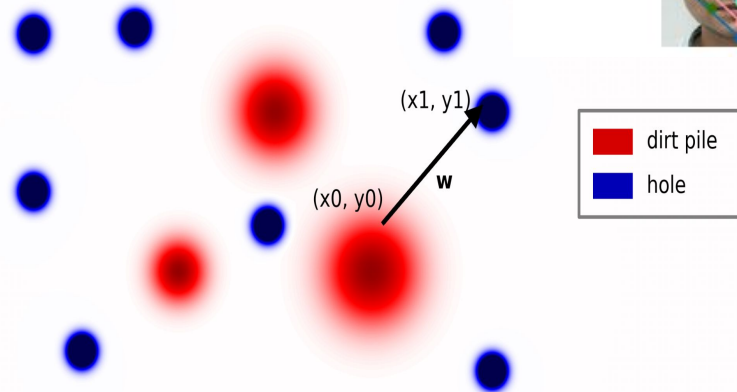
How to get meaningful semantic correspondence? Cross-attention with Optimal Transport



Cross-Attention^[10] (CA)

- Face has skin texture, eyes, nose, mouth.
- Skull has bone structure, eye socket, nasal bone.

CA address→Which skull features should a face feature look at to become more informative?



Optimal Transport^[11] (OT)

- Nasal bone → nose face = low cost.
- Eye socket → jaw face = high cost.

OT address→What is the best overall matching between skull regions and face regions?

[10] Vaswani, Ashish, et al. "Attention is all you need." *Advances in neural information processing systems* 30 (2017).

[11] Kantorovich, Leonid V. "On the Translocation of Masses." *Journal of mathematical sciences* 133.4 (2006).

