

Automatic Keypoint Detection on 3D Faces Using a Dictionary of Local Shapes

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Aim

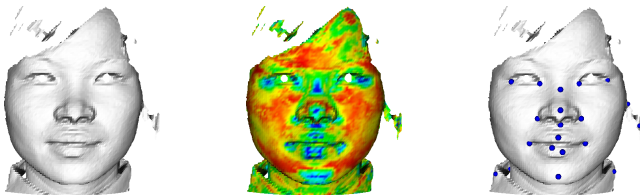
What

Why

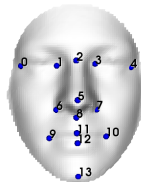
How

Results

Conclusion

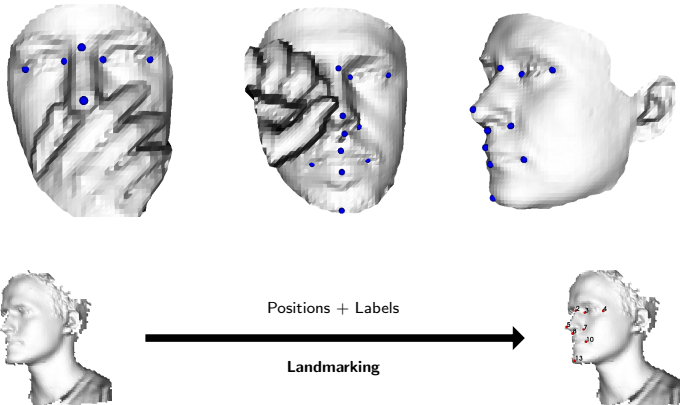


- Keypoints detection (NOT LANDMARKS)
- Similar to any of 14 learnt features (Dictionary of local shapes)



Part of a bigger project

What
Why
Long Term
Objective
Gap in Research
How
Results
Conclusion



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What

Why

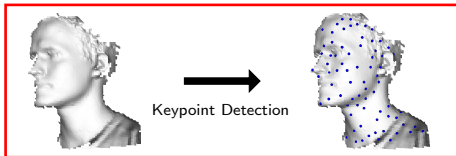
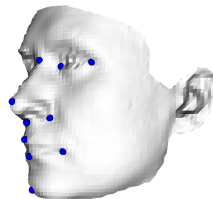
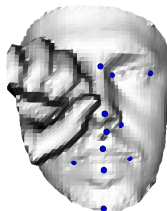
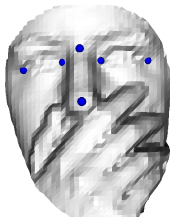
Long Term
Objective

Gap in Research

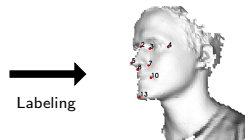
How

Results

Conclusion



Keypoint Detection



Labeling

What

Why

Long Term
Objective

Gap in Research

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Most literature:

- 3 points max or single-point-of-failure design
- Weak features often discarded
- Almost no work on combining more than 2 descriptors
- Little literature that examine multiple descriptors over multiple scales
- Most people focused on landmarking, without giving the intermediate results on candidate detection (keypoints)

What

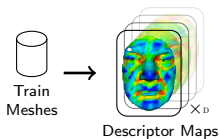
Why

How

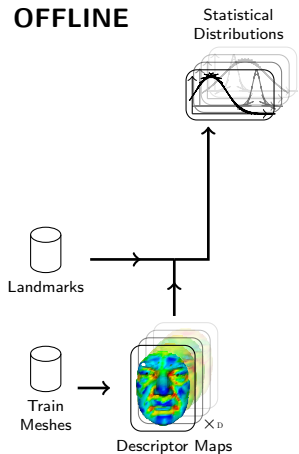
Results

Conclusion

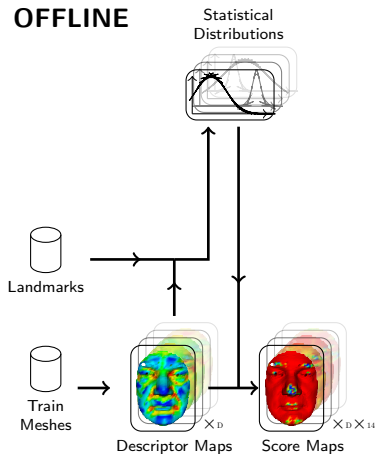
OFFLINE

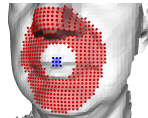


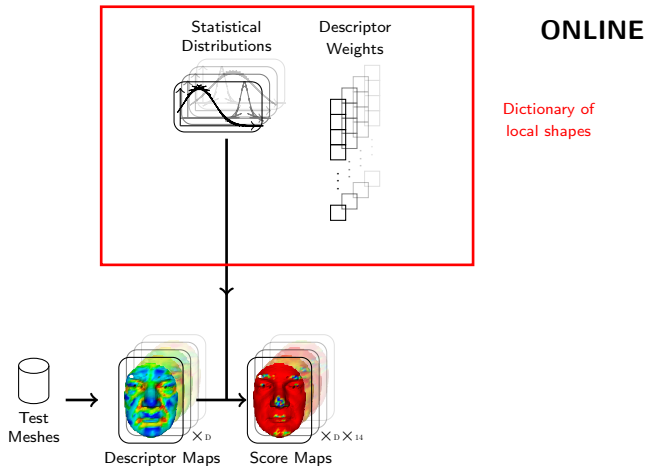
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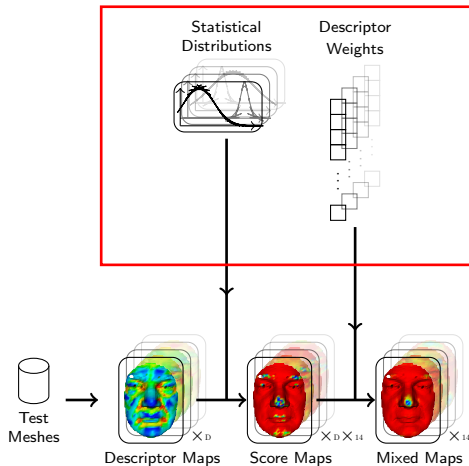


OFFLINE



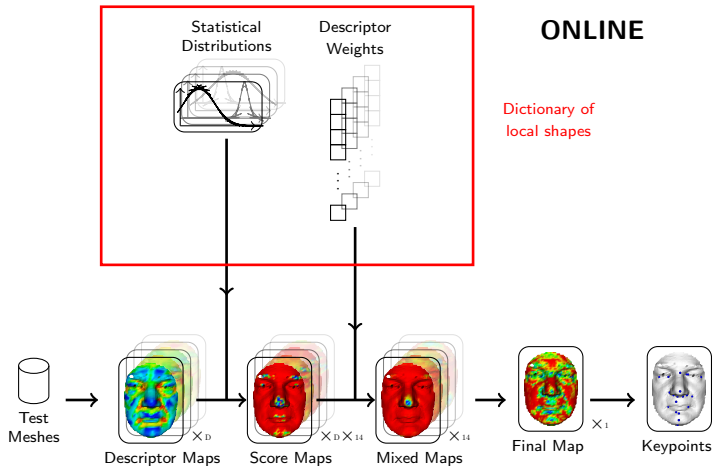






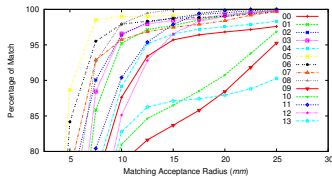
ONLINE

Dictionary of
local shapes

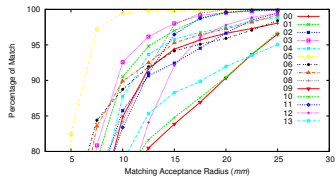


Results

- Sparse selection (max 1%)
- Repeatable (same subject registration)
 - ~75% (at 10 mm)
- Close to human hand-placed landmarks
 - average All: ~85% (at 10 mm)
 - average Nose: ~99% (at 10 mm)
 - average Eyes: ~90% (at 10 mm)
- High proportion of the local shapes retrieved
 - ~11.88/14 (at 10 mm)



Configuration 1



Configuration 2

Examples

What

Why

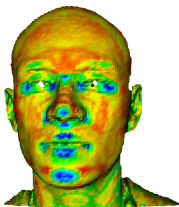
How

Results

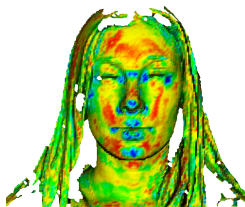
Results

Examples

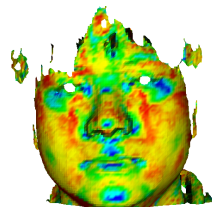
Conclusion



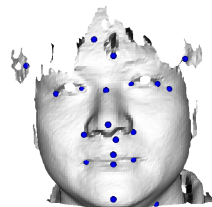
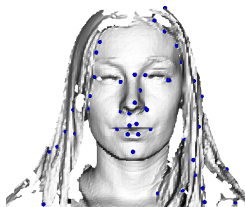
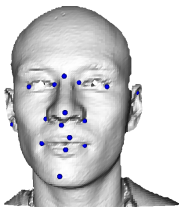
0.224 0.612 1.00



0.265 0.632 1.00



0.270 0.635 1.00



Conclusion

- Good points:
 - Detects "weak" features
 - No single-point-of-failure design
- Limitations:
 - Can be time consuming
article: 7s, now: 0.5s (8 desc.)
 - Linear combination of scores
- Future Work:
 - Non linear methods (boosting, kernel methods)
 - Structural matching to deduce correspondences
 - Comparison with a new clustering technique for keypoint detection

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Thank You For Listening!

<http://www.cs.york.ac.uk/~creusot>