### Landmark Labelling for 3D Faces

Clement Creusot, Nick Pears, Jim Austin





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- Motivation
- Problem
- Solution
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- Conclusion



- Non-cooperative Recognition at a distance
- Modality
- Difficult Cases
- Review
- Assumptions

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### **Motivation**

## Non-cooperative Recognition at a distance

#### Motivation

Non-cooperative Recognition

- at a distance
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- Application:
  - Surveillance
  - Human-Machine Interaction





## Non-cooperative Recognition at a distance

#### Motivation

Non-cooperative Recognition

- at a distance
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Results

- Application:
  - Surveillance
  - Human-Machine Interaction
- Problems:
  - Pose
  - Occlusion
  - Speed



From [Savran et al., 2008]



From [Savran et al., 2008]



Non-cooperative Recognition

at a distance

ModalityDifficult Cases

Review

Assumptions

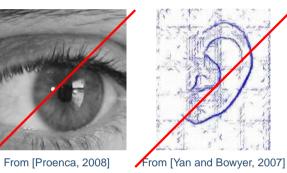
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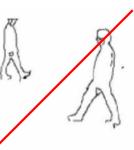
Conclusion

# Non-Cooperative > Anti-cooperative Proved possible for big database









From [Havasi et al., 2007]



Non-cooperative Recognition

at a distance

ModalityDifficult Cases

Review

Assumptions

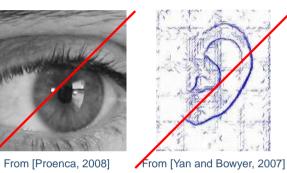
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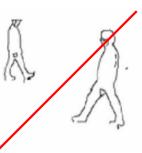
Conclusion

# Non-Cooperative > Anti-cooperative Proved possible for big database



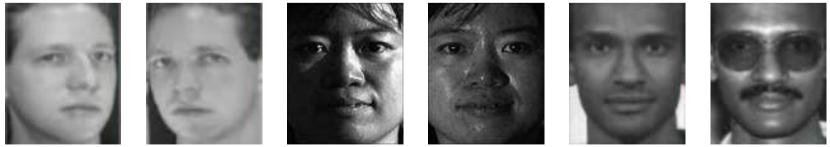


From [Phillips et al., 2005]



From [Havasi et al., 2007]

### ■ 2D or 3D ?







### **Difficult Cases**

	vat	

 Non-cooperative Recognition at a distance
 Modality

• Difficult Cases

Review

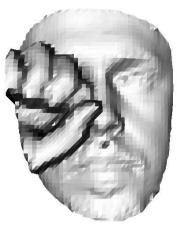
Assumptions

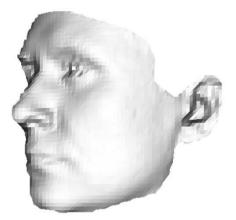
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- Recognition:
  - Holistic method  $\rightarrow$  Need for good Registration
  - Feature based method  $\rightarrow$  Need for good Feature Localisation



 Non-cooperative Recognition at a distance
 Modality

Difficult Cases

Review

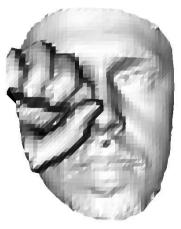
Assumptions

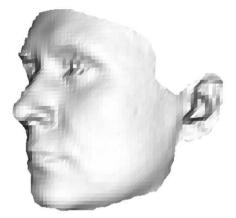
Problem

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- Recognition:
  - ◆ Holistic method → Need for good Registration
  - ◆ Feature based method → Need for good Feature Localisation
- Will often fail at preprocessing
  - Naive methods for feature detection
  - Strong assumptions



### **Difficult Cases**

	vat	

 Non-cooperative Recognition at a distance
 Modality

Difficult Cases

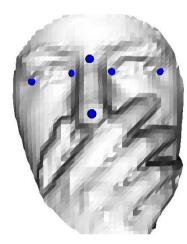
Review

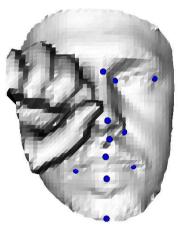
Assumptions

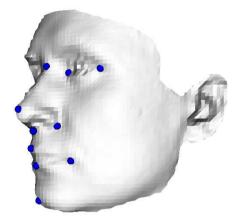
Problem

Solution

Results







- Recognition:
  - Holistic method  $\rightarrow$  Need for good Registration
  - Feature based method  $\rightarrow$  Need for good Feature Localisation
- Will often fail at preprocessing
  - Naive methods for feature detection
  - Strong assumptions
- Recquire better feature detection



 Non-cooperative Recognition at a distance

Modality

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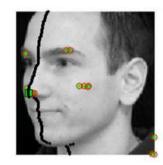
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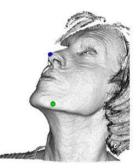
Conclusion

### Almost all paper expect non occluded frontal face

- A few that don't:
  - Some orientation change:
    - [Colbry et al., 2005]: Curvature + ICP + Relaxation
    - [Lu and Jain, 2006]: Directional Maximum
    - [Faltemier et al., 2008]: Rotated Profile Signature







[Colbry et al., 2005] [Lu and Jain, 2006] [Faltemier et al., 2008]

- Almost all papers expect the nose will be present
- Most papers recquire two well defined inner corners of the eyes

### Assumptions

#### Motivation

- Non-cooperative Recognition at a distance
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Assumptions

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Results

- The ones we needed to make:
  - At least half of the face is visible
  - There exist features repeatable across individual
- The ones we did not make:
  - All landmark are present and will match there descriptor
  - Candidates for one landmark descriptor are rare
- The ones we made (only in post-processing)
  - The face is roughly convex
  - Faces are not too flexible (≠ hand)
  - Only 1 face per scene



#### Problem

• The landmark Detection Problem

Input Generation

Solution

Results

Conclusion

### **Problem**

## **The landmark Detection Problem**

Activation Problem The landmark Detection Problem Input Generation Solution	Input Mesh	Positions + Labels	Landmarks
Conclusion		Landmarking	

### Landmark = Position + Label

- Two Approaches:
  - Select One Label + Find Corresponding Position
  - Find All Positions + Find Corresponding Labels





Problem

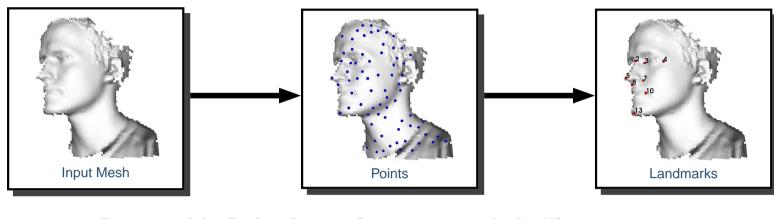
 The landmark Detection Problem
 Input Generation

• Input Genera

Solution

Results

Conclusion



**Repeatable Point Detection** 

Labelling

Landmark = Position + Label

- Two Approaches:
  - Select One Label + Find Corresponding Position
  - Find All Positions + Find Corresponding Labels

## **The landmark Detection Problem**



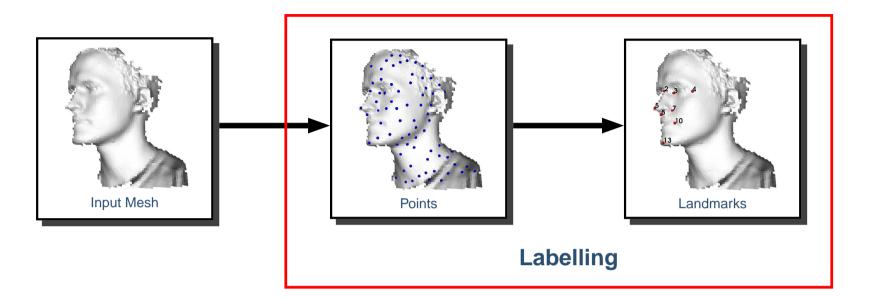
#### Problem

The landmark Detection
 Problem
 Input Generation

Solution

Results

Conclusion



### Landmark = Position + Label

- Two Approaches:
  - Select One Label + Find Corresponding Position
  - Find All Positions + Find Corresponding Labels

### **Input Generation**

Motivation

REFERENCES

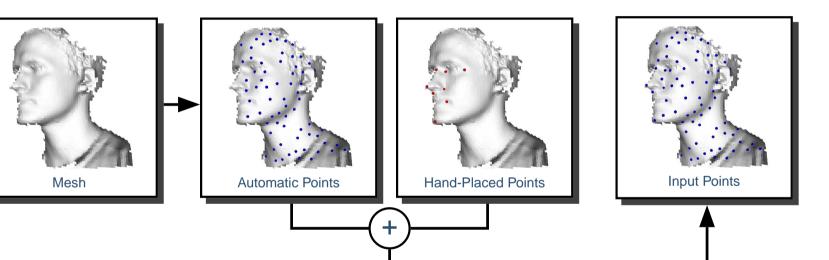
Problem

The landmark Detection
 Problem

Input Generation

Solution

Results





Problem

#### Solution

Our Strategy

Graph Generation

• Graph Matching

Elimination

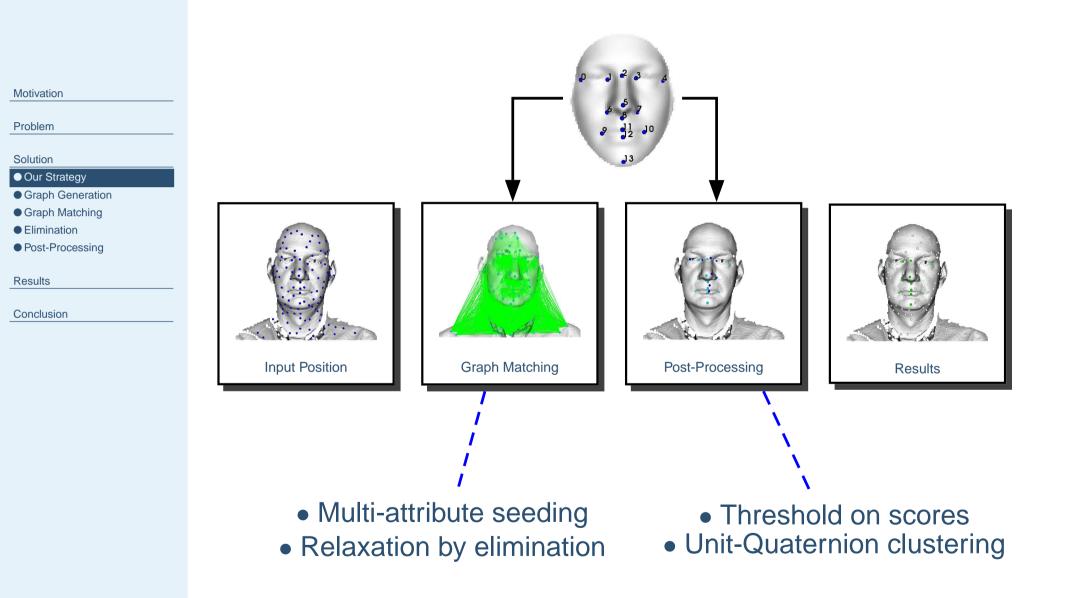
Post-Processing

Results

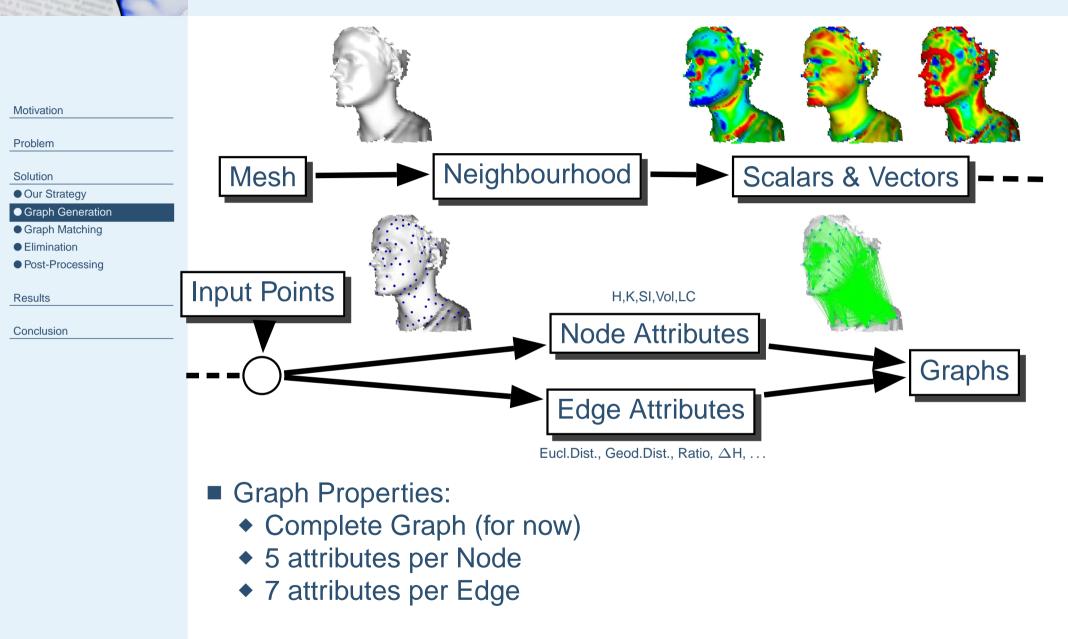
Conclusion

### **Solution**





### **Graph Generation**





#### Problem

Solution

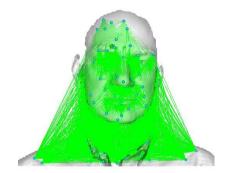
- Our Strategy
- Graph Generation
- Graph Matching
- Elimination
- Post-Processing

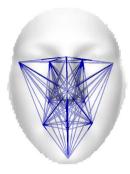
Results

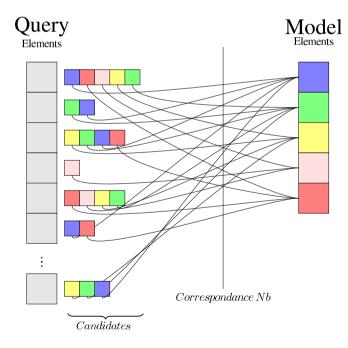
Conclusion

### Structure

- list of candidates
- Associated scores







## **Graph Matching**

#### Motivation

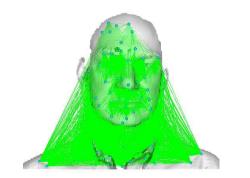
#### Problem

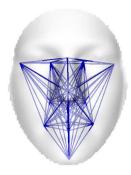
Solution

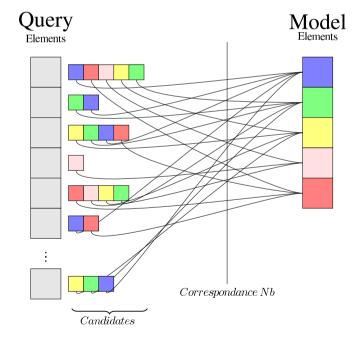
- Our Strategy
- Graph Generation
- Graph Matching
- Elimination
- Post-Processing

Results

- Structure
  - list of candidates
  - Associated scores
- Objective:
  - Reduce correspondence Nb







## **Graph Matching**

#### Motivation

#### Problem

Solution

- Our Strategy
- Graph Generation
- Graph Matching
- Elimination
- Post-Processing

Results

Conclusion

- Structure
  - list of candidates
  - Associated scores
- Objective:
  - Reduce correspondence Nb

Query 🌩

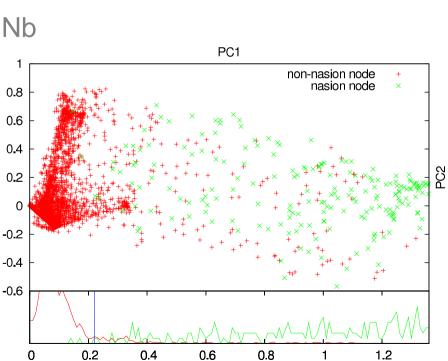
 $\sum$ 

1.1

 $\times 0.2$ 

 $0.5 \times 0.2$ 

- Seeding
  - Partial scores  $\stackrel{LDA}{\rightarrow}$  Score



• N properties

 $\times 0.5$ 

5.6

 $\times 0.3$ 

0.9×0.3

← Model

= Score

120.0

₽

 $0.65 \times 0.5$ 

## **Graph Matching**

#### Motivation

#### Problem

Solution

- Our Strategy
- Graph Generation
- Graph MatchingElimination
- Post-Processing

Results

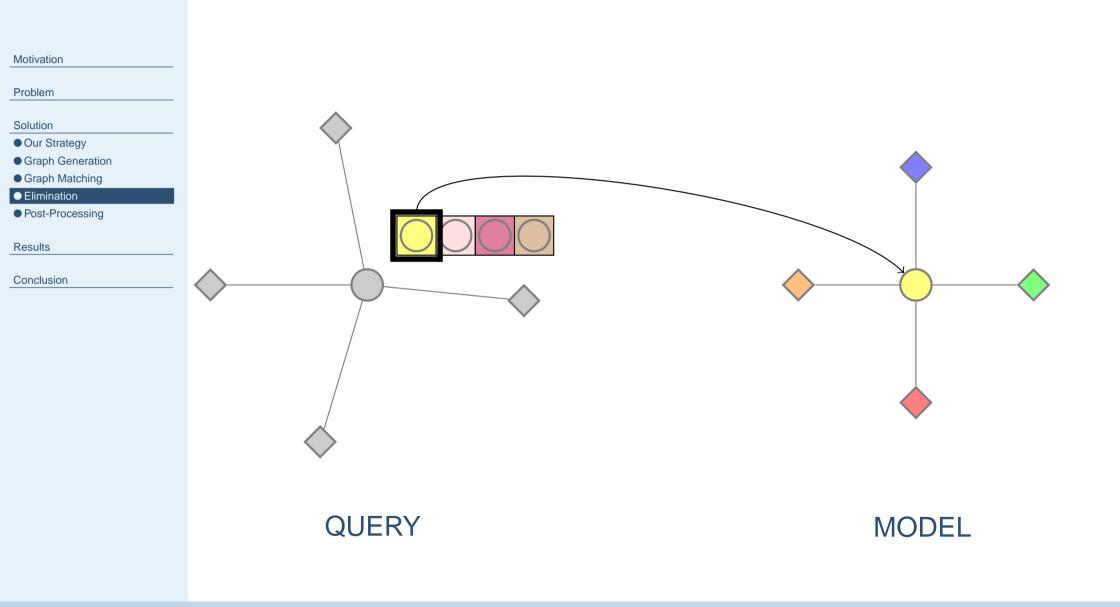
- Structure
  - list of candidates
  - Associated scores
- Objective:
  - Reduce correspondence Nb
- Seeding
  - Partial scores  $\stackrel{LDA}{\rightarrow}$  Score
- Relaxation on hyperedges (≠ [Christmas et al., 1995])



Motivation	
Problem	
Solution  Our Strategy Graph Generation Graph Matching  Image: Strategy Results Conclusion	
	QUERY

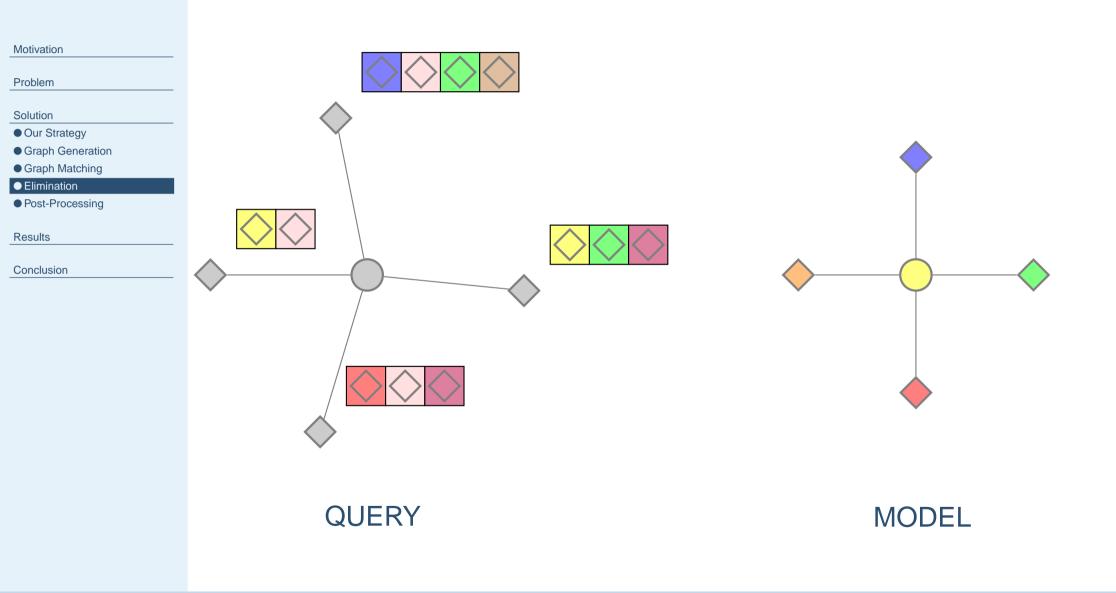
REFERENCES



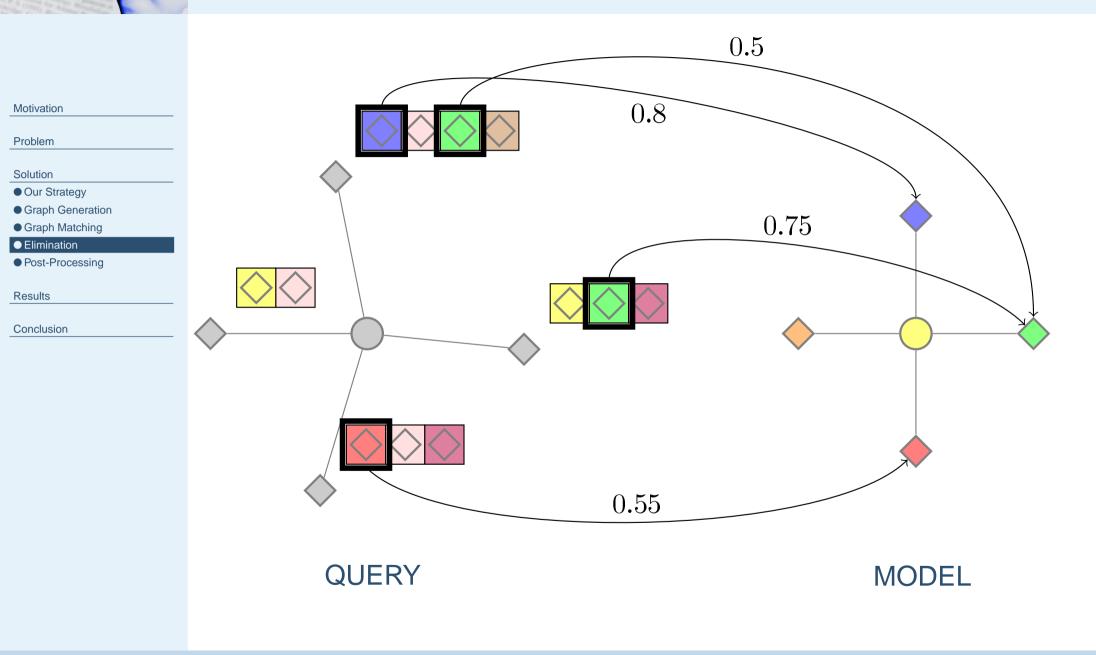


REFERENCES

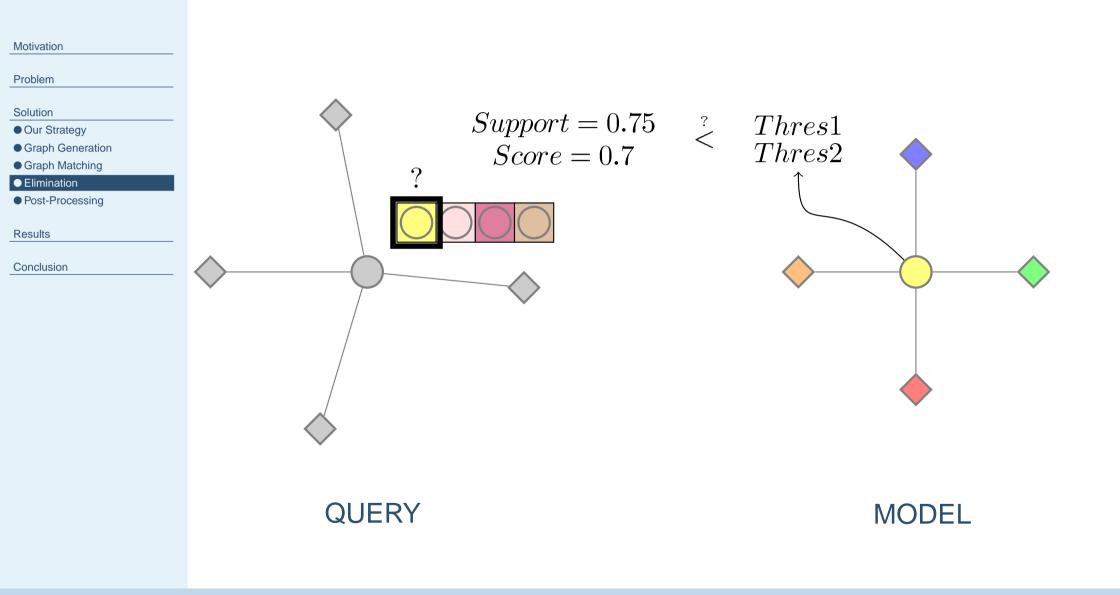




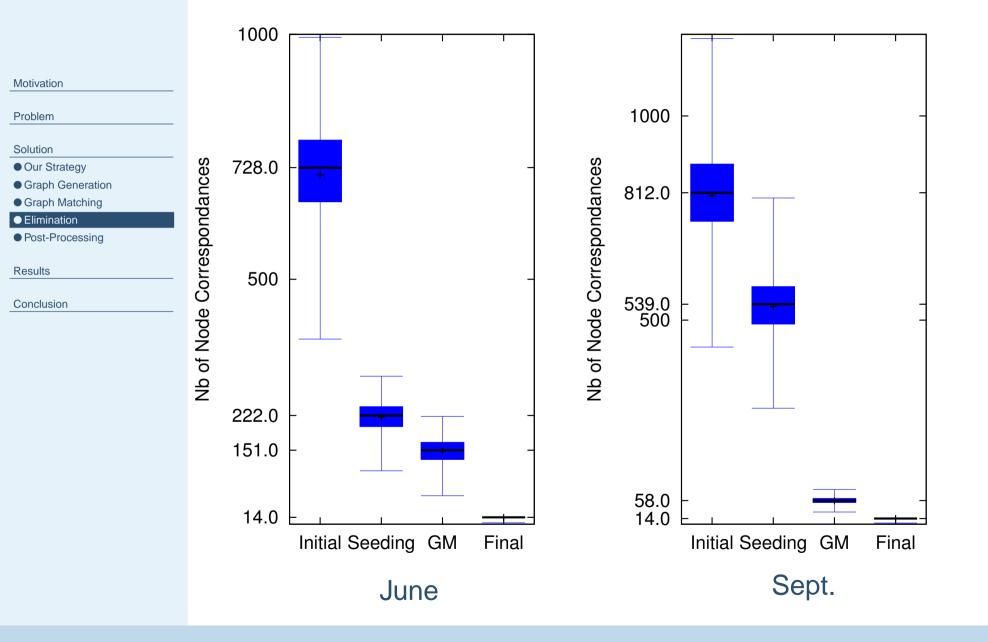
### Elimination











### **Post-Processing**

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Problem

Solution

Our Strategy

Graph Generation

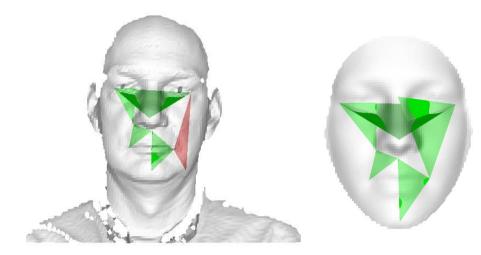
Graph Matching

Elimination

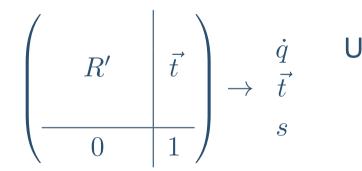
Post-Processing

Results

Conclusion



### Transformation Matrix 4x4:



Unit Quaternion Translation Scale

### **Post-Processing**

1\/	OTIN.	/ation	
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Problem

Solution

• Our Strategy

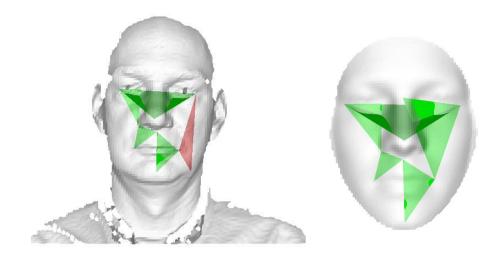
Graph Generation

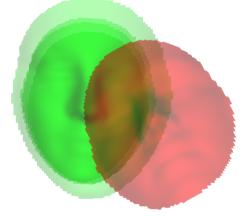
Graph Matching

Elimination

Post-Processing

Results





- Clustering
- Mean Transformation
- Final Correspondence



Problem

Solution

#### Results

Databases

Results

Conclusion

### **Results**

### Databases

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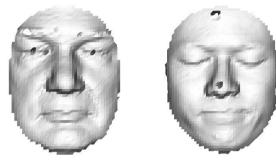
Problem

Solution

Results

Databases

Results





- FRGC v2
  - ◆ 4950 faces from 557 people
  - 200 in train set
  - 4750 in test set (3108 Neutral, 1642 Expression)
  - cropped
- Bosphorus
  - 4666 faces from 105 people
  - Occlusion, Expression, Rotation
  - 99 in train set (20 for profile)

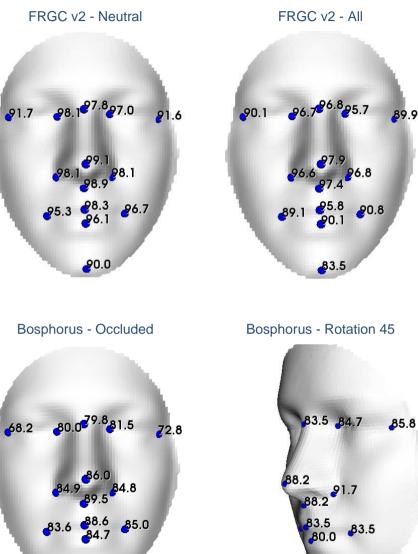


## Results

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Results
<ul> <li>Databases</li> </ul>
● Results
Conclusion

For now:

- ◆ 6.3% bad final registration
- If automatic landmarks only:
  - ◆ 10.4% bad final registration
- The system doesn't collapse when dealing with occlusion or pose variation



68.7

72.2



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### Conclusion

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Conclusion

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### Good

- Very few assumptions on the input data
- Graphs are very versatile
- Bad
  - Non optimised (preliminary results)
  - Naive post-processing
- Future Work
  - Try different graph topologies
  - Improve robustness to missing points
  - Deal with non-cropped faces
  - Try higher order hyperedges

### Conclusion

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### Good

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# Thank you !

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