# Wall Painting Reconstruction Using a Genetic Algorithm

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#### Motivation

Wall paintings recovered from Akrotiri (Santorini), Greece







Part of a reconstructed fresco

Antelope Fresco

#### **Related Problems**

#### **Reconstruction of Torn Documents**





#### Reconstruction of Fragmented Objects



[Huang '06]

**RGB-D** Reconstruction





[Henry '10]





Real Fragments

**Reconstruction Pipeline** 



Real Fragments

3D Models

Reconstruction Pipeline





# Step 1: Scanning

#### Use a laser scanner to acquire a 3D surface





**Surface Scanning** 



**Physical Fragments** 

**Digital Scan of a Fragment** 

# Step 2: Matching

#### Use an algorithm to find and rank potential pairwise matches between fragments



# Step 3: Global Assembly

Find a globally optimal arrangement in any cluster of fragments

50K ranked matches, most are wrong

#### Challenge

?

Which matches are part of the correct assembly?

50K ranked matches, most are wrong

### Challenge



50K ranked matches, most are wrong

# Existing Assembly Strategies

Dense Cluster Growth

 clusters of fragments are merged by means of best candidate match



[Goldberg et al. 2002]

# Existing Assembly Strategies

#### **Hierarchical clustering**

- Start with all fragments as singleton clusters and merge in priority order
- Fragment alignments are optimized in each iteration
- Process terminates when no good merges are possible





Need small assemblies to construct larger ones

### Approach



#### Initialization



### Population

Clusters of fragments joined by matches



#### Recombination

Create a new cluster from two parent clusters



By Fragment

#### Recombination

Create a new cluster from two parent clusters





## Cluster Optimization

Find rotation angles for each fragment with respect to the global coordinate system

- Find unknown angles

$$\theta_1, \theta_2, \ldots, \theta_n \in [0, 2\pi]$$

- from known pairwise matches

$$\delta_{ij} = \theta_i - \theta_j$$

Resulting set of equations:  $\theta_i - \theta_j = \delta_{ij} \mod 2\pi$ 

Solving the above is a non-linear problem



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#### Selection

- Cluster Fitness
- Selection Algorithm
  - Round Robin
  - Diversification

#### Cluster Fitness

Assign clusters scores based on their intrinsic properties



#### Convex Hull Ratio





#### **Round-Robin Selection**

Limit number of clusters with the same fragment



### Diversification



## Genetic Algorithm



#### Evaluation

- Reconstruction quality?
- Algorithmic contributions?

Method	# of Fragments	F-score
Our Full System (GA)	90	0.823
Hierarchical Clustering (HC)	42	0.411
Dense Cluster Growth (DCG)	7	0.082

Performance comparison of different wall painting reconstruction methods with our fitness and feasibility criteria.

F-score = 2 \* Precision \* Recall Precision + Recall

#### of correct matches



[Dense Cluster Growth]





[Castaneda et al. 2011]





[Dense Cluster Growth]







[Dense Cluster Growth]





[Our Result]



[Ground Truth]



[Dense Cluster Growth]





[Castaneda et al. 2011]

#### Evaluation: Does Our Cluster Optimization Help?



Comparison of performance of our method with various cluster optimization routines.

#### Evaluation: Do Larger Clusters Have Higher Precision?



Precision - Recall comparison of different iteration of our algorithm (aggregated over 10 runs)

## Summary

- Genetic algorithms are not greedy they defer selection of matches until large assemblies are formed
- Our genetic algorithm is able to find larger assemblies than previous methods
- More robust cluster optimization helps find larger assemblies

#### Limitations

- GA is very time consuming (tens of hours)
- Hand tuned cluster scoring function

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