

Shape Recognition in 3D Point-Clouds

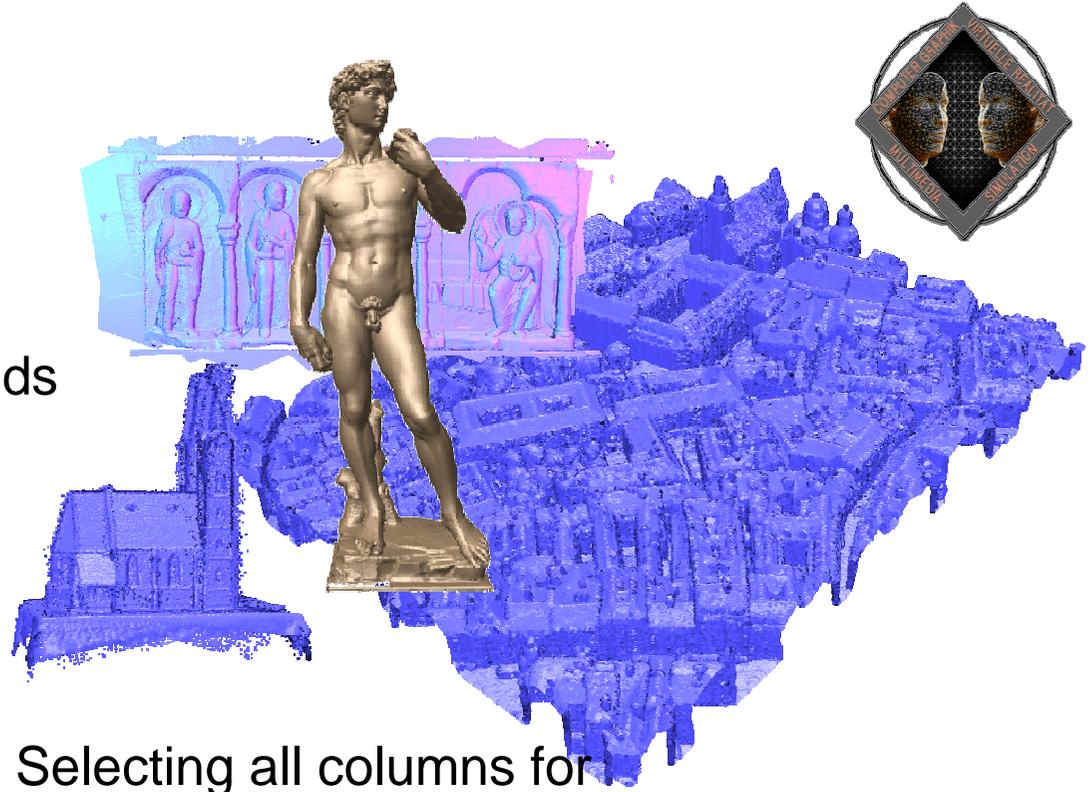
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Introduction

- Size of geometric datasets increases quickly
- In particular: Large point-clouds from 3d scanning



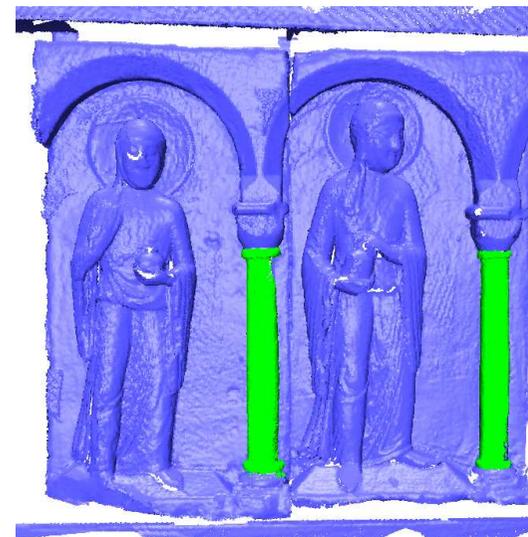
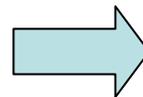
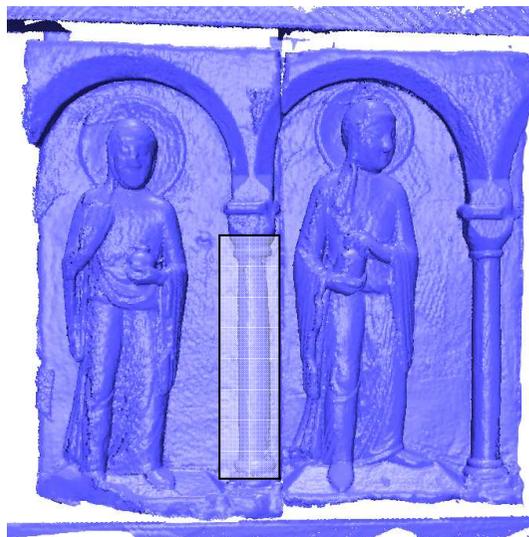
Problem:

- Interaction still limited, e.g.: Selecting all columns for assigning them colors or textures
- Common approach:
 - Selecting columns with mouse
 - Time consuming and imprecise



Introduction

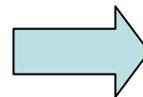
- Desirable approach:
 - Select single column
 - find all other columns automatically





Introduction

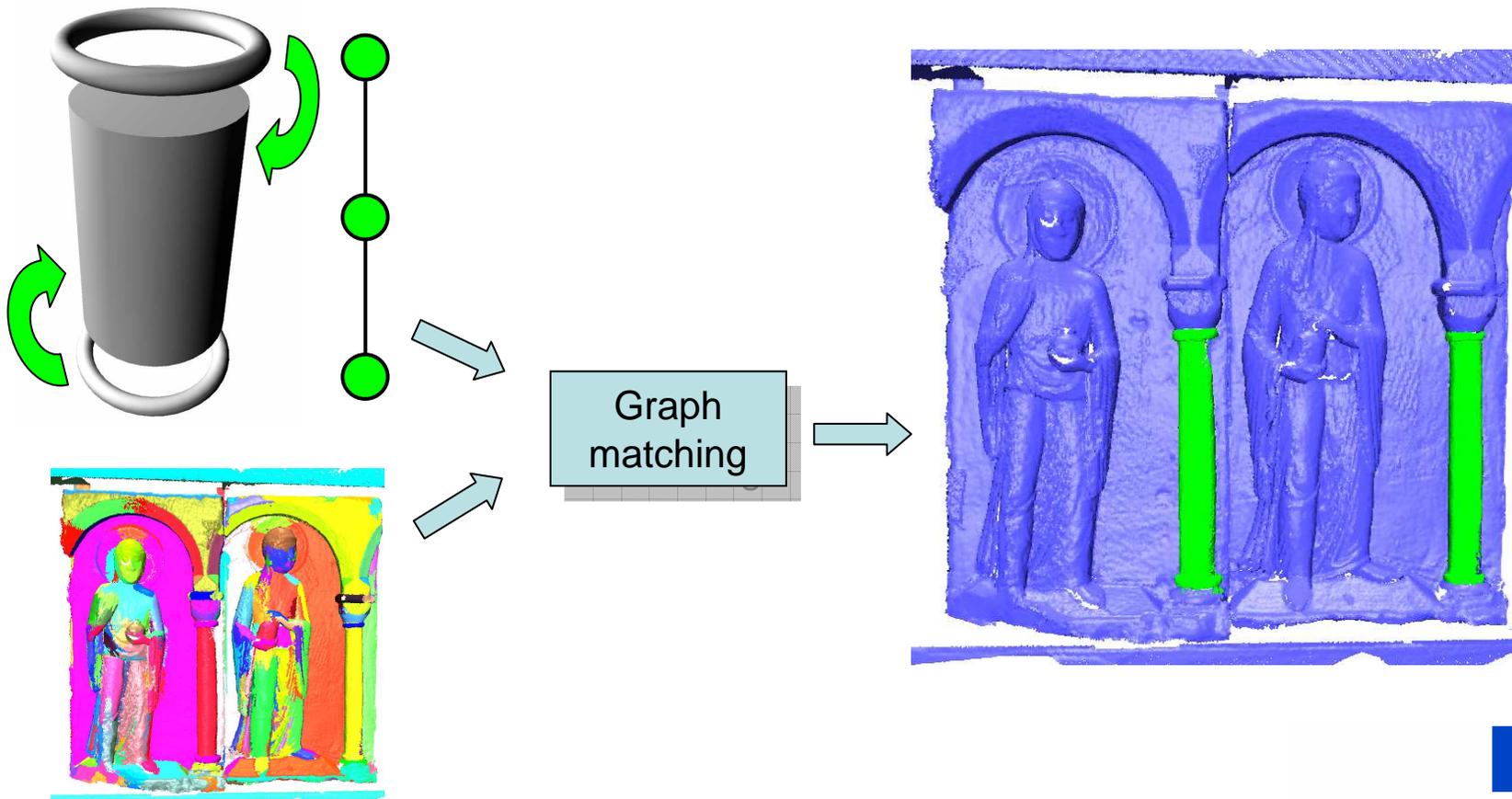
- Our approach:
 - Based on primitive shapes (plane, cylinder, cone,...)
 - Well suited for man-made objects
 - E.g.: column \approx cylinder
 - Decompose point-cloud into primitive shapes





Introduction

- User defines „query graph“
 - Desired configuration of primitive shapes
- Search for configuration in decomposed point-cloud
 - Graph matching





Related Work

- 3D city reconstruction
 - Verma, Kumar, Hsu: *3D Building Detection and Modeling from Aerial Lidar Data*, CVPR 2006
 - detection of roof types
 - configurations of planes (region growing)
 - limitations: only planes, no node and graph constraints
- Graph based shape matching
 - Zuckerberger, Tal, Shlafman: *Polyhedral surface decomposition with applications* Computers and Graphics, 2002
 - segmentation into *meaningful* units
 - unconstrained subgraph matching -> prohibitively time consuming
 - restricted to triangle meshes -> segmentation not stable

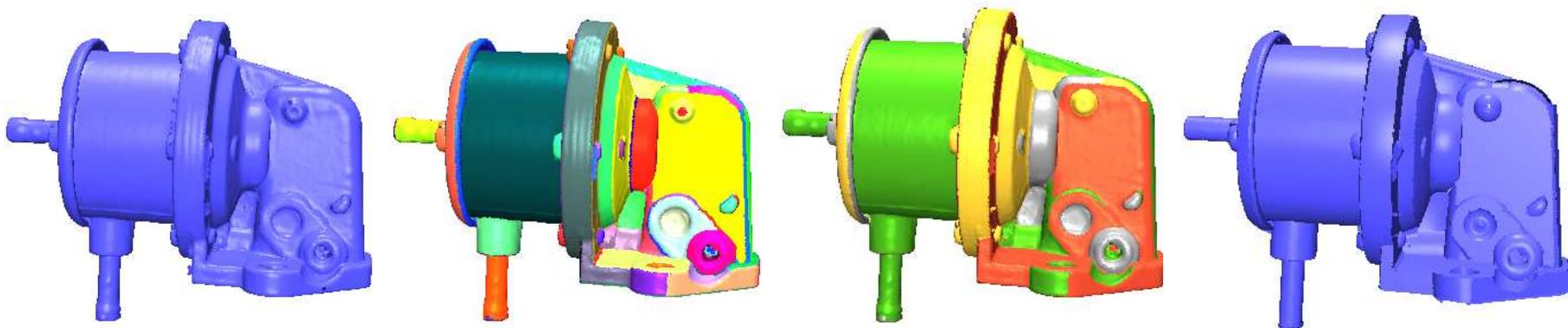


Point Cloud Decomposition

- Efficient RANSAC
- Detection of planes, cylinders, spheres, cones and tori
 - Fast, robust

$$P = S_{\phi_1} \cup \dots \cup S_{\phi_A} \cup R$$

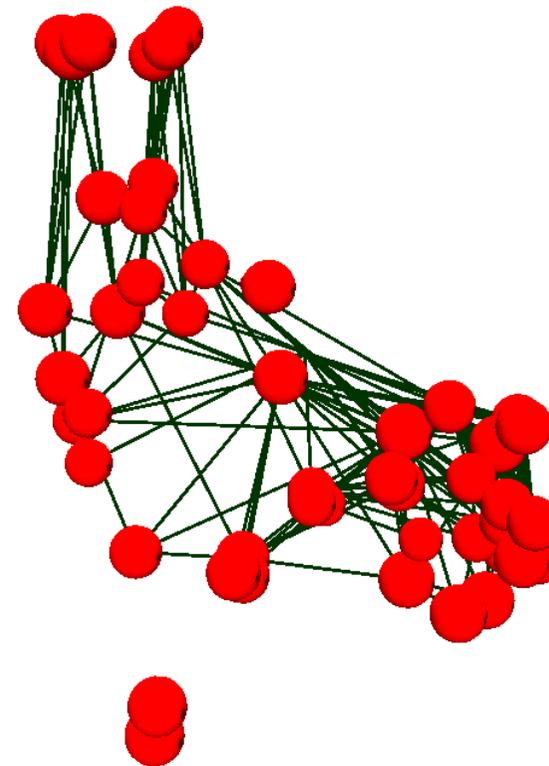
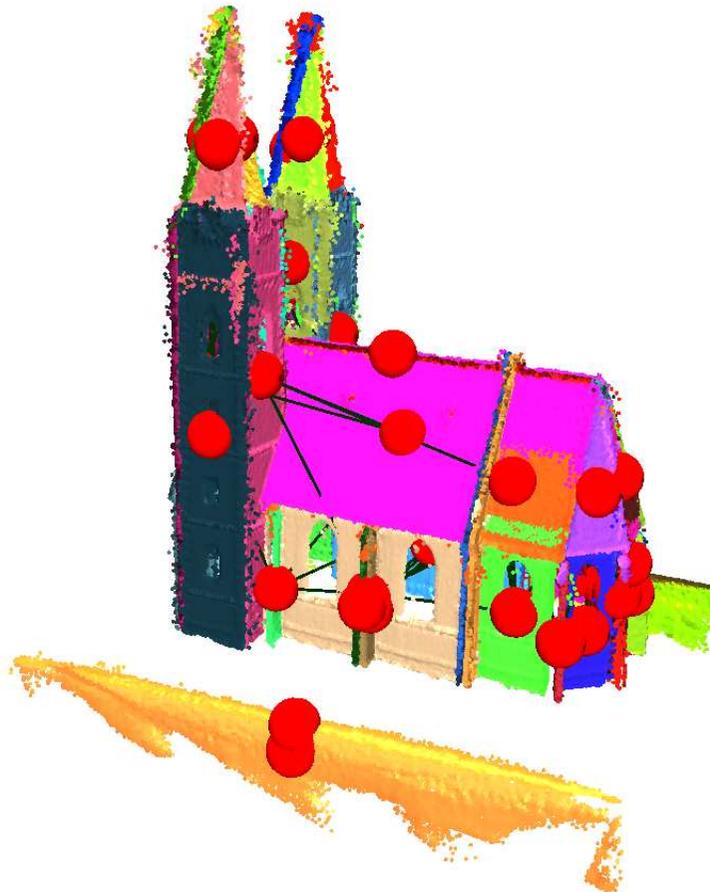
- Set of remaining points will be ignored
 - Outliers, noise, complex geometry





Topology Graph

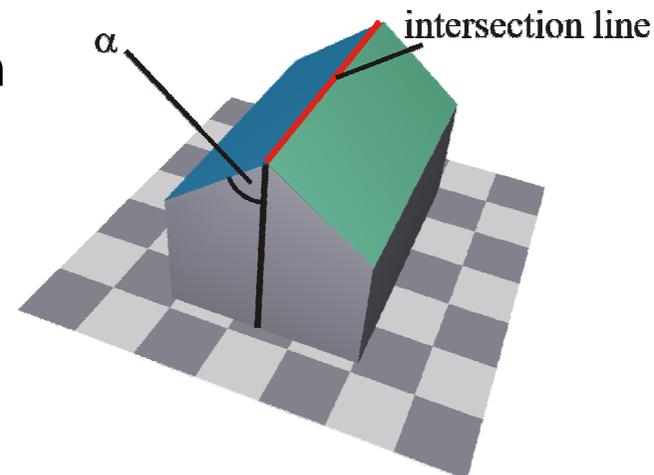
- Captures spatial relation between primitives
- Represent primitive shapes as **nodes**
- Neighboring nodes are connected by **edges**
 - controlled by proximity threshold t





Defining Query Graphs

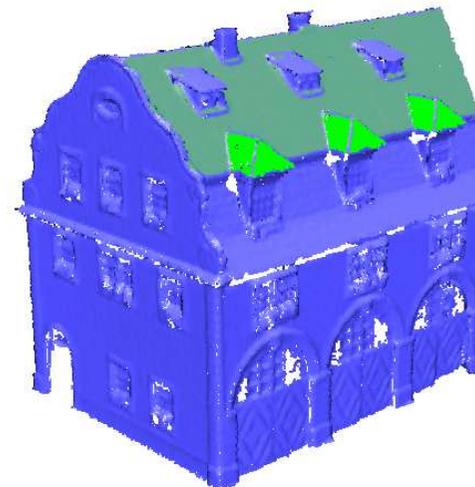
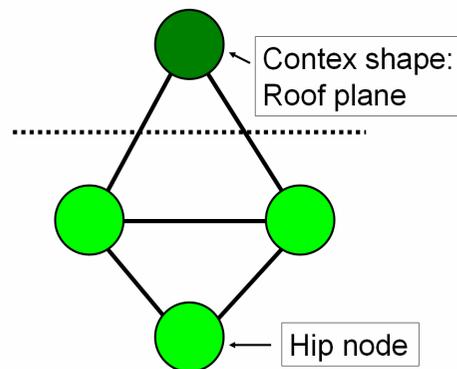
- Node Constraints
 - type of underlying primitive shape (e.g. plane, cylinder,...)
 - assign additional constraints manually (e.g. normals, angles, radius,...)
- Edge Constraints
 - describes spatial configuration between two shapes (e.g. angles, intersections, relative sizes,...)
- Graph Constraints
 - configuration constraints between unconnected nodes





Defining Query Graphs

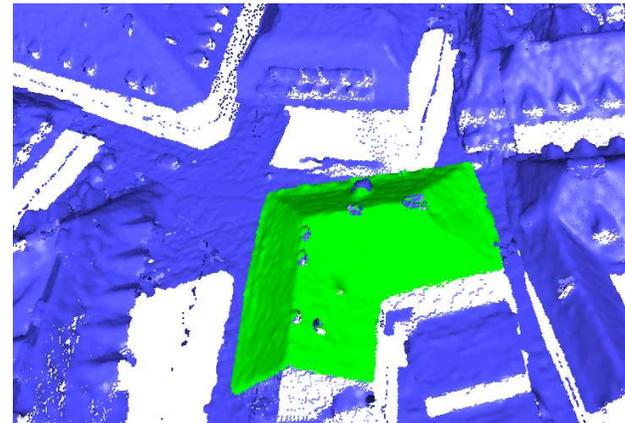
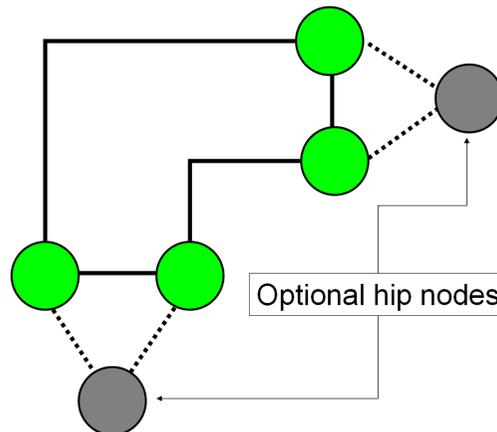
- Context Nodes
 - certain features benefit from *context objects* to distinguish them from other structures
 - model context objects as additional nodes in query graph
 - after matching, remove context nodes from result





Defining Query Graphs

- Optional Nodes
 - represent variants of one shape
- Matching
 - first match query graph without optional nodes
 - try to maximally extend results by optional nodes

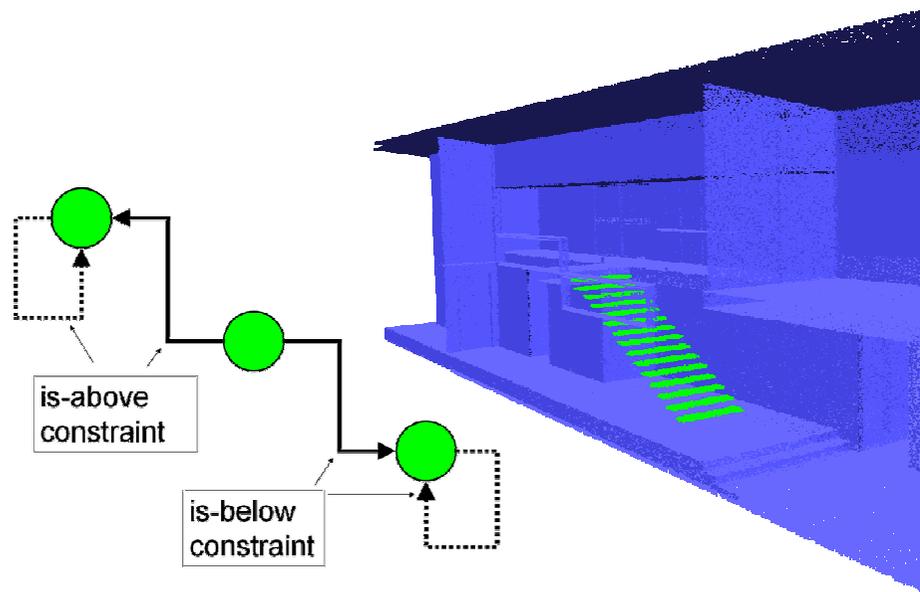


- Optional graph components
 - use graph constraints to assure completeness of component



Defining Query Graphs

- Multinodes
 - represent repetitive patterns in shapes (e.g. stairs)
 - modeled as self loops in the query graph



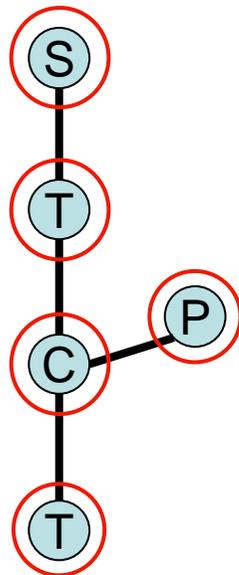
- use of directed edges
- restricted to one outgoing edge per multinode

Graph Matching

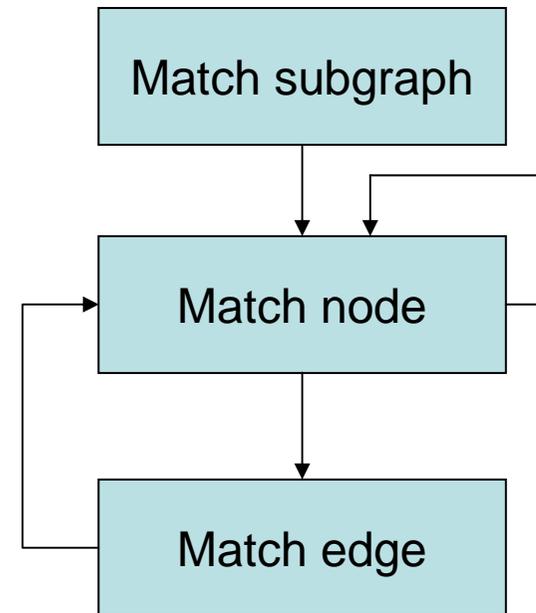
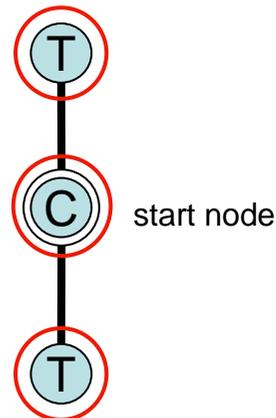


- Recursive graph matching
 - node and edge constraints are checked early on
 - graph constraints are checked when all nodes and edges have been correctly matched

Topology graph

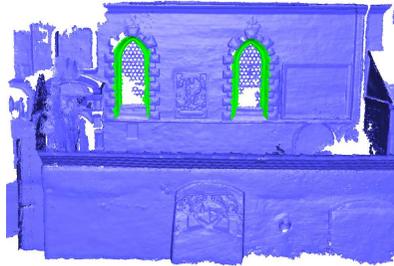


Query graph



checking edge constraints...
edge constraints match!!

Timings



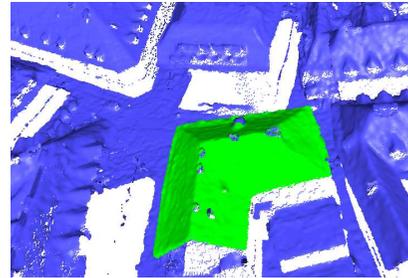
chapel

nodes: 232

edges: 406

graph construction: 1.8 sec

matching: < 10 msec



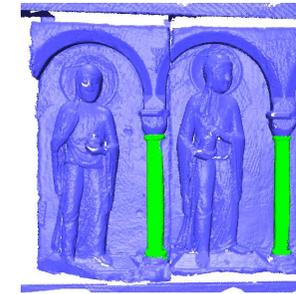
city model

nodes: 431

edges: 351

graph construction: 2.5 sec

matching: < 10 msec



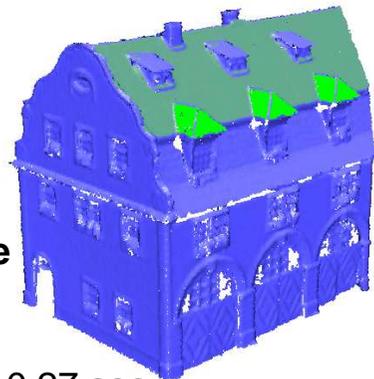
choir screen

nodes: 537

edges: 2731

graph construction: 1.8 sec

matching: < 10 msec



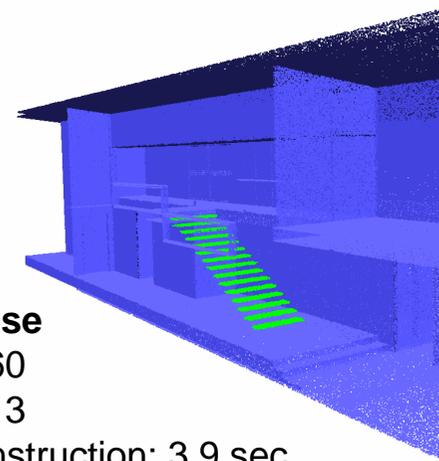
dormer roof house

nodes: 106

edges: 138

graph construction: 0.27 sec

matching: < 10 msec



CAD house

nodes: 160

edges: 513

graph construction: 3.9 sec

matching: < 10 msec



Conclusion

- Object detection in point clouds based on primitive shape decomposition
 - suited even for large amounts of data
 - applicable to nearly every data source like point-clouds, polygon soups, meshes
 - very fast due to constrained graph matching

- Limitations
 - restricted to objects consisting of primitive shapes
 - Not suited e.g. for natural objects like flowers or animals
 - large graphs might lead to degeneration of retrieval performance



Future Work

- User Interface
 - more comfortable interface for unexperienced users
 - easily define query graphs and constraints
- Compression
 - detect self-similarities
 - replace instances of query graphs by generic representations
- Editing
 - Apply basic point cloud operations like copy and paste of semantic units



Thanks for your attention !