Matching Geometric Models via Alignment

Alignment is the most common paradigm for matching 3D models to either 2D or 3D data. The steps are:

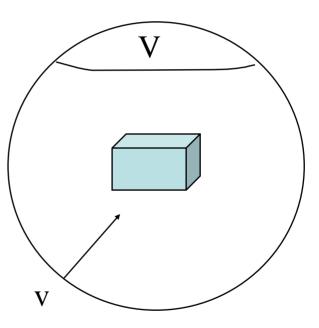
- 1. **hypothesize a correspondence** between a set of model points and a set of data points
- 2. From the correspondence **compute a transformation** from model to data
- 3. **Apply the transformation** to the model features to produce transformed features
- 4. **Compare** the transformed model features to the image features to verify or disprove the hypothesis

2D-3D Alignment

- single 2D images of the objects
- 3D object models
 - full 3D models, such as GC or SEV
 - view class models representing characteristic views of the objects

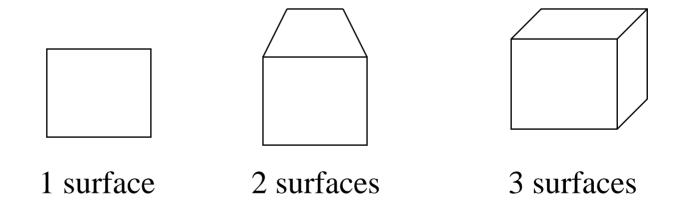
View Classes and Viewing Sphere

- The space of view points can be partitioned into a finite set of characteristic views.
- Each view class represents a set of view points that have something in common, such as:



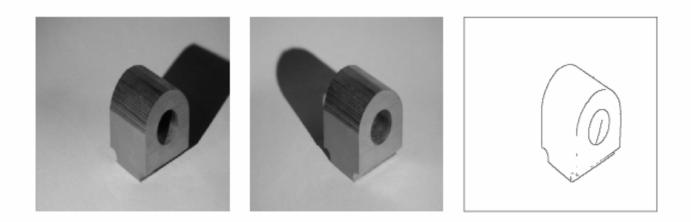
- 1. same surfaces are visible
- 2. same line segments are visible
- 3. relational distance between pairs of them is small

3 View Classes of a Cube



RIO: Relational Indexing for Object Recognition

- RIO worked with industrial parts that could have
 - planar surfaces
 - cylindrical surfaces
 - threads

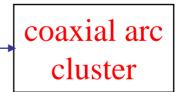


Object Representation in RIO

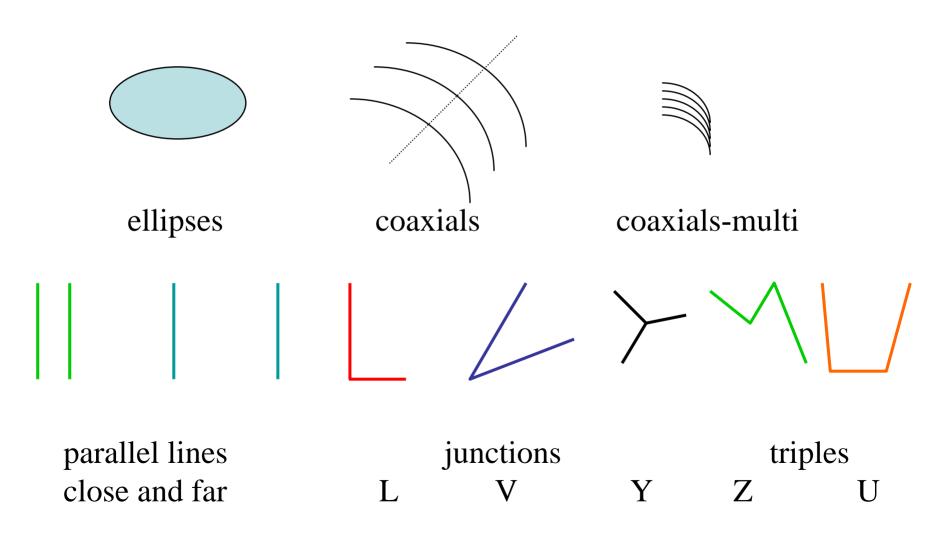
- 3D objects are represented by a 3D mesh and set of 2D view classes.
- Each view class is represented by an attributed graph whose nodes are features and whose attributed edges are relationships.
- For purposes of indexing, attributed graphs are stored as sets of 2-graphs, graphs with 2 nodes and 2 relationships.



share an arc

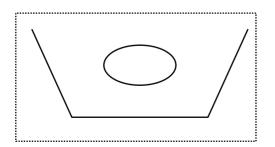


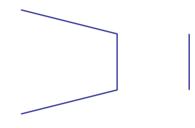
RIO Features



RIO Relationships

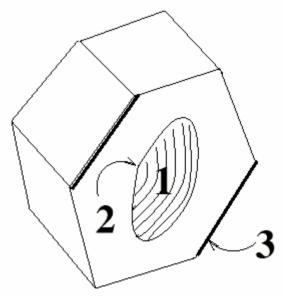
- share one arc
- share one line
- share two lines
- coaxial
- close at extremal points
- bounding box encloses / enclosed by





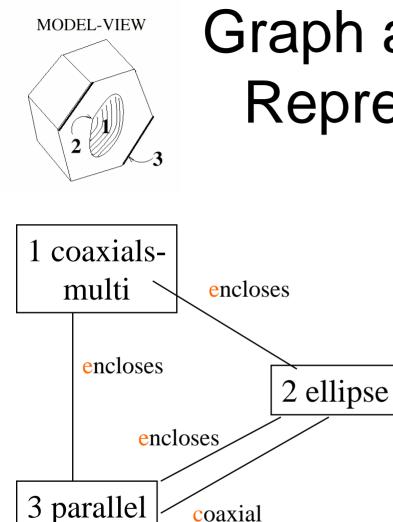
Hexnut Object

MODEL-VIEW

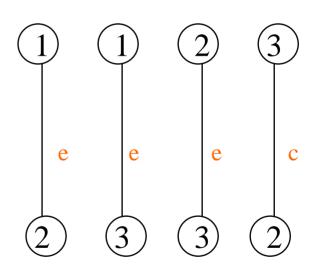


RELATIONS: a: encloses b: coaxial What other features and relationships can you find?

FEATURES: 1: coaxials-multi 2: ellipse 3: parallel lines



lines



Graph and 2-Graph Representations

Relational Indexing for Recognition

Preprocessing (off-line) Phase

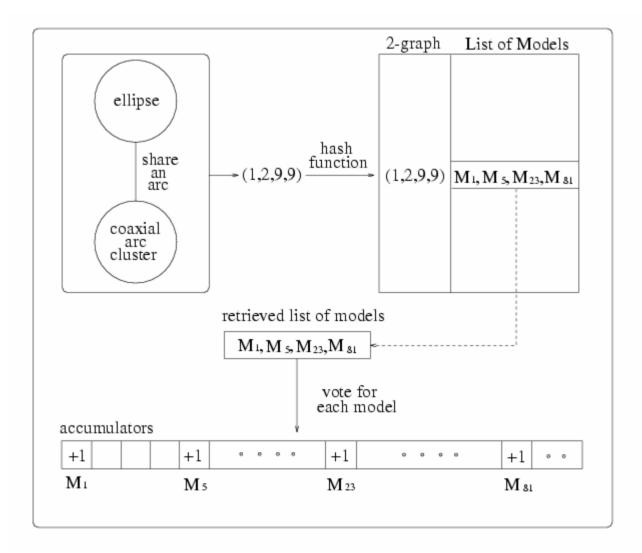
for each model view M_i in the database

- encode each 2-graph of M_i to produce an index
- store M_i and associated information in the indexed bin of a hash table H

Matching (on-line) phase

- 1. Construct a relational (2-graph) description D for the scene
- 2. For each 2-graph G of D
 - encode it, producing an index to access the hash table H
 - cast a vote for each M_i in the associated bin
- 3. Select the M_i s with high votes as possible hypotheses
- 4. Verify or disprove via alignment, using the 3D meshes

The Voting Process



Verification

- 1. The matched features of the hypothesized object are used to determine its **pose**. Pose is computed from correspondences between 2D and 3D points, lines, and circles.
- 2. The **3D** mesh of the object is used to project all its features onto the image using perspective projection and hidden feature removal.
- 3. A verification procedure checks how well the object features line up with edges on the image, using a Hausdorf distance between expected and existing edges.

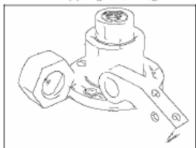
Feature Extraction



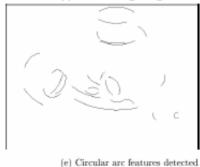
(a) Original left image

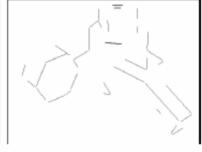


(b) Original right image

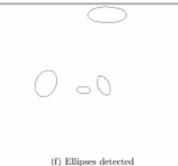


(c) Combined edge image

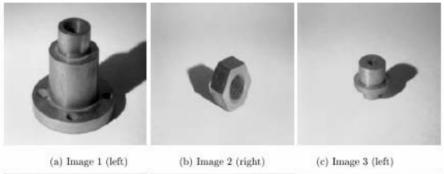




(d) Linear features detected



Some Test Scenes





(d) Image 4 (left)

(e) Image 5 (left)



(f) Image 6 (right)



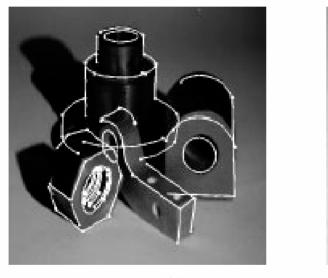
(g) Image 7 (left)

(h) Image 8 (right)

(i) Image 9 (right)

Sample Alignments 3D to 2D Perspective Projection

(b)



(a)

RIO Verifications

incorrect hypothesis



