

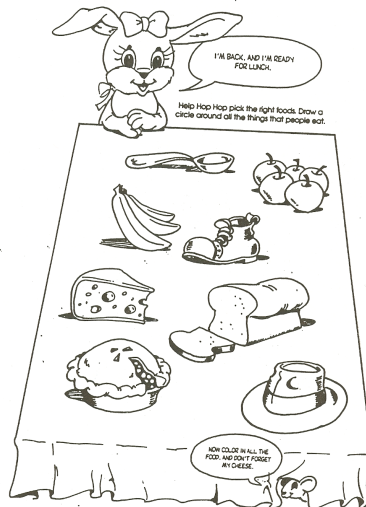
High-Level Vision

Object Recognition



CS332 Visual Processing
Department of Computer Science
Wellesley College

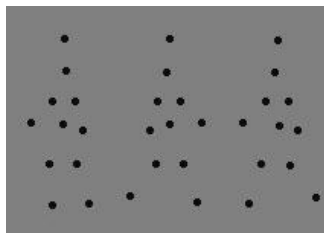
Recognition from geometric shape



Objects can be recognized easily from the shape of their image contours

Very young children can easily recognize a wide variety of common objects

Other recognition cues



characteristic motion



color



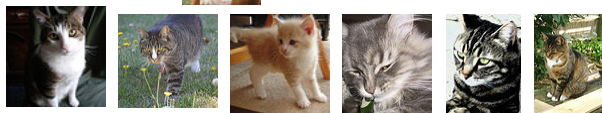
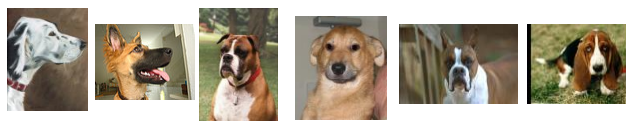
texture



context



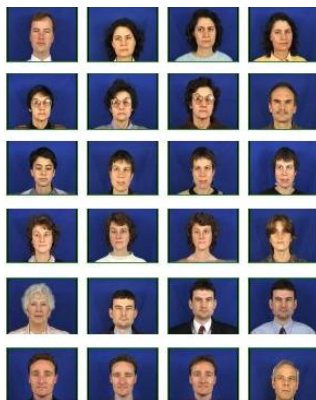
Why is recognition difficult?



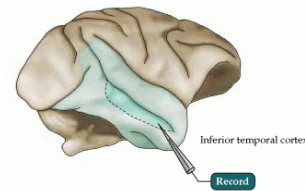
What is a chair?




Face recognition




AR Face Database



← → ↻ www.faceplusplus.com/demo-detect/

 [SIGN IN](#)

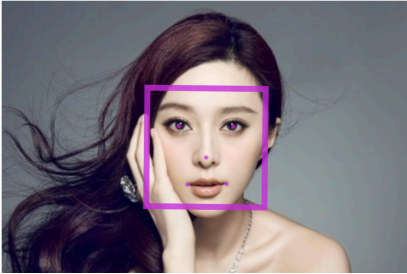
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[Demo](#)

- [Face Detection](#) >
- [Face Search](#) >
- [Face Landmark](#) >
- [Face Mask](#) >
- [Interactive Demo](#) >

Tips:



REST URL:
`http://apius.faceplusplus.com/api_key=DEMO_KEY&api_secret=2F%2Fwww.faceplusplus.com%aceplusplus%2Fassets%2Fimg`

RESPONSE JSON:

```
{
  "face": [
    {
      "attribute": {
        "age": {
          "range": 5,
          "value": 24
        },
        "gender": {
```

Early approaches to recognition...

... differ in how *regularities* are used to constrain the interpretation of the viewed object

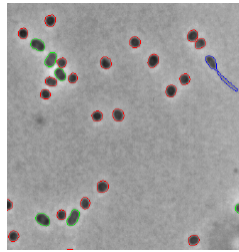
Three main approaches:

- invariant properties
- parts decomposition
- alignment

Invariant properties

Every instance of each object class exhibits certain properties

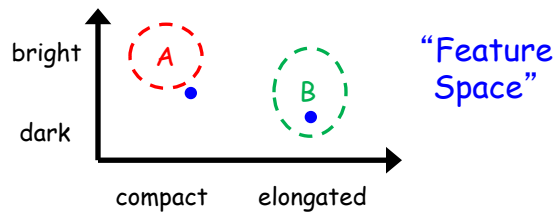
- (1) measure properties of viewed object
- (2) apply decision procedure



classify cells in culture

Properties:

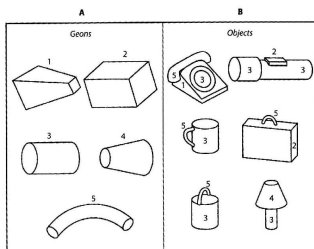
- (1) ratio: perimeter/area
- (2) brightness



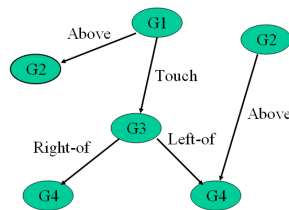
Parts decomposition

Every instance of each object class shares certain parts, arranged in a certain way

- (1) find object parts
- (2) recognize objects by presence of parts with proper relationship



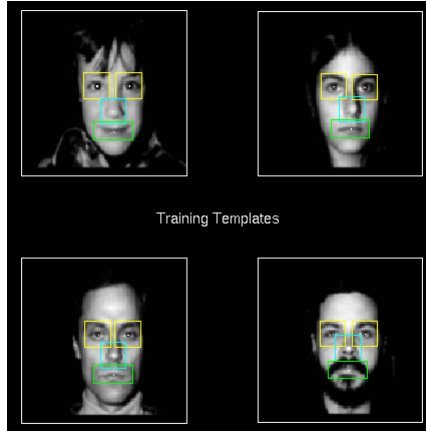
Biederman's Geons



"Structural Description"

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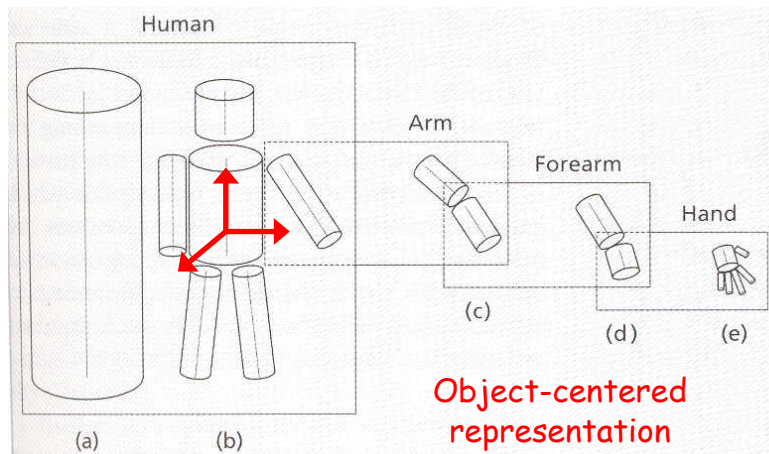
Face recognition by parts decomposition



MIT Media Lab Vision & Modeling Group

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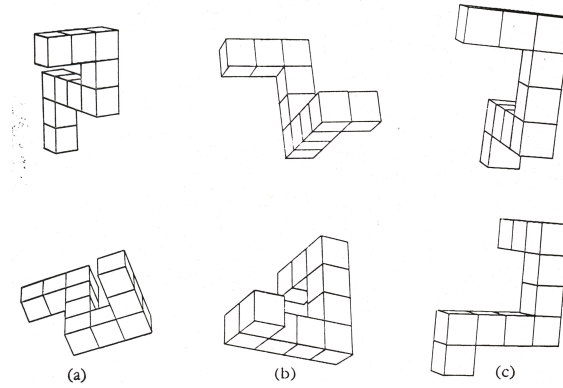
Feature hierarchies



Marr & Nishihara

1-14

Mental rotation

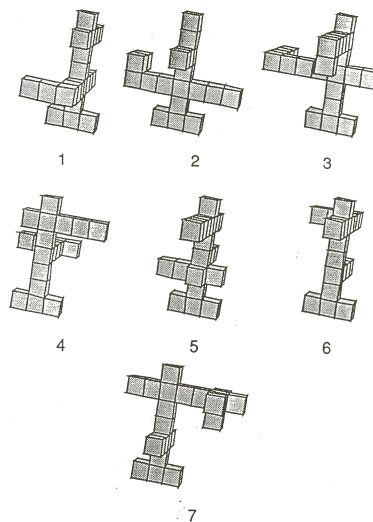


Time needed to determine whether pair of objects are *the same* is proportional to angle of rotation between pair

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Viewer-centered object representation?

Tarr, '95: After learning to recognize 3-D objects from a small set of specific 2-D views, the time needed to recognize a novel view is proportional to the 3-D angle between the new view and closest learned view



1-16

The debate continues...

Viewpoint invariant
object representations



Viewpoint dependent
object representations



Biederman

Object-centered!

Viewer-centered!



Bülhoff

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Alignment methods

Find an object model and geometric transformation
that *best match* the viewed image

V viewed object (image)

M_i object models

T_{ij} allowable transformations between viewed
object and models

F measure of fit between V and the expected
appearance of model M_i under the
transformation T_{ij}

GOAL: Find a combination of M_i and T_{ij} that
maximizes the fit F

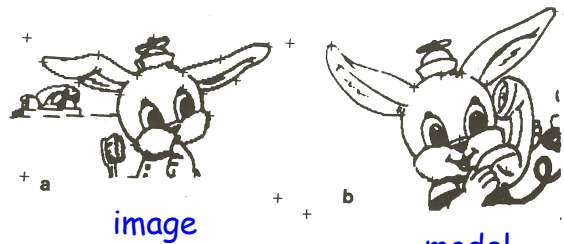
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Alignment method: recognition process

- (1) Find best transformation T_{ij} for each model M_i (optimizing over possible views)
- (2) Find M_i whose best T_{ij} gives the best match to image V

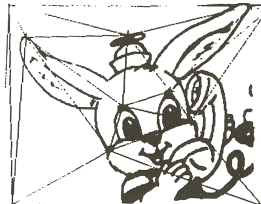
1-19

Aligning pictorial models



image

model



triangulated model



transformed model
superimposed on image

1-20

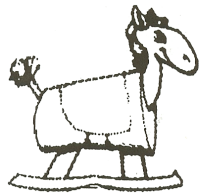
When the model doesn't fit...



image



model



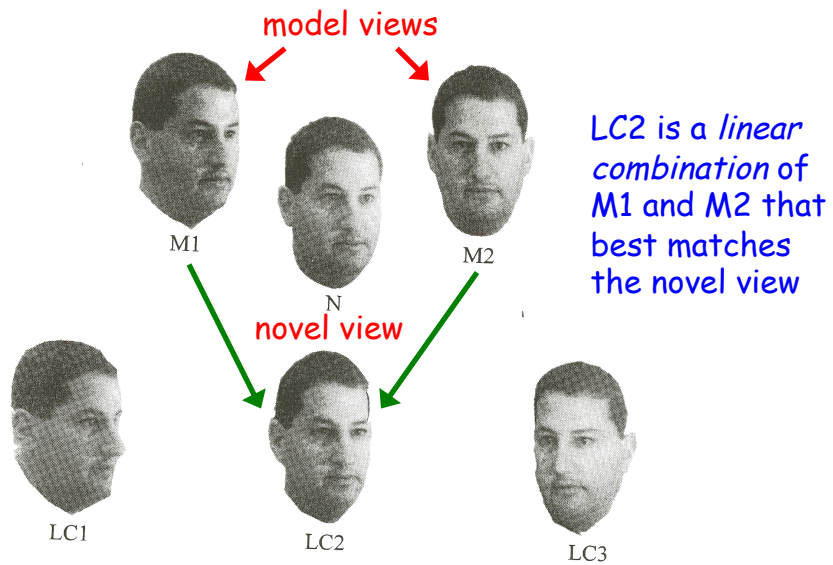
transformed
model



transformed model
superimposed on image

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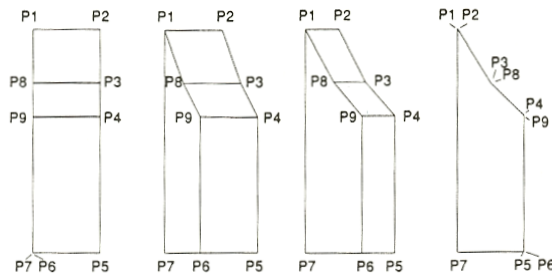
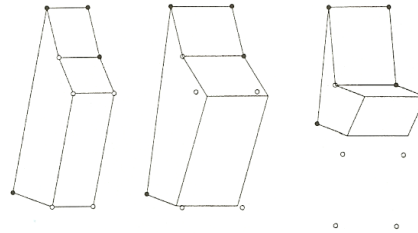
Recognition by linear combination of views



1-22

Obelisk, jukebox or seat?

Each object model consists of multiple 2-D views

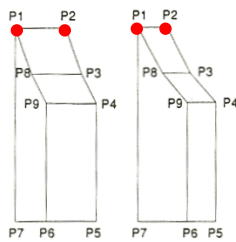


obelisk model

Goal: recognize novel views of these objects

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Predicting object appearance



I_1

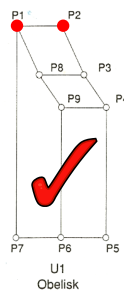
I_2

two known views of obelisk

$$X_{P1I_0} = \alpha X_{P1I_1} + \beta X_{P1I_2}$$

$$X_{P2I_0} = \alpha X_{P2I_1} + \beta X_{P2I_2}$$

I_0



U1
Obelisk

I_0



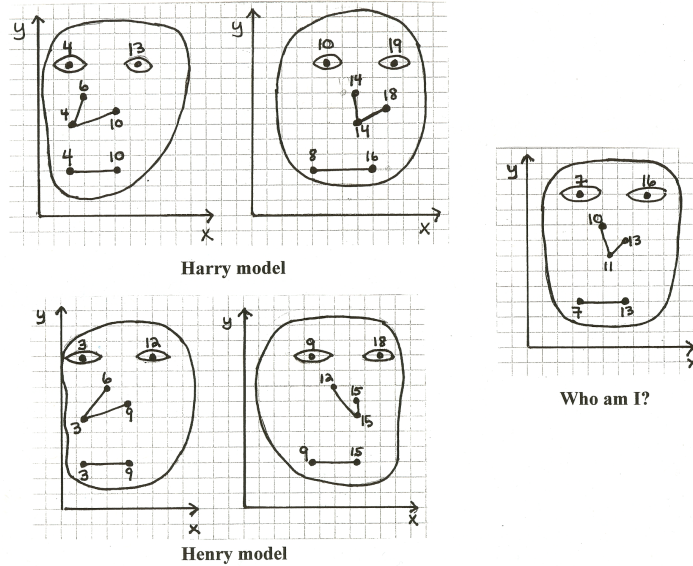
U2
Jukebox

Recognition process:

- (1) compute α, β that predict P1 and P2
- (2) use α, β to predict other points
- (3) evaluate fit of model to image

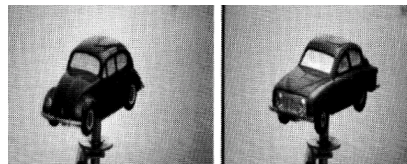
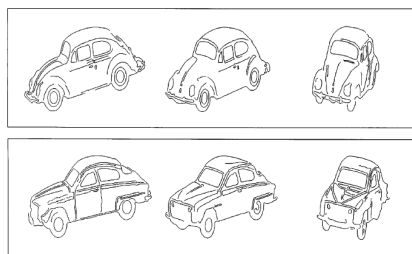
1-24

Face recognition by linear combination of views



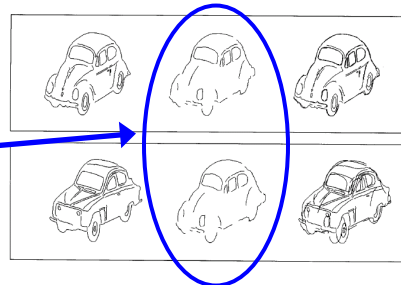
1-25

Ullman & Basri



Object models: edge maps from multiple 2D views

Template: linear combination of locations of edge points from model views that "best fits" edge map from image of unknown object



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