

# Computer Vision at Cambridge:

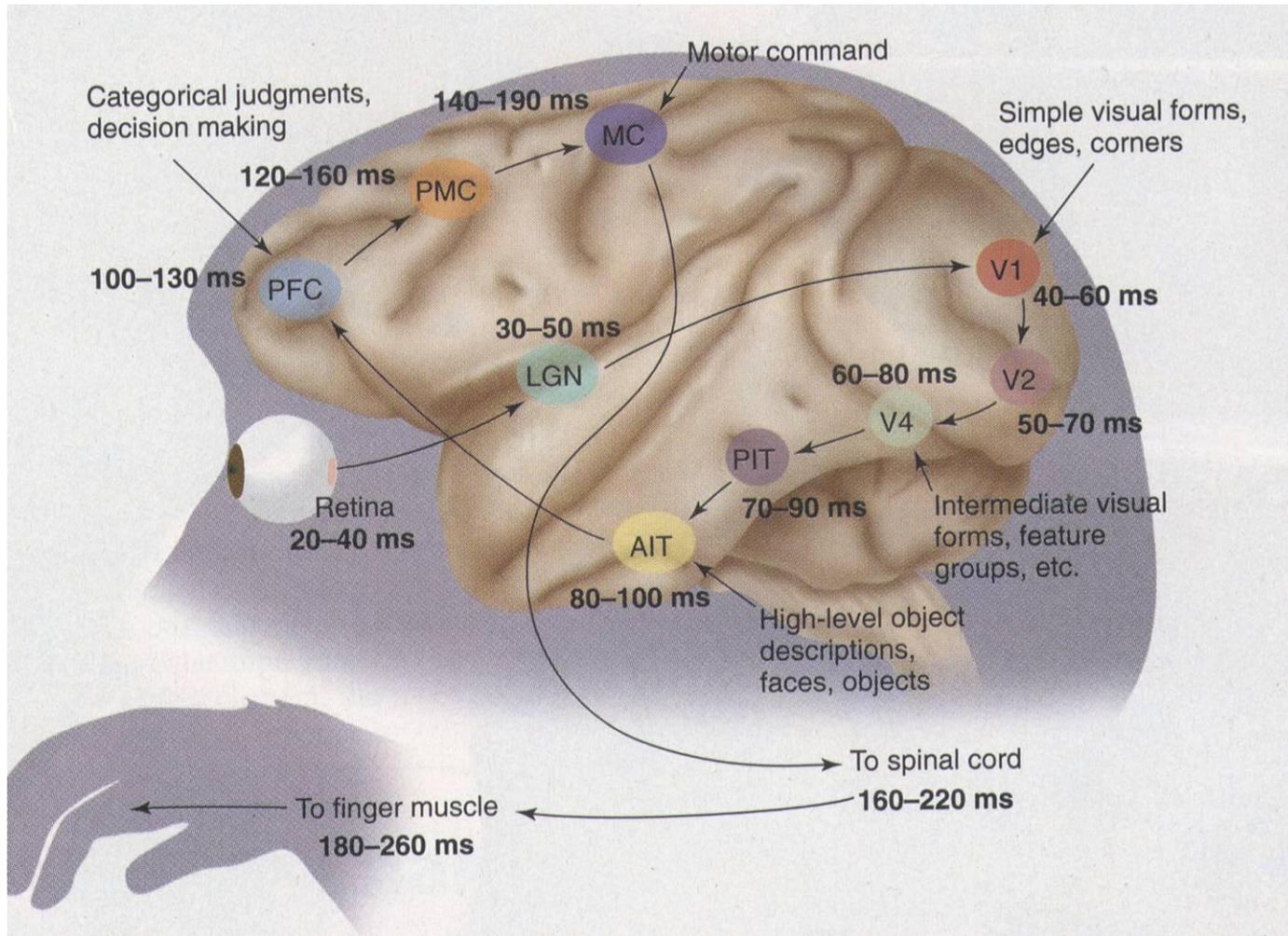
## Reconstruction, Registration and Recognition

Roberto Cipolla

Research team

<http://www.eng.cam.ac.uk/~cipolla/people.html>

# Cognitive Systems Engineering



# New applications

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Computer vision has now found a place in consumer products

- Mobile phones and PDAs
- Games
- Cars
- Image and video search
- Internet and shopping

# Smart erase on a mobile phone

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# I. Reconstruction:

## Recovery of accurate 3D shape from uncalibrated images

Cipolla and Blake 1992

Cipolla and Giblin 1999

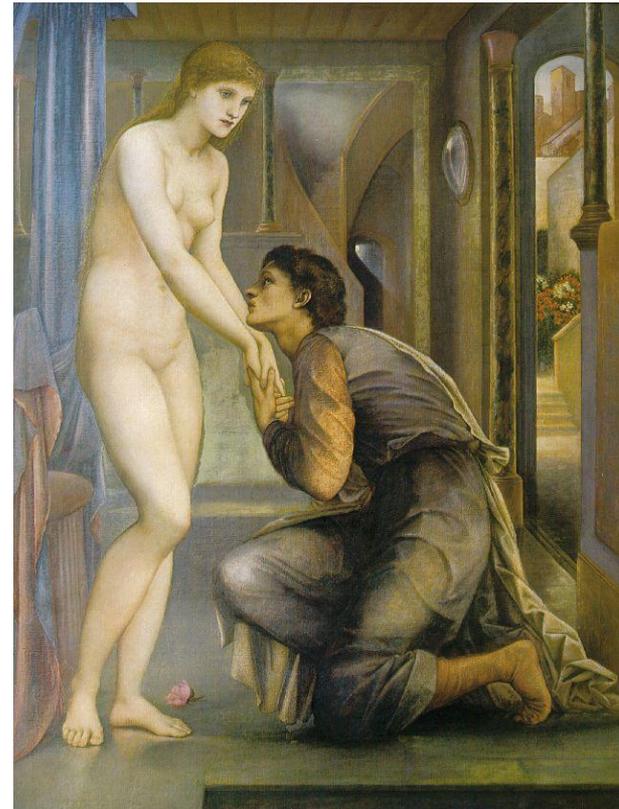
Mendonca, Wong and Cipolla 1999-2005

Vogiatzis, Hernandez and Cipolla 2006-2007

Hernandez, Brostow and Cipolla 2007

# Digital Pygmalion – the myth

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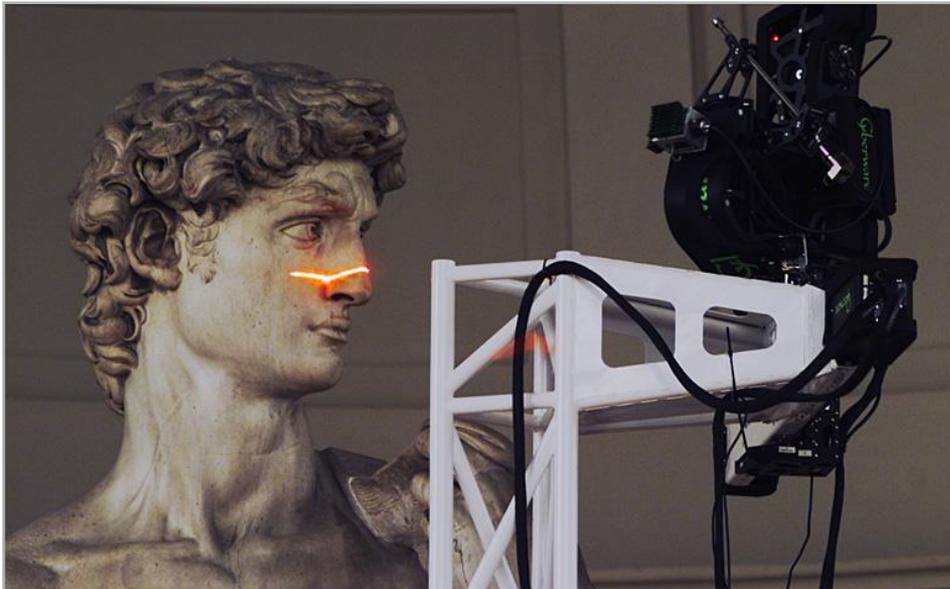
# Digital Pygamlion Project



# Scanning technologies

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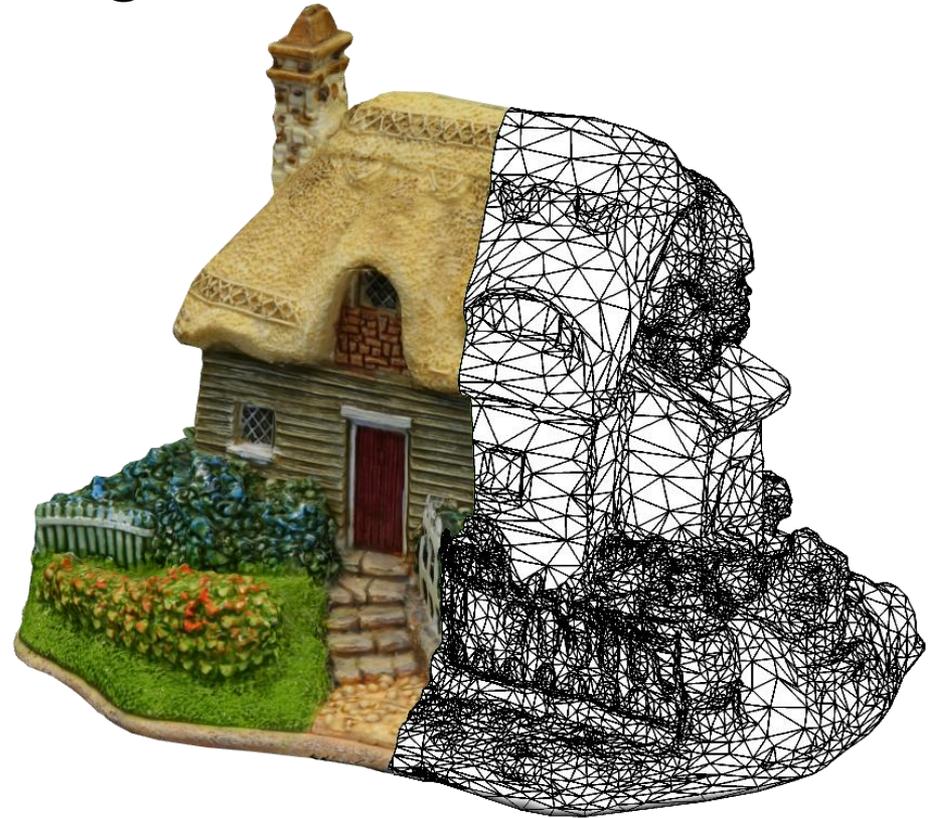
- Laser range finders
  - Very accurate
  - Very expensive
  - Complicated to use



# 3D models

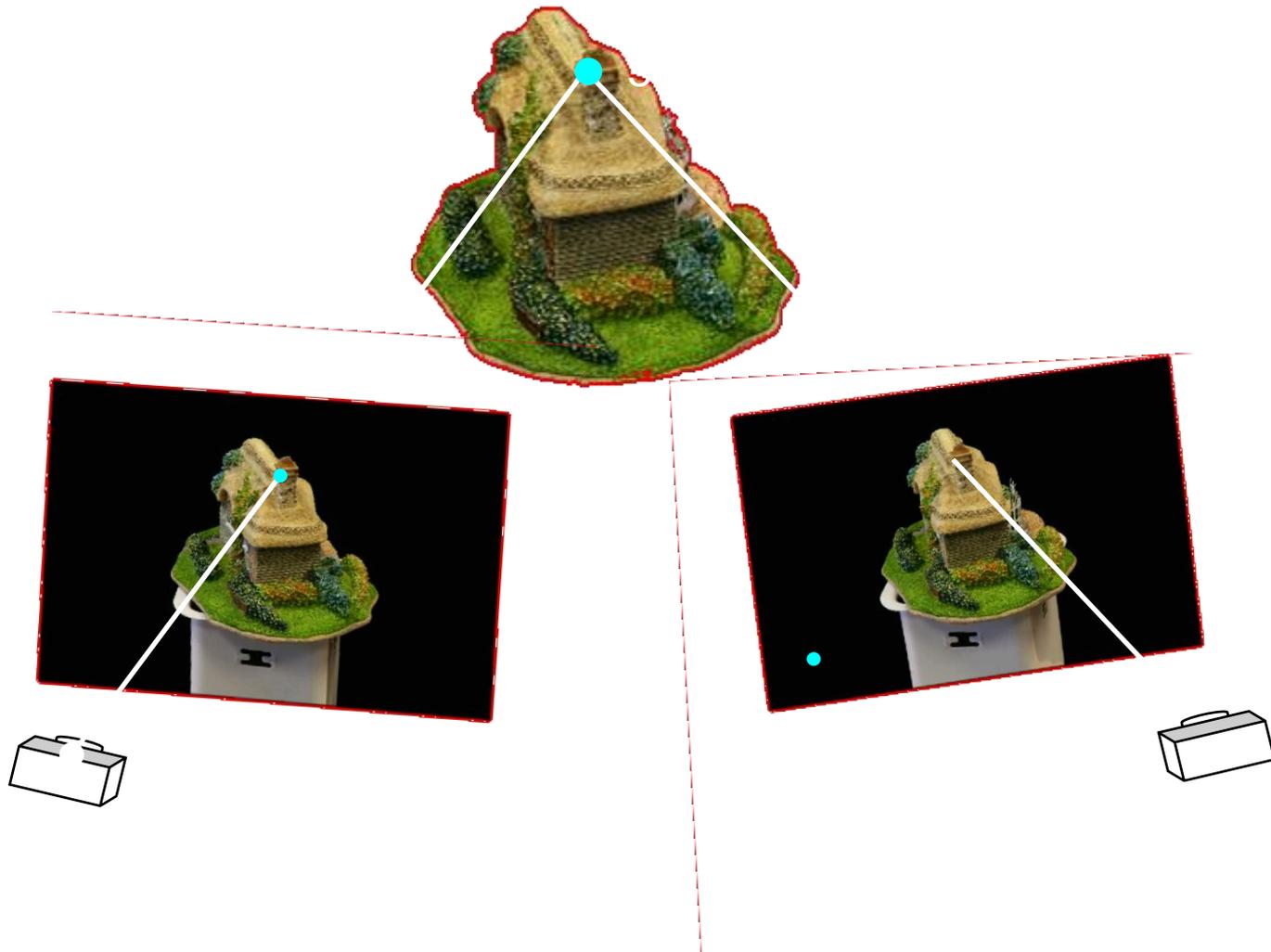
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- We need a way to get them that is
  - practical
  - fast
  - non-intrusive
  - low cost



# Stereo vision

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# 3D Models



# 1. Camera motion and multiview stereo

Cipolla and Blake 1992

Cipolla and Giblin 1999

Mendonca, Wong and Cipolla 1999-2005

Hernandez, Schmidt and Cipolla 2007

Vogiatzis, Hernandez and Cipolla 2007

# Input Images

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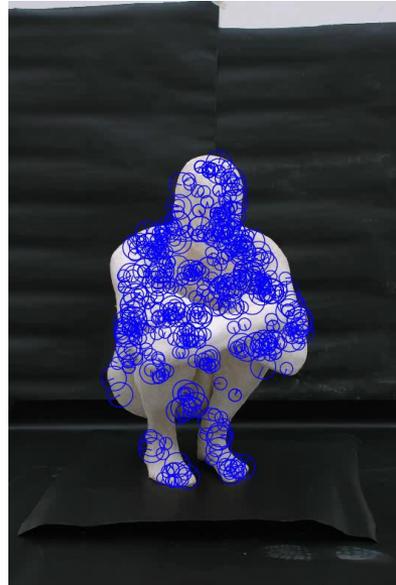


# Recovery of camera motion

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Input images



Feature  
extraction



Feature  
matching

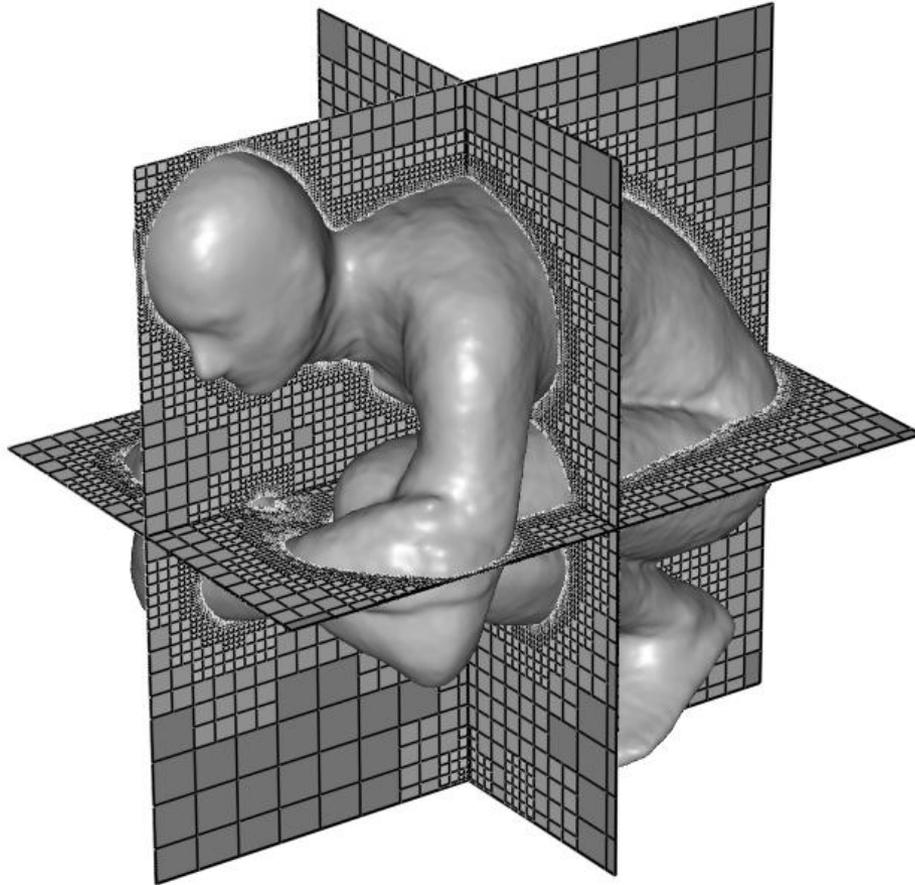


Bundle  
adjustment

# 2. Probabilistic 3D segmentation using graph cuts

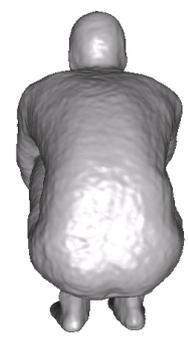
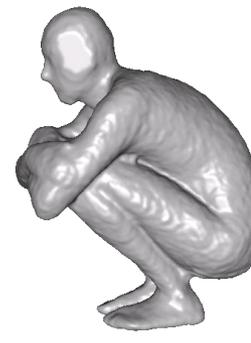
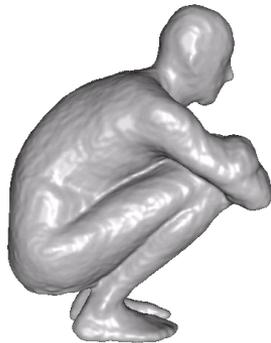
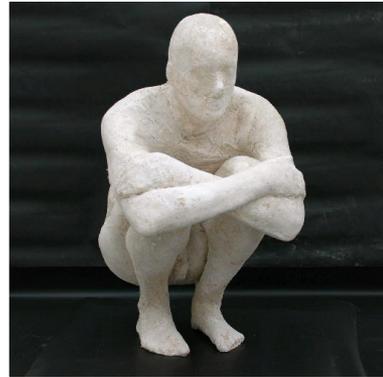
Vogiatzis, Hernandez and Cipolla 2007

# Probabilistic 3D segmentation UNIVERSITY OF CAMBRIDGE



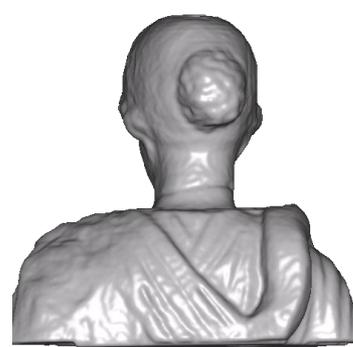
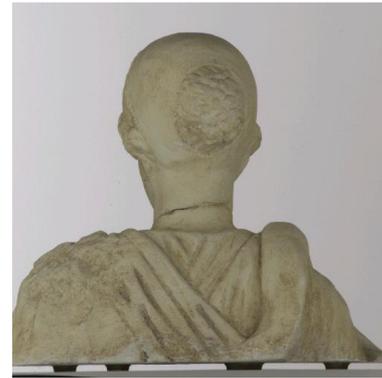
# 3D models

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# 3D models

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# Advantages

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- Low cost
- Non intrusive
- Accurate
- Simple
  - Can work with about 5-30 images
- Fast
  - Approximately 15-30 minutes of computation for these models
  - We believe we can bring this down to minutes

# Application requirements

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In order of importance

1. Simplicity ~25 images, no calibration
2. Cost Digital camera + PC
3. Visual accuracy Sub-pixel
4. Intrusiveness Completely non-intrusive
5. Speed Recently down to 15min
6. Robustness transparency, deformations
7. Geometric accuracy 1 part in 1000

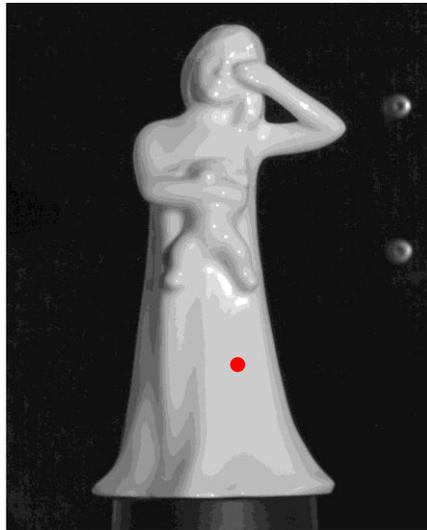
# 3. Multiview photometric stereo

Vogiatzis, Hernandez and Cipolla 2006

# Untextured objects

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- Almost impossible to establish correspondence



# Untextured objects

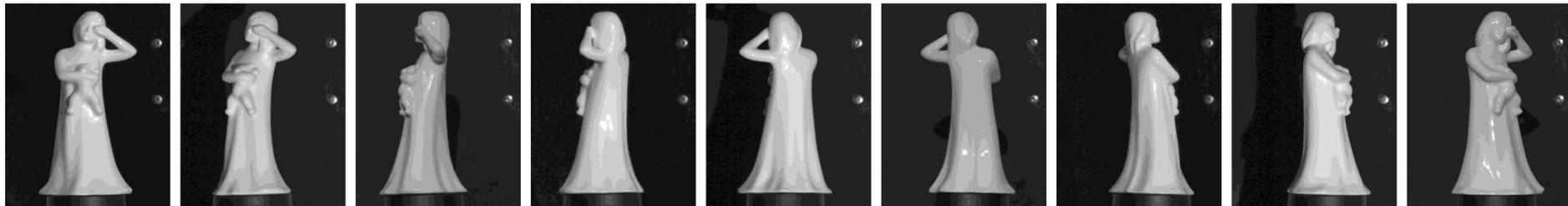
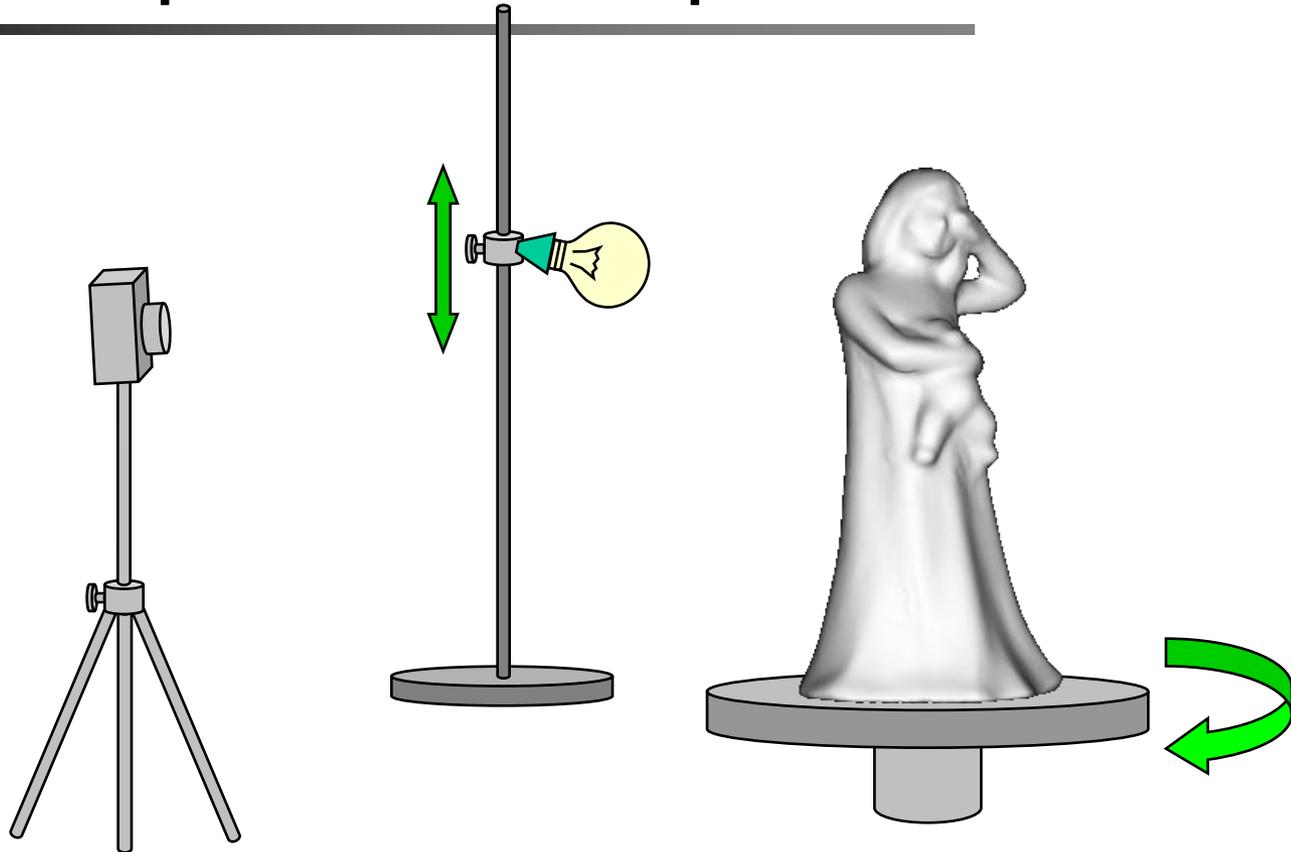
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Changing lighting uncovers fine geometric detail

- Assumptions:
  - Single, distant light-source
  - Silhouettes can be extracted
  - No texture, single colour

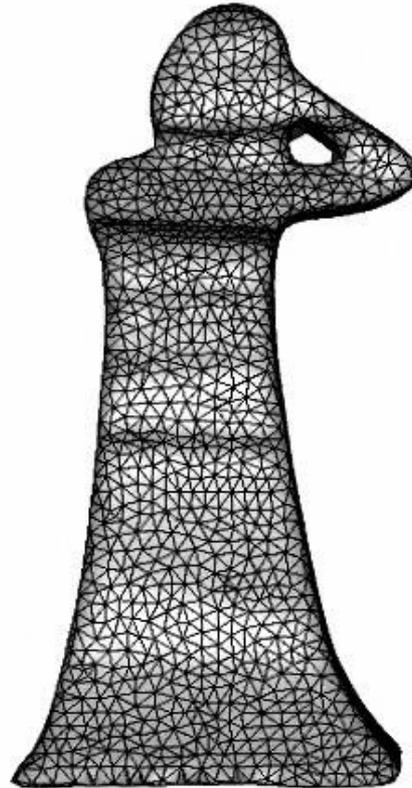


# Image acquisition setup



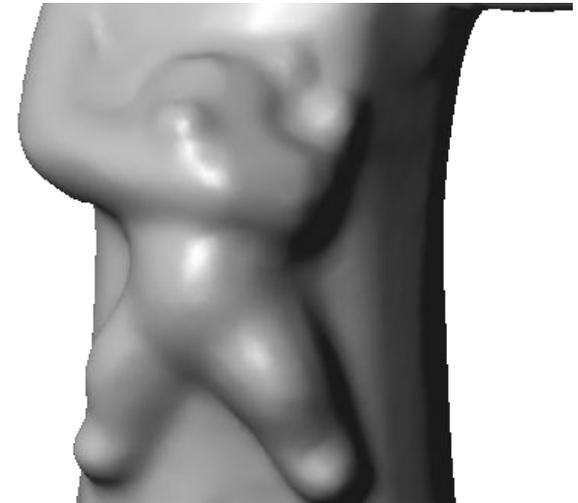
# Surface Evolution: 3D Mesh

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# 3D Models

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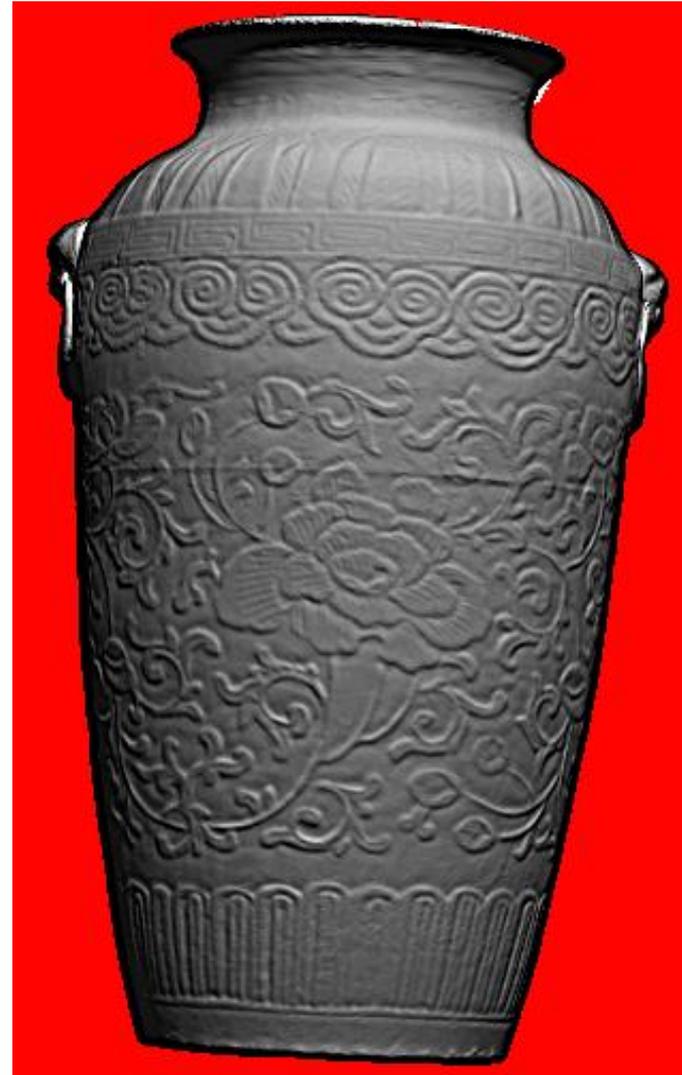
# 3D Models

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# 3D Models

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# 3D Models

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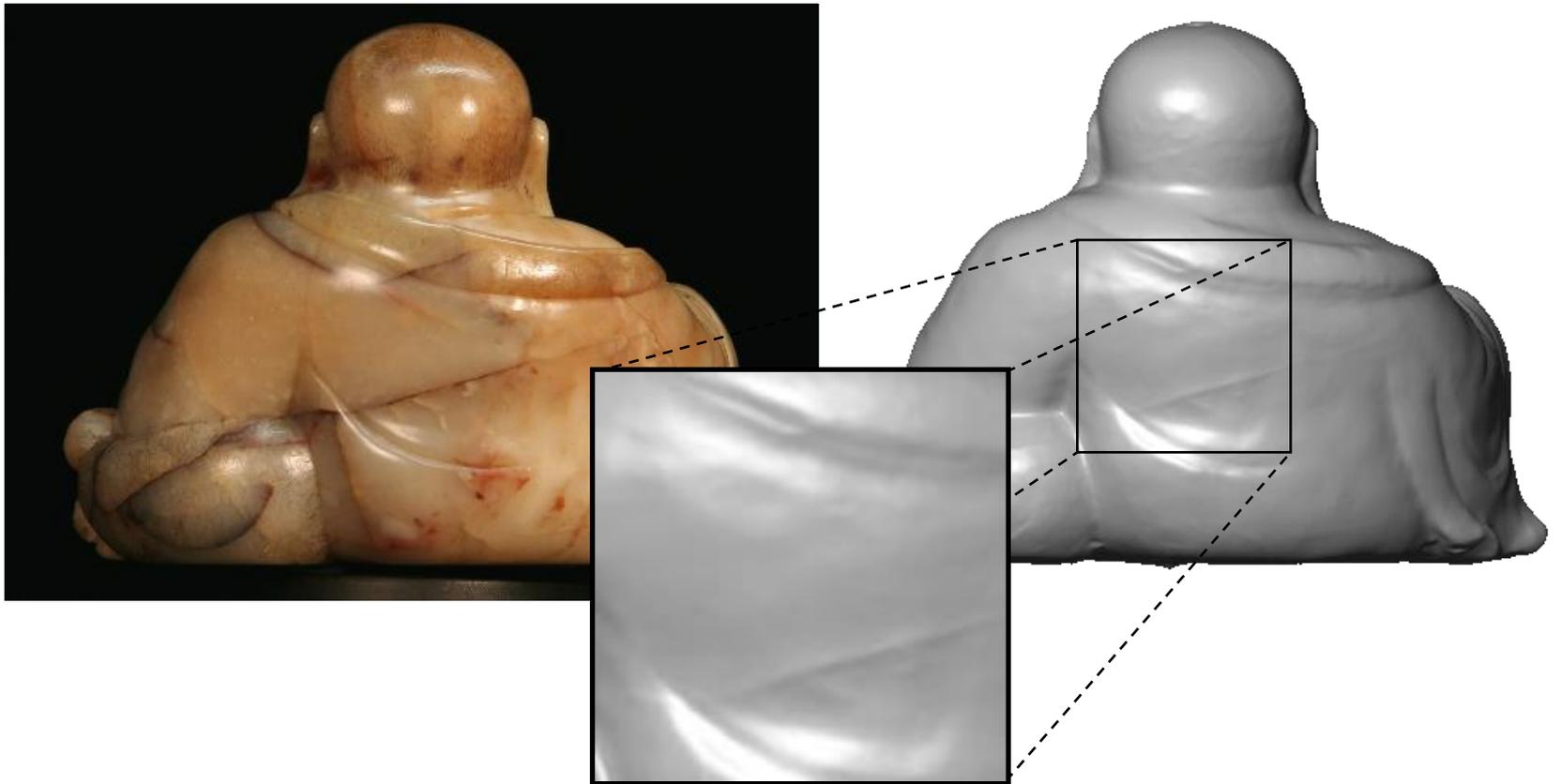
# 3D Models

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# 3D Models

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## II. Registration:

Target detection and pose  
estimation

## 4. Registration:

Where am I?

What am I looking at?

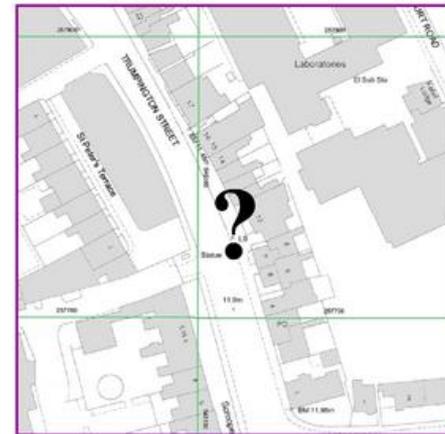
Johansson and Cipolla 2002

Robertson and Cipolla 2004

# Where I am?



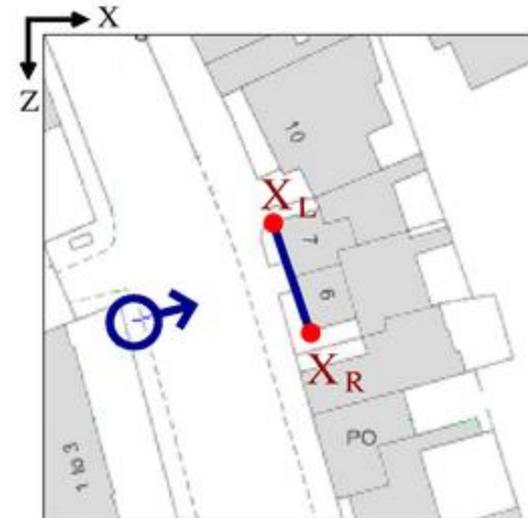
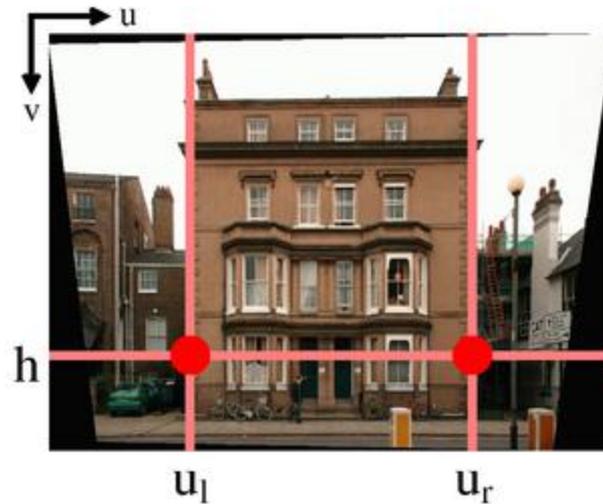
?  
=



Determine pose from single image by matching

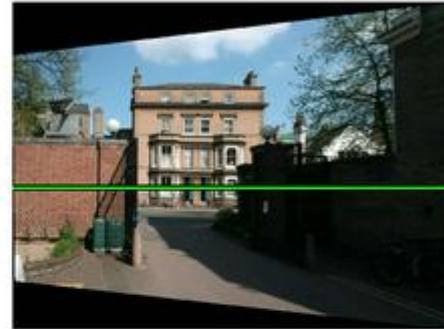
# Register database view

First align database view to map



# Registration of input image

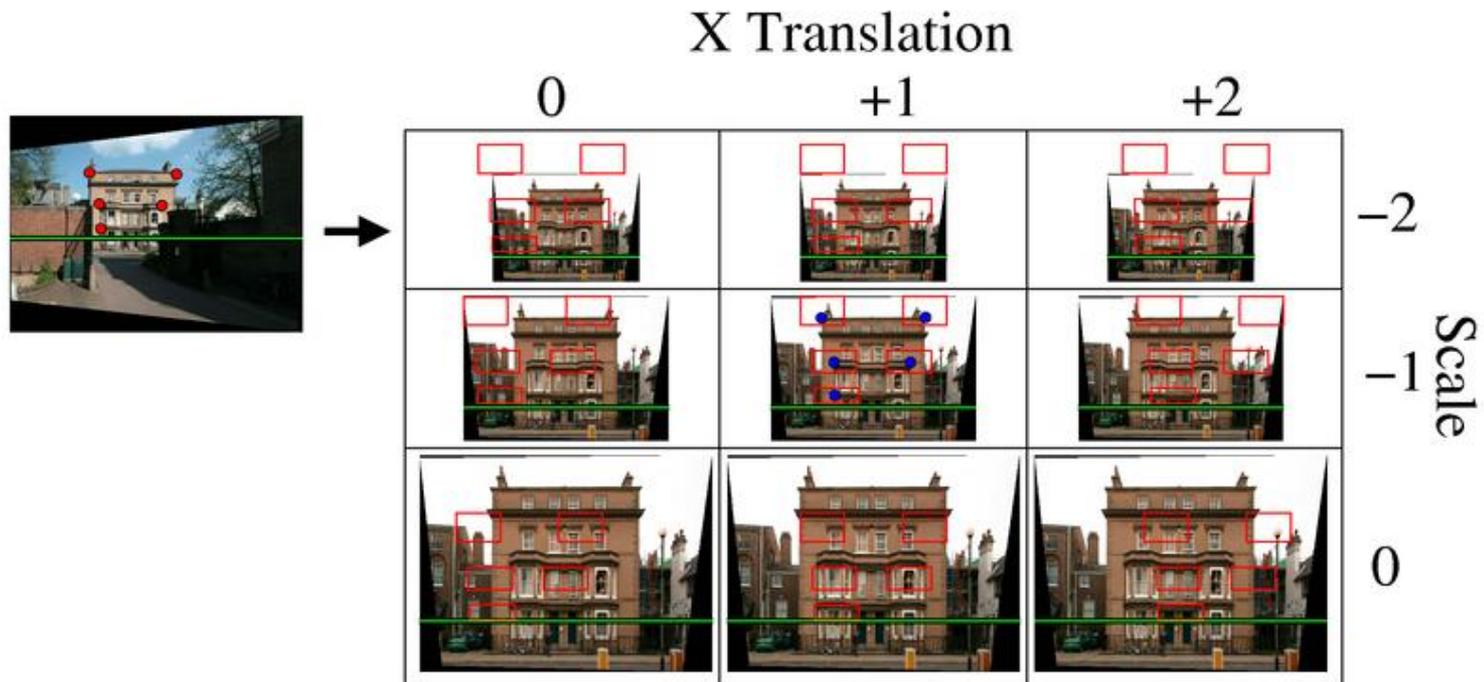
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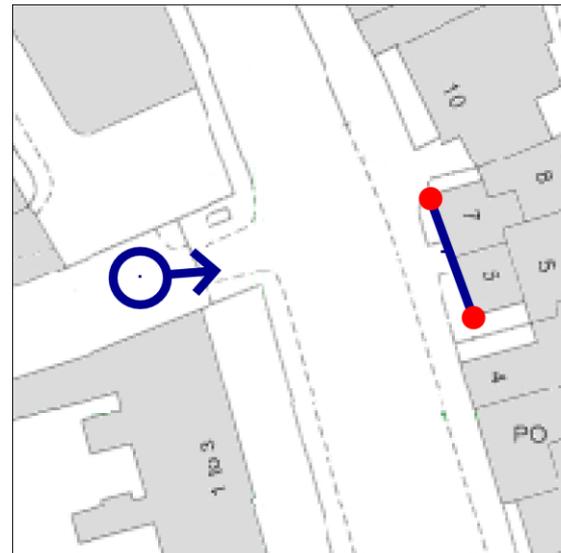
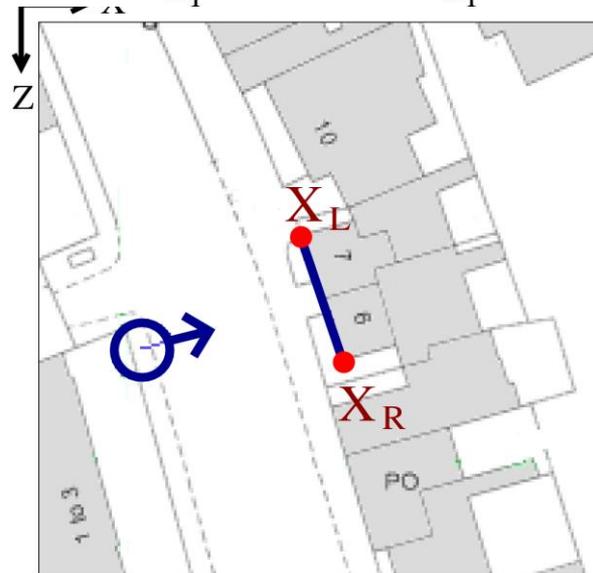
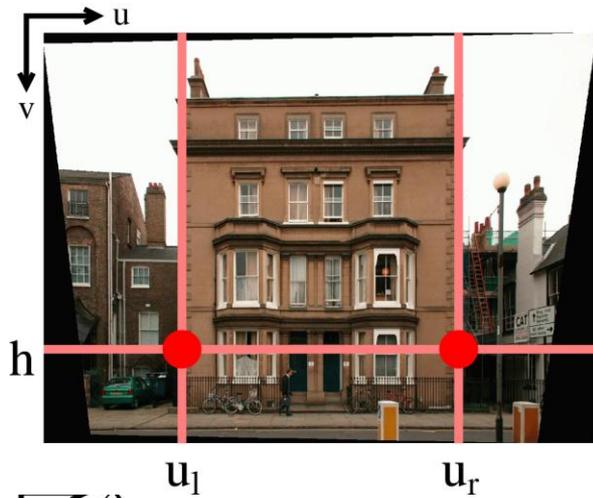
**Only difference is now scale + x translation**

# Matching

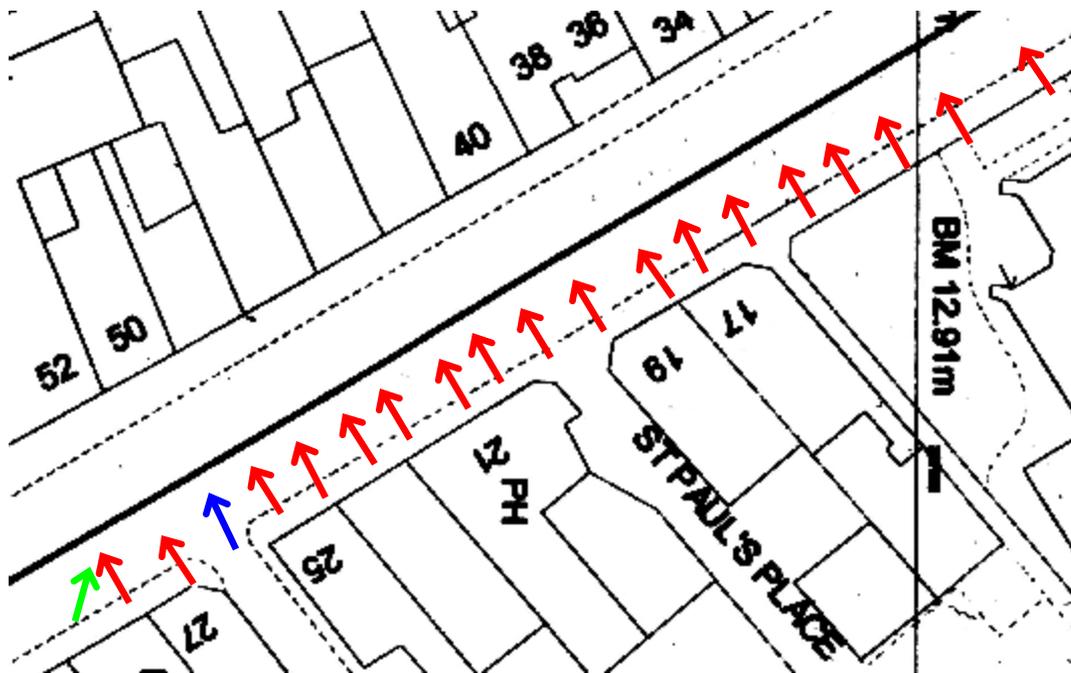
With only 2 params ( $s, t_x$ ), can search rather than RANSAC.



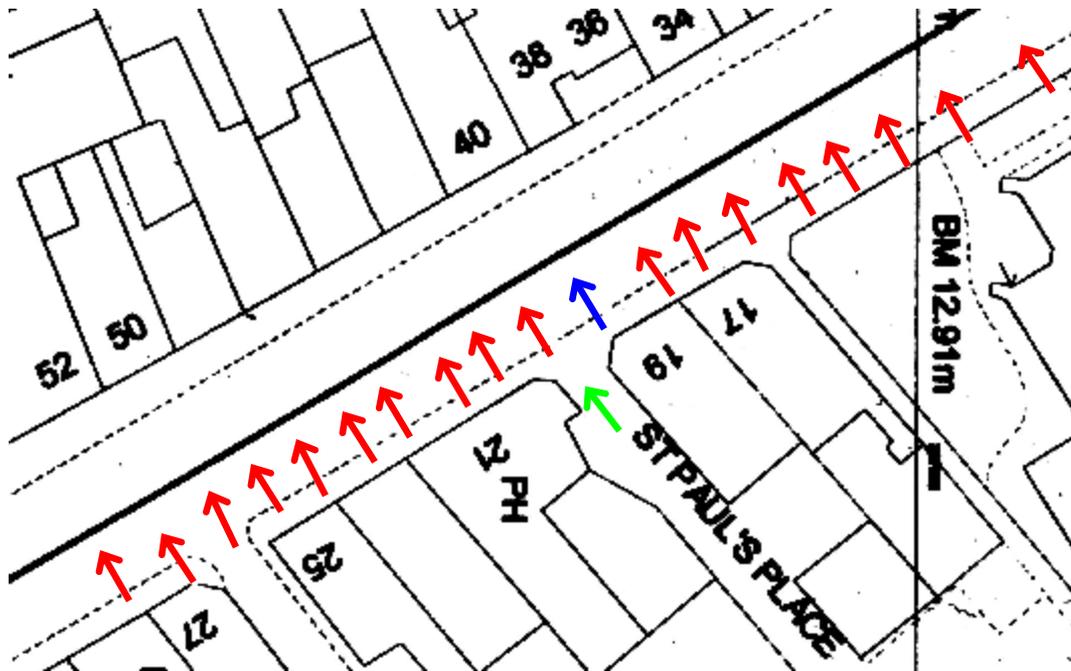
# Localisation of query view



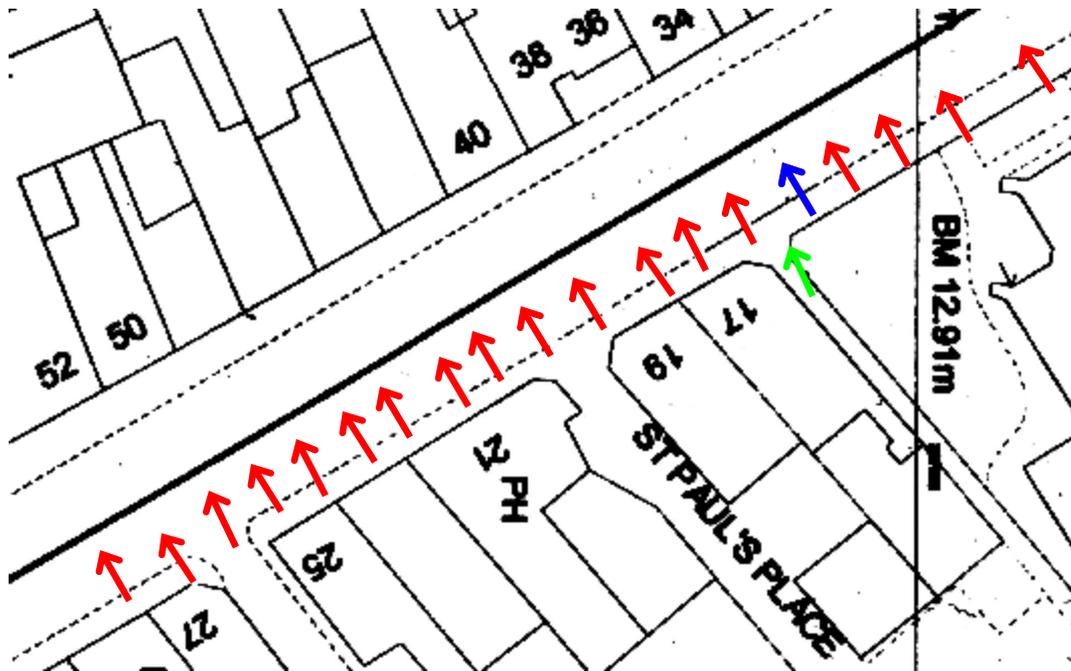
# Image-based localisation



# Image-based localisation



# Image-based localisation



# 5. Finding 2D shapes and applications to HCI

Stenger, Thayananthan, Torr and Cipolla 2003

Williams, Blake and Cipolla 2003 and 2006

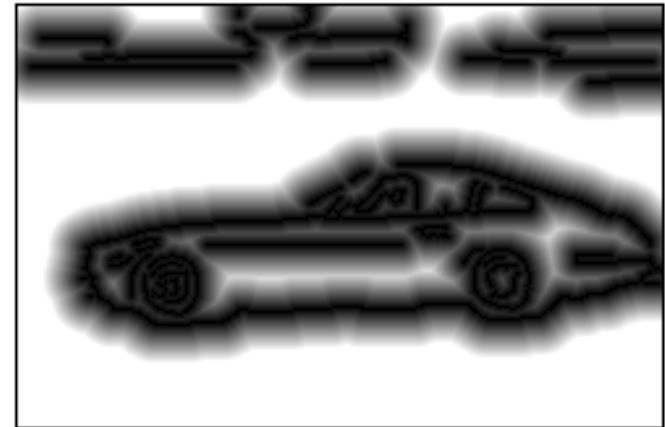
Ramanan, Fitzgibbon and Cipolla 2006-2007

# Matching shape templates

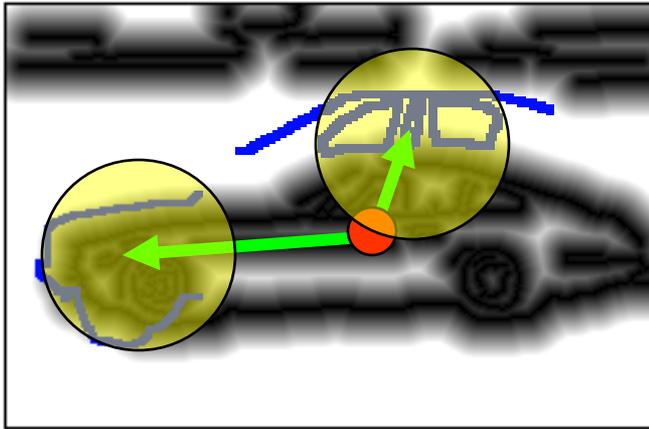
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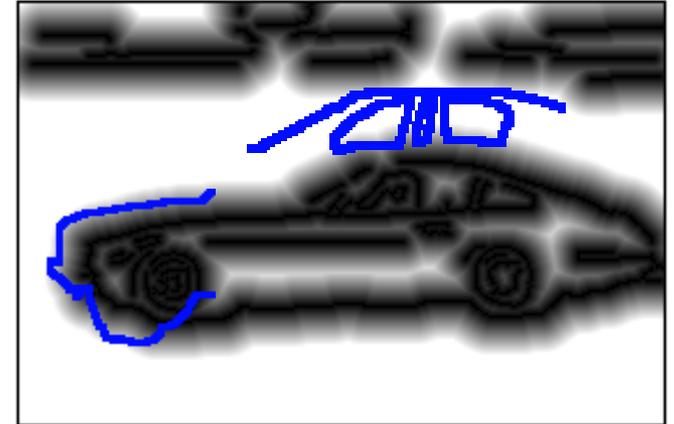
Oriented  
Carry  
Distance  
Edge  
Transform  
Detector



# Matching shape templates



Oriented  
Chamfer  
Matching



# Hand detection system

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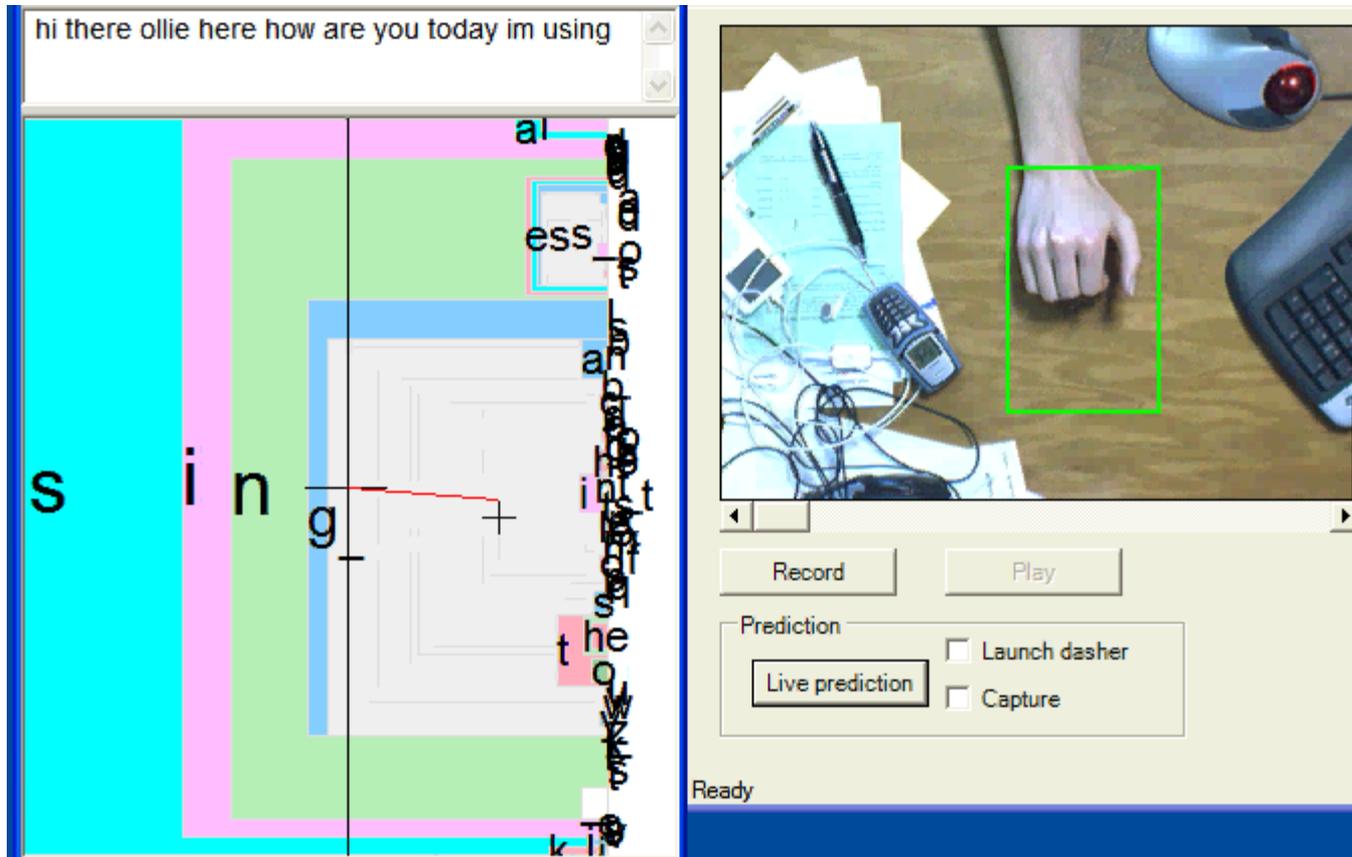


# Tracking - 3D mouse

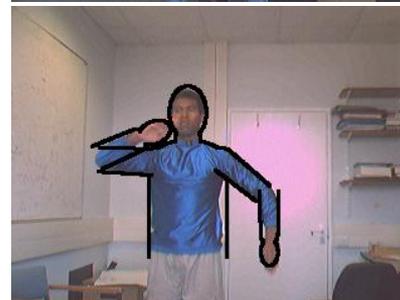
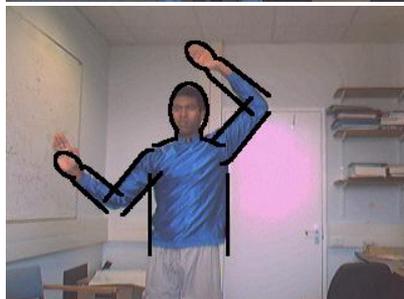
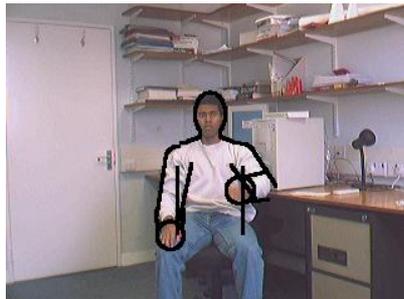
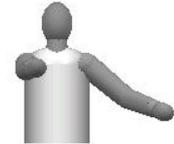
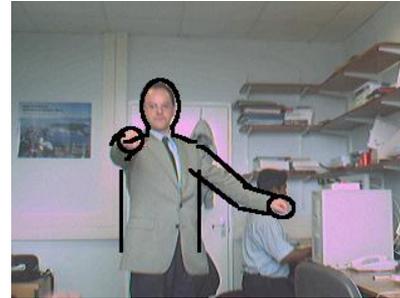
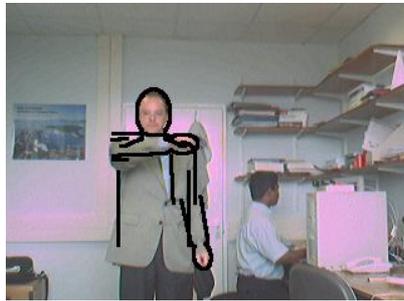
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# Real-time visual controller for Dasher



# People and pose detection



# People and pose detection



# III. Object recognition and machine learning

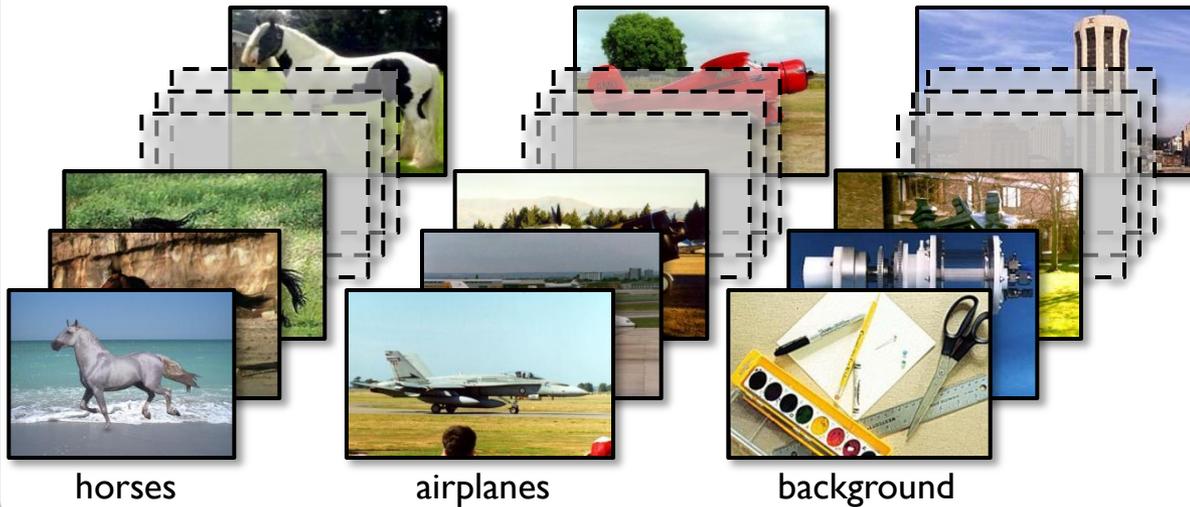
Shotton, Blake and Cipolla 2005-2007

Kim, Kittler and Cipolla 2006

Johnson and Cipolla 2007

# Overview

## image classification



## categorical object detection



## semantic segmentation



# 6. Using interest points and visual words

Johnson and Cipolla 2007

# Image matching

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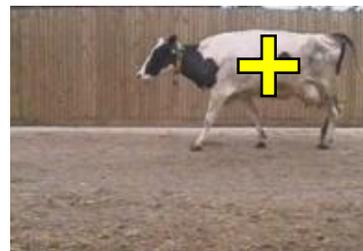
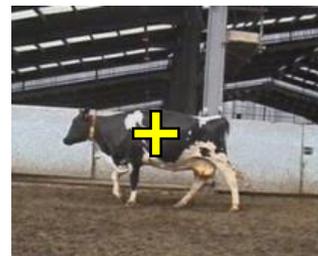
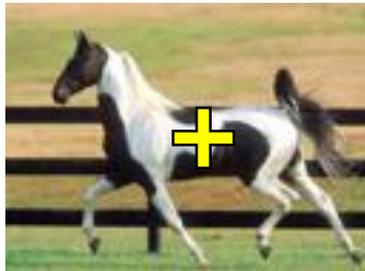
# 7. Using contour and shape

Shotton, Blake and Cipolla 2005-2007

# Supervised learning

- Learn to recognise images of a particular class, localised in space and scale
- i.e. find the horse/cow/car etc!

Desired  
Results

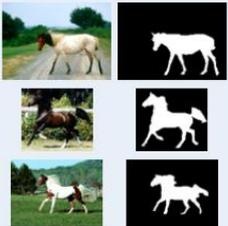


# Learning and Adaptability

## Training Data

### Class

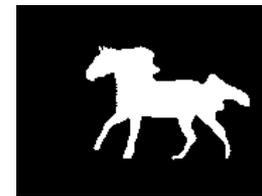
#### Segmented: $D_S$



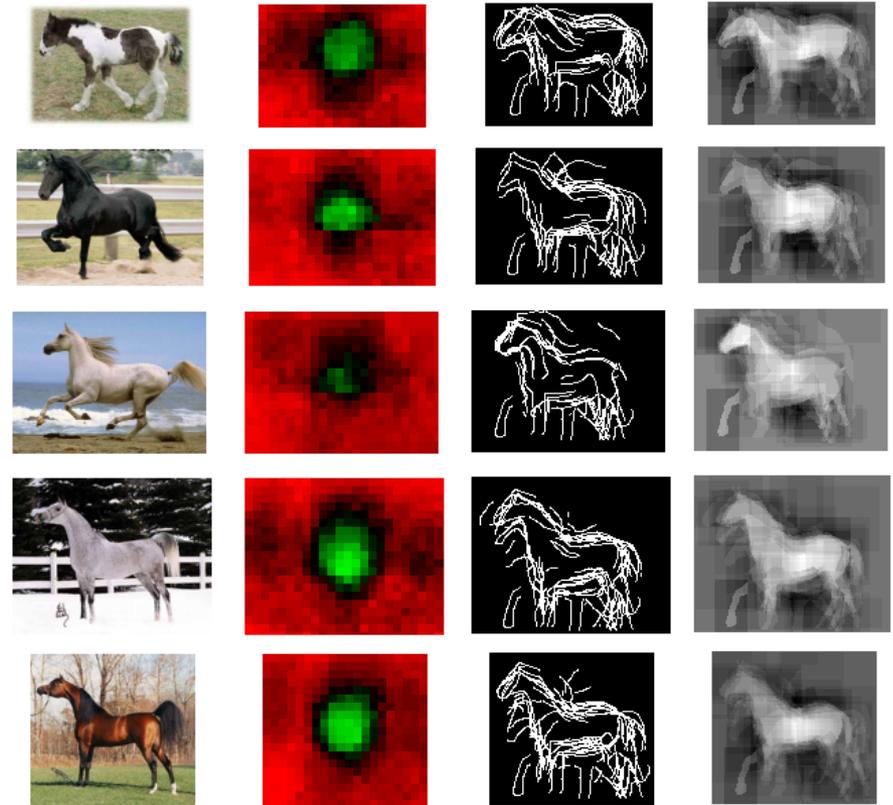
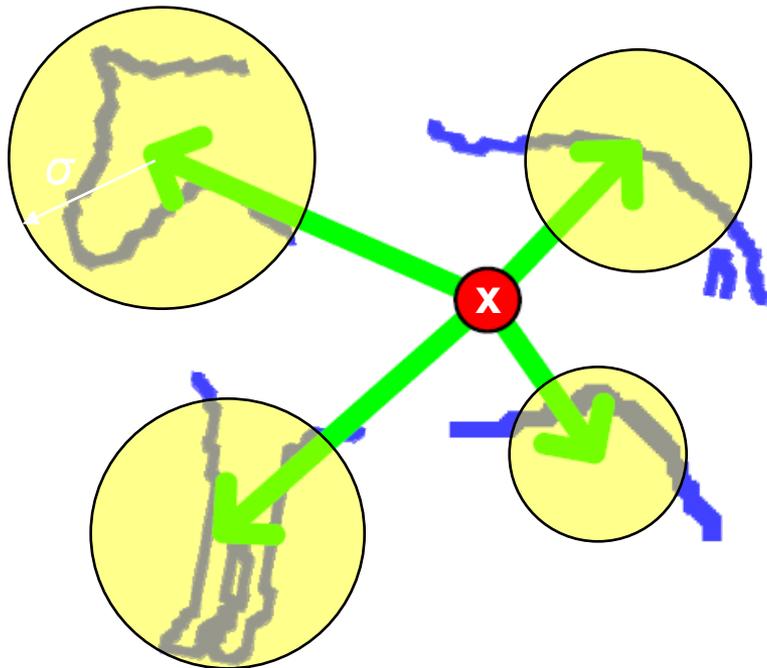
#### Unsegmented: $D_U$



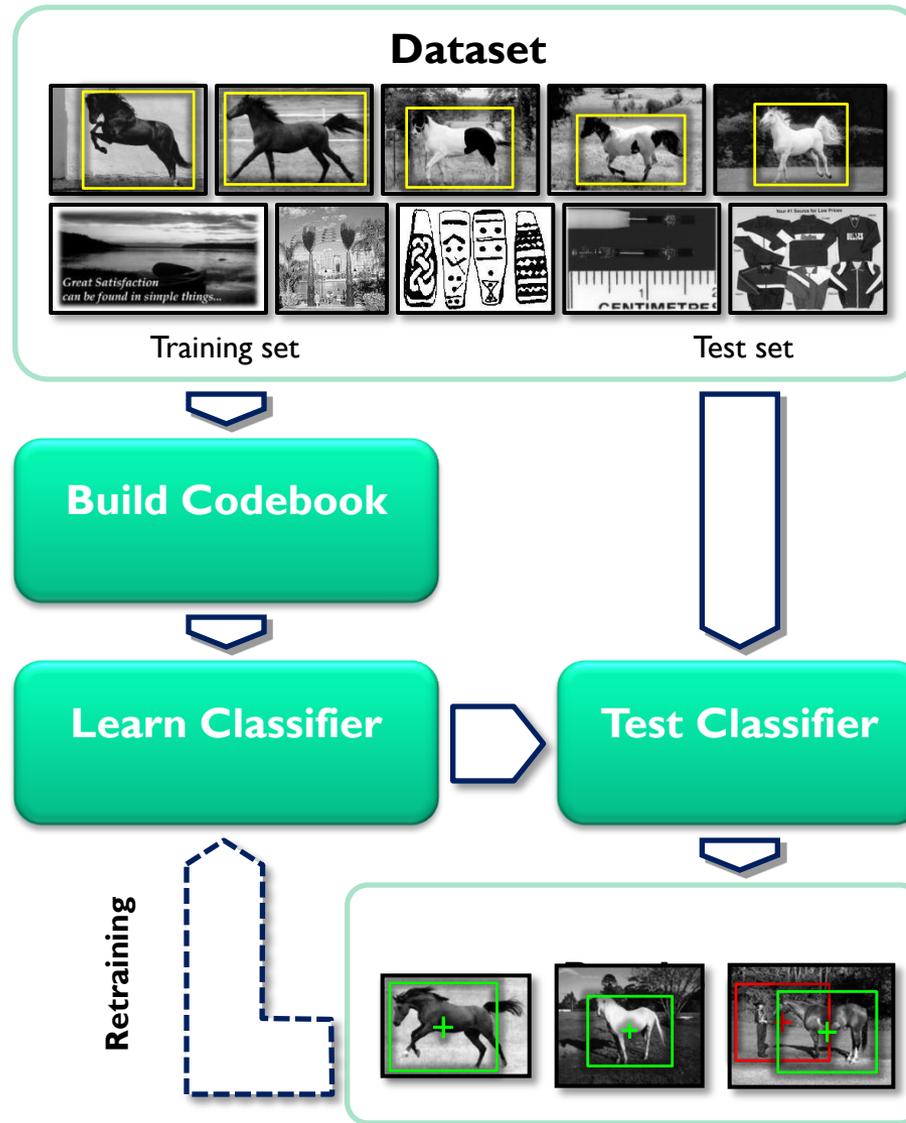
#### Background: $D_B$



# Object Model



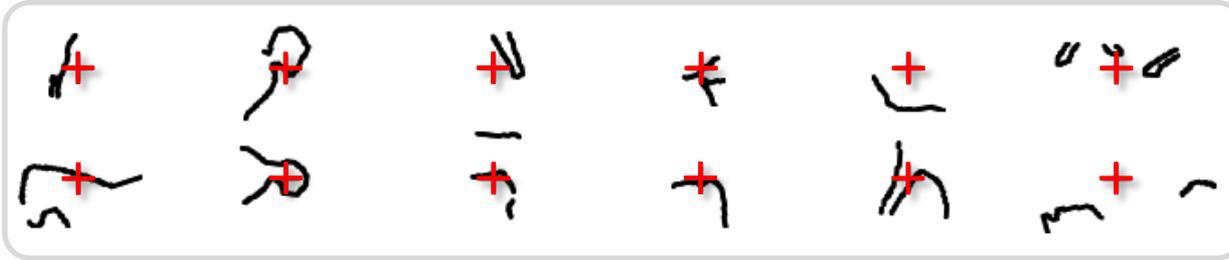
# Approach



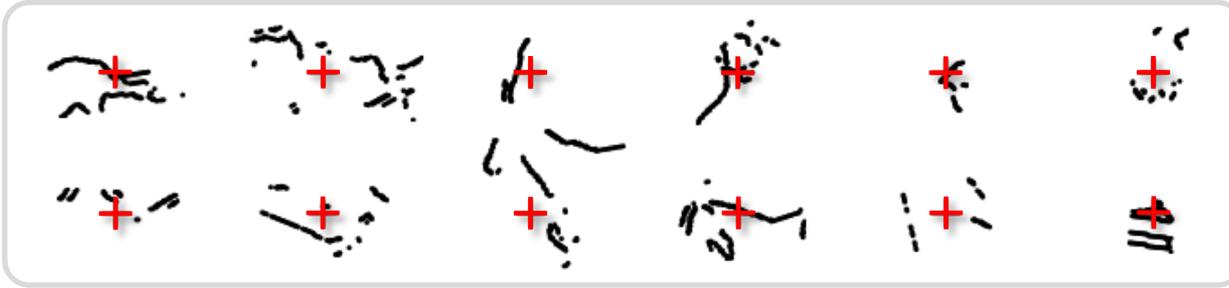
# Dictionary of contour fragments

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Segmented

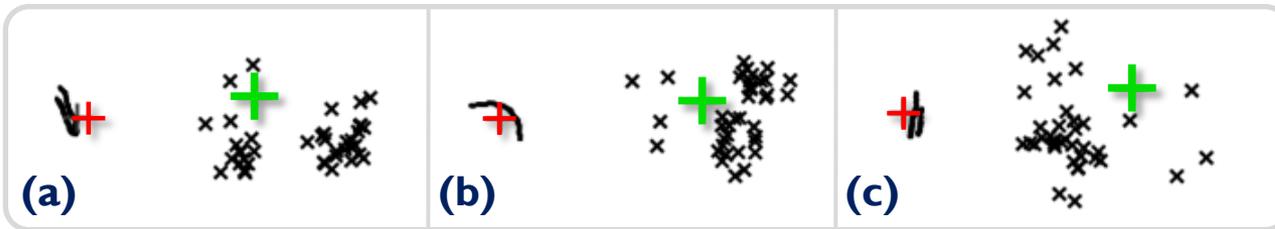


Unsegmented

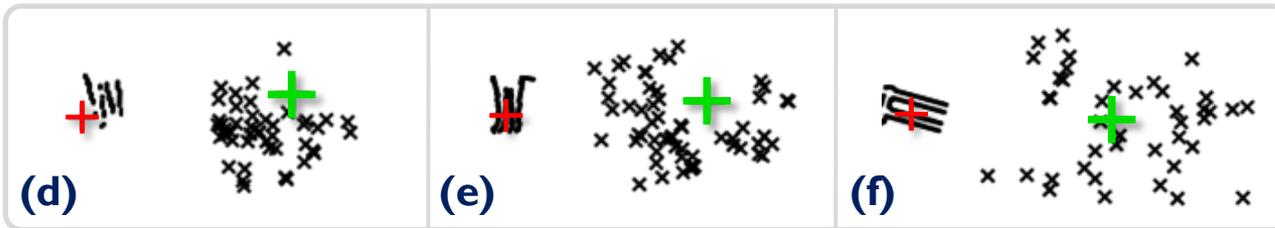


# Shape

Segmented



Unsegmented



Exemplar

Centroid votes

Exemplar

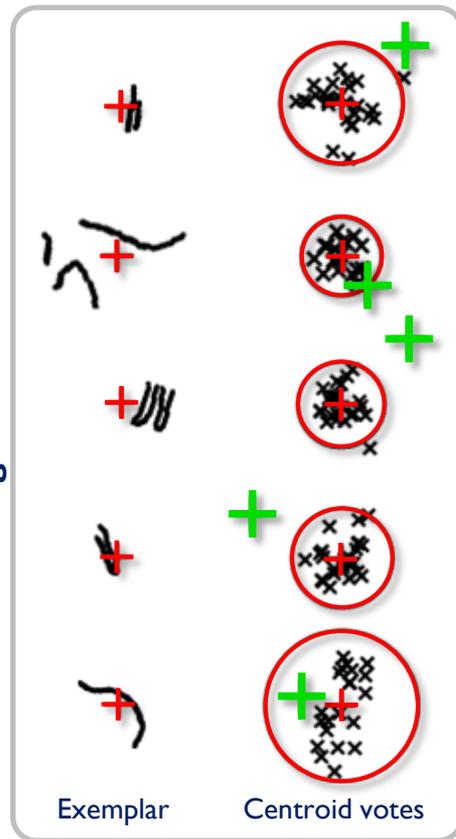
Centroid votes

Exemplar

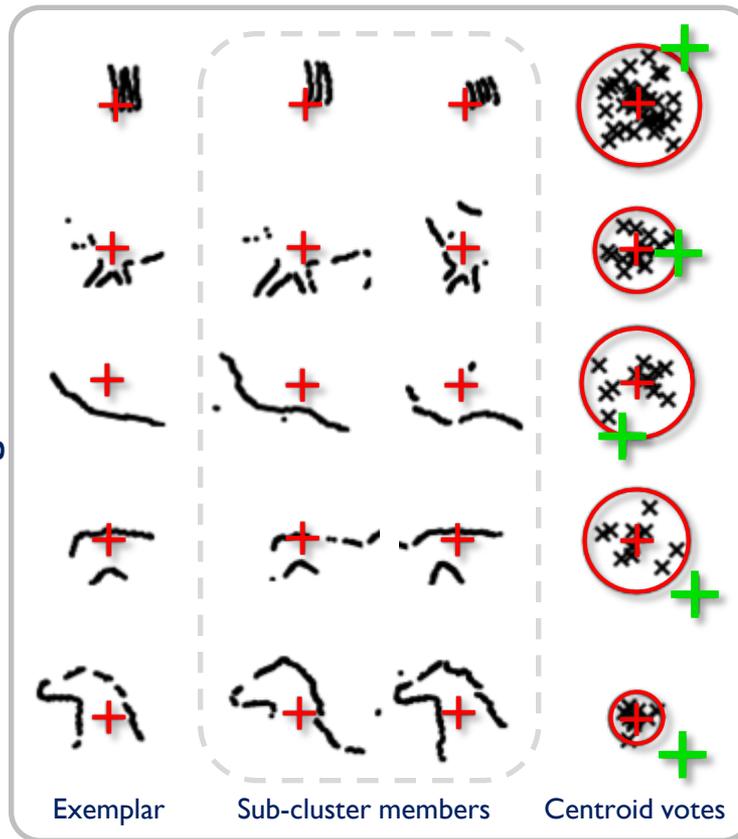
Centroid votes

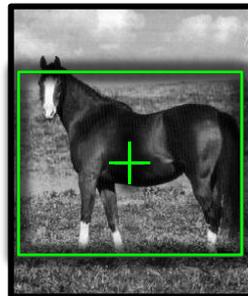
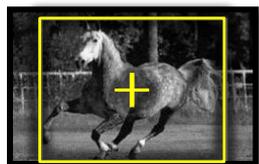
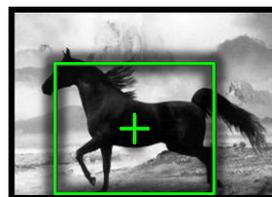
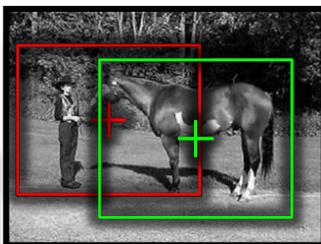
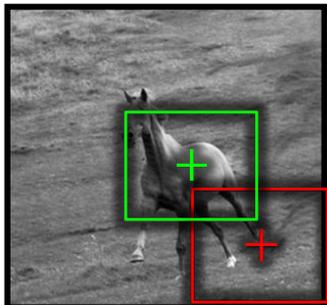
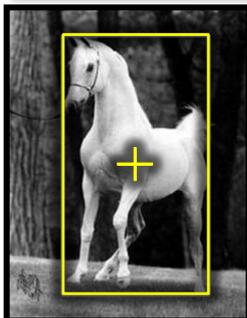
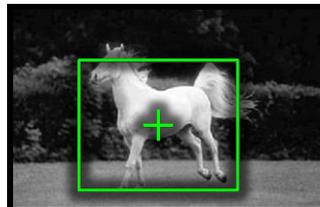
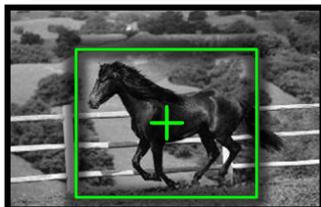
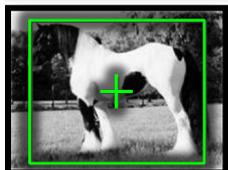
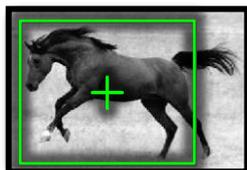
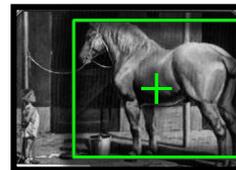
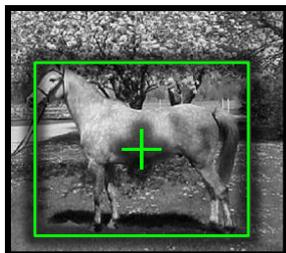
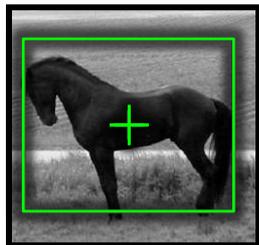
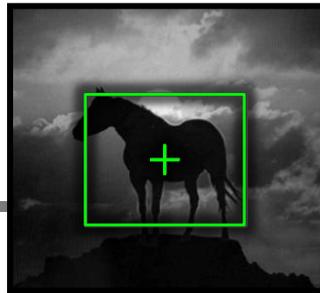
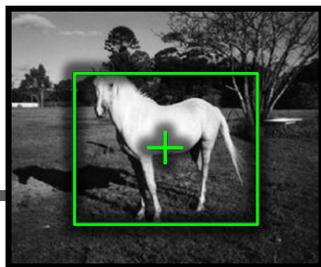
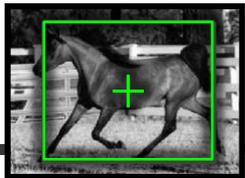
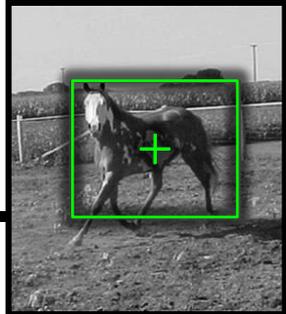
# Shape

Segmented



Unsegmented









# 8. Using texture and contour

Shotton, Blake and Cipolla 2005-2007

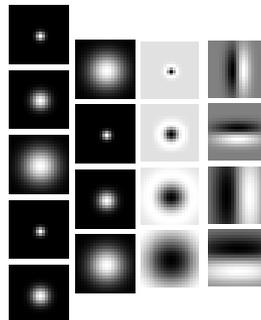


# Extracting textons

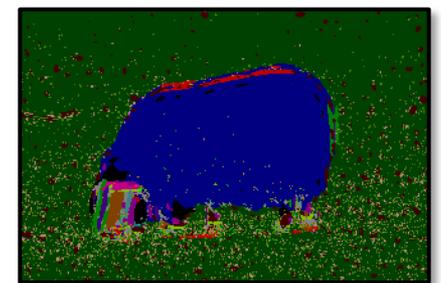
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input image



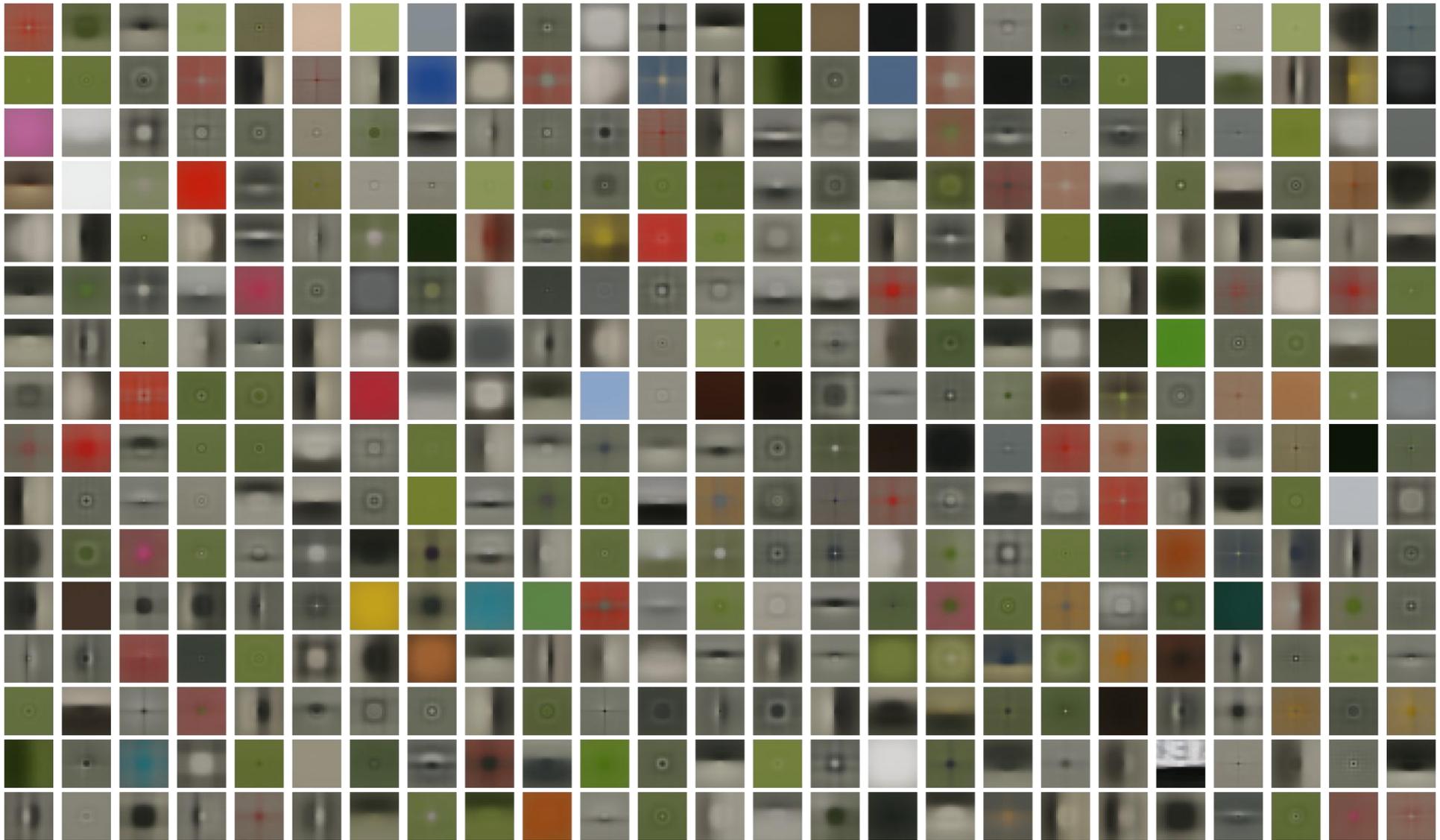
filter bank



texton map  
(colours  $\Leftrightarrow$  texton indices)

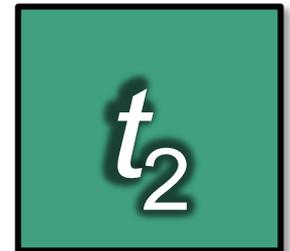
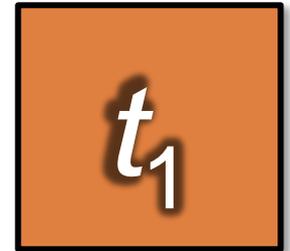
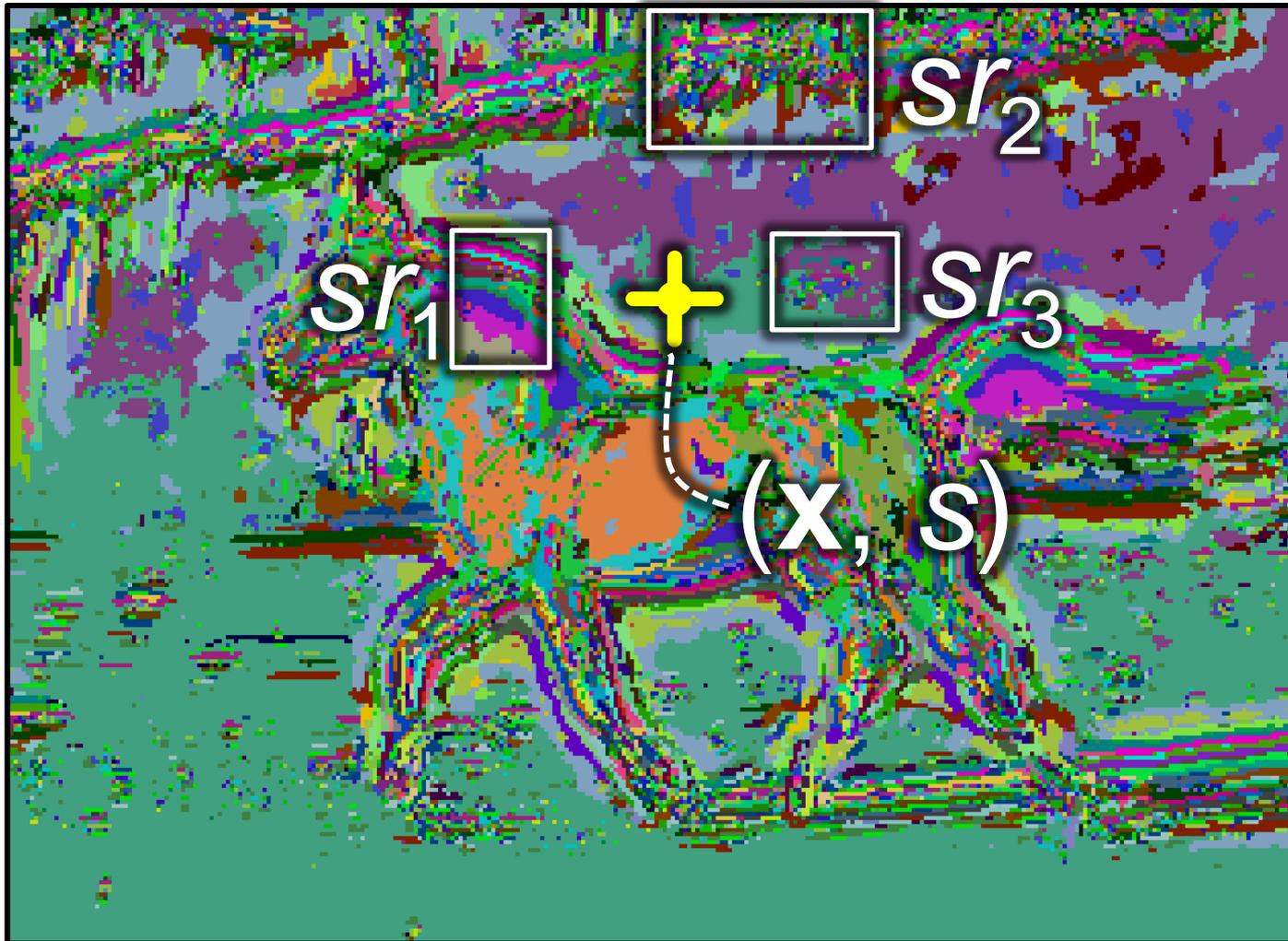
# Extracted texton dictionary

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# Use with contour shape model UNIVERSITY OF CAMBRIDGE

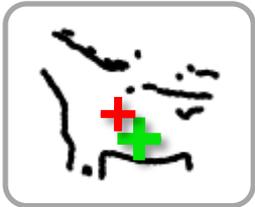
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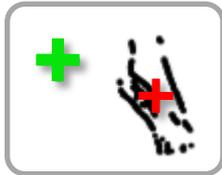
textons

# Learned contour and texture

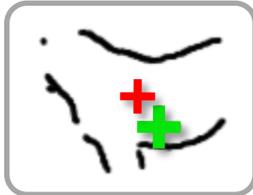
(1)



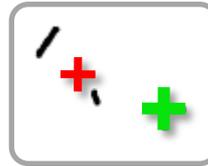
(2)



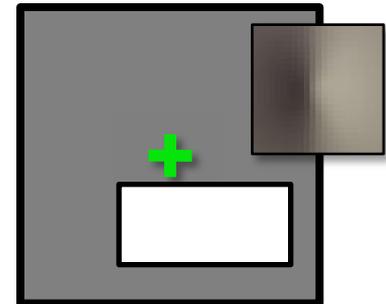
(3)



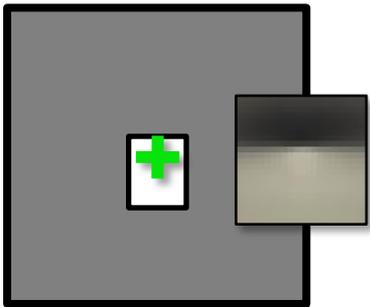
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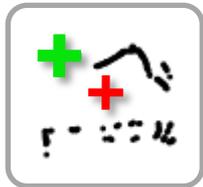
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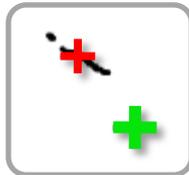
(6)



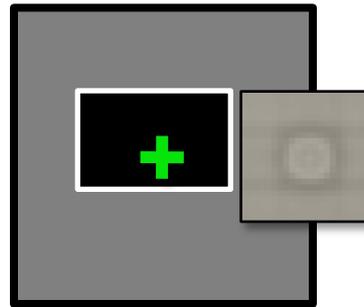
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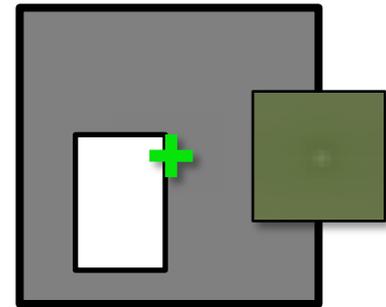
(8)



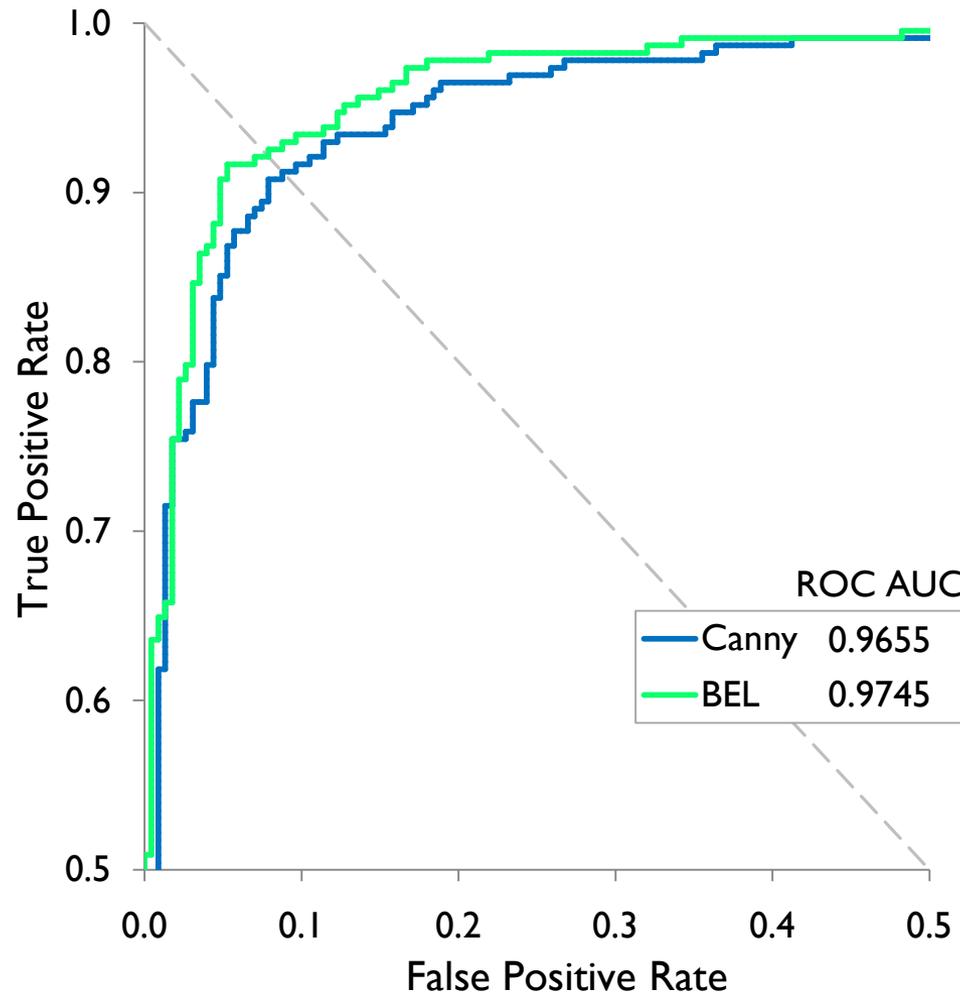
(9)



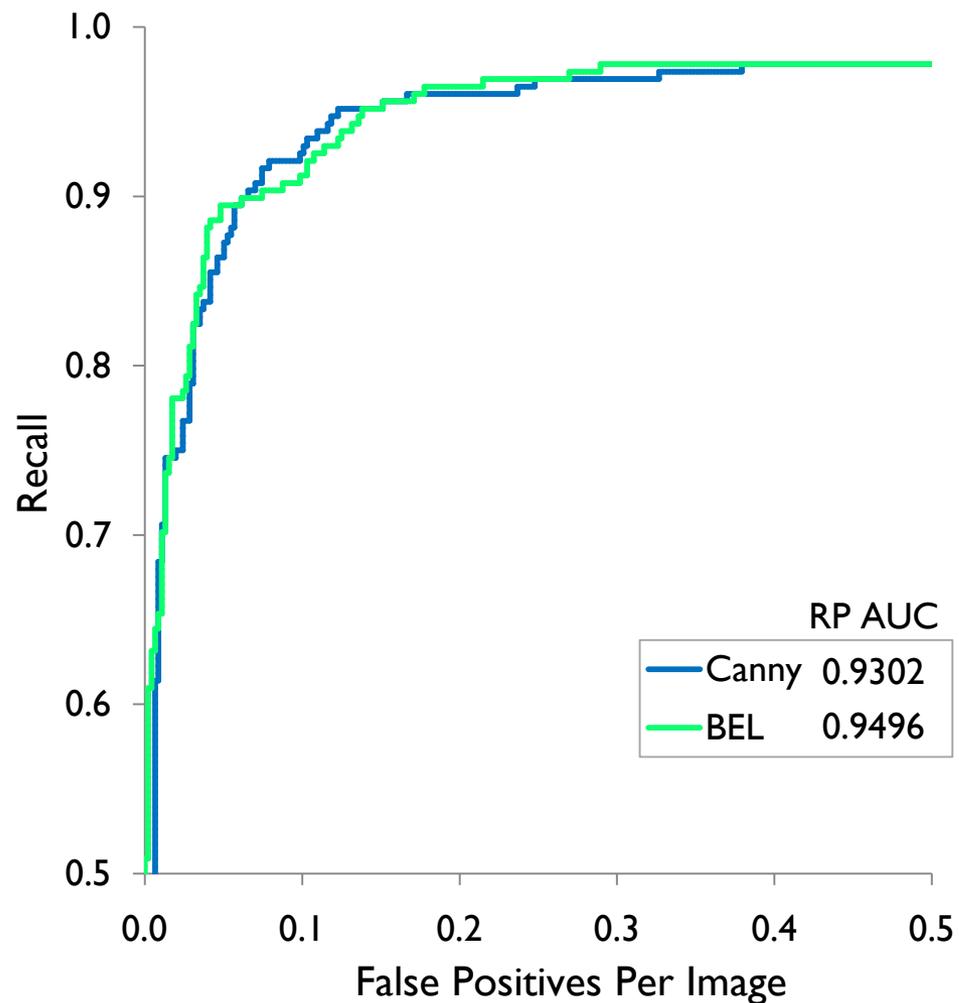
(24)



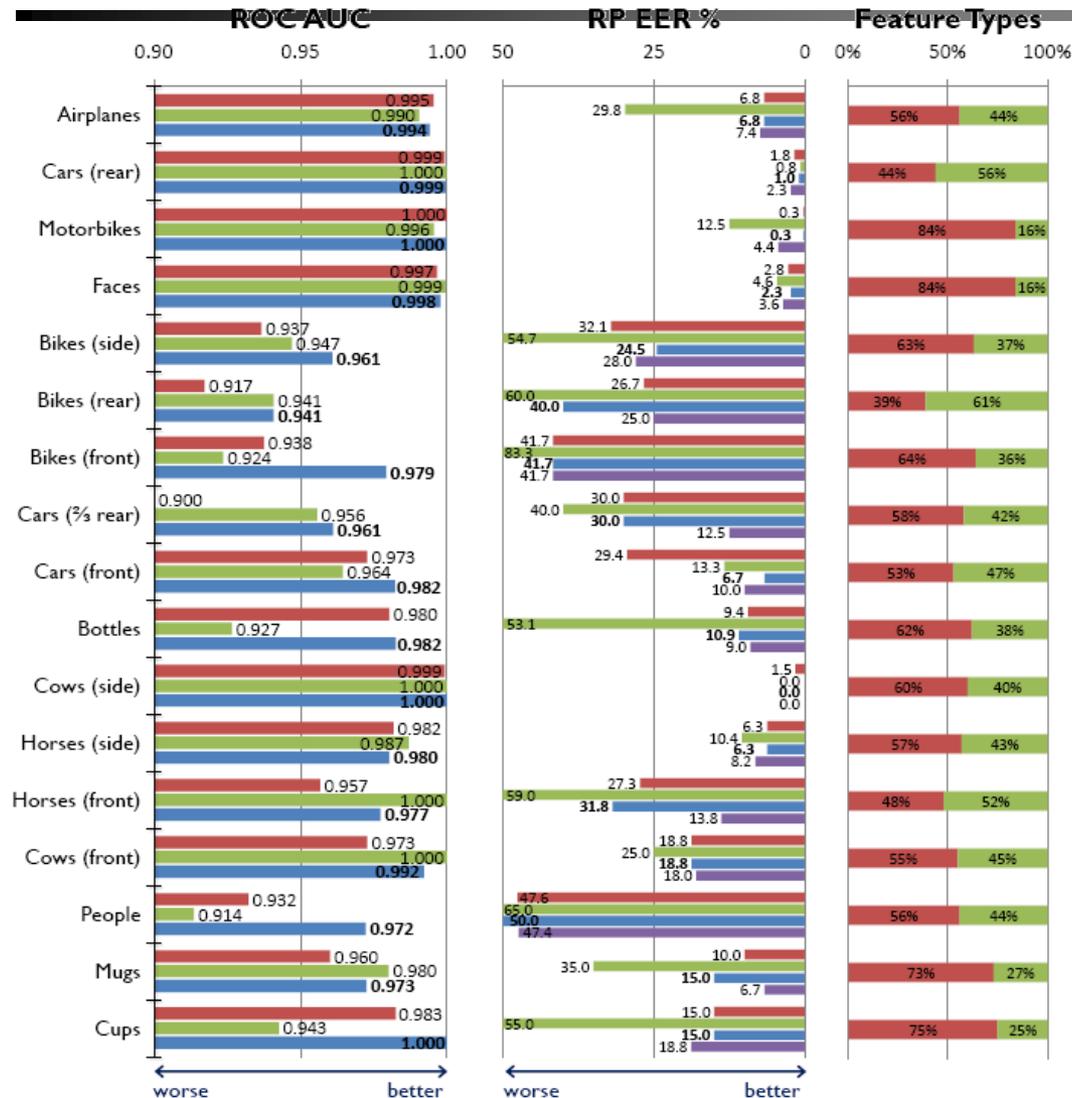
# Performance



# Performance



# Performance



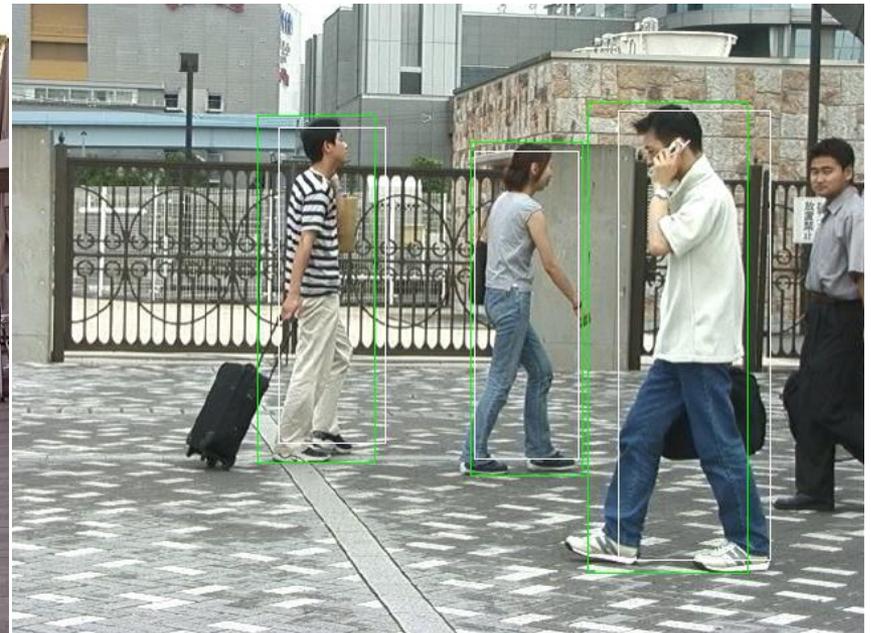
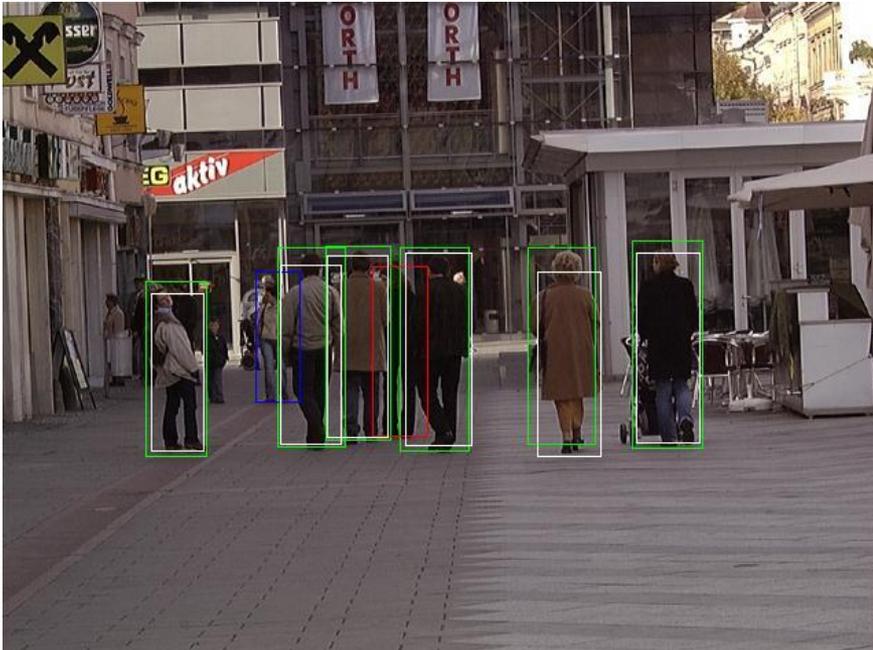
# 9. Detection and tracking of people

Brostow and Cipolla 2006

Bucciarelli and Cipolla 2007

# Pedestrian detection

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# Tracking people in crowds

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Detecting People in  
Crowds by Bayesian Clustering

Brostow & Cipolla, 2005

# 10. Recognition in video using CCA

Arandjelovic and Cipolla 2006

Kim, Wong and Cipolla 2006-2007

# Face recognition

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Overcome appearance variations due to:

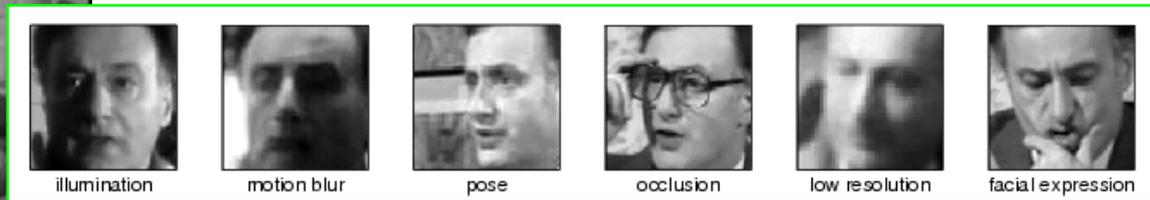
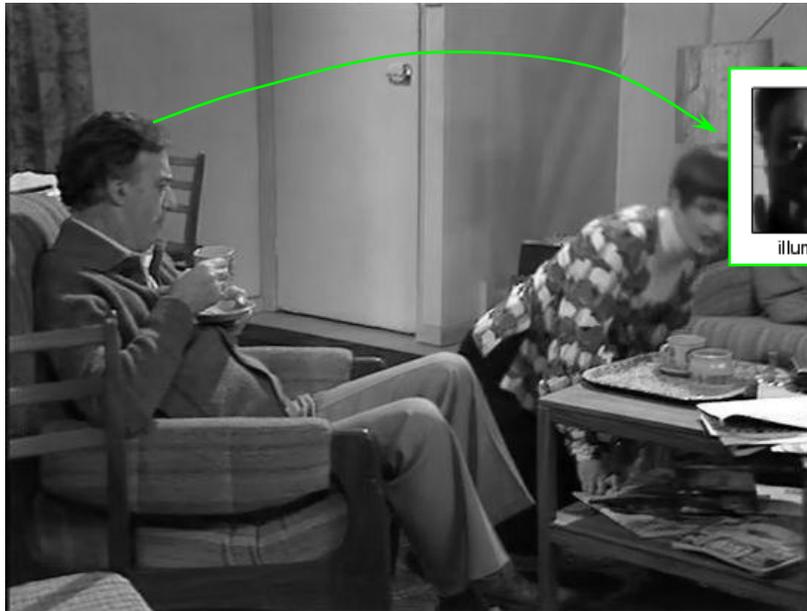
Lighting condition



Scale, pose, motion pattern



# Automatic cast listing



Problem difficulties

# Automatic cast listing

“Simple clustering”  
results

Sir Hacker:



Miss Hacker:



Humphrey:



Secretary:



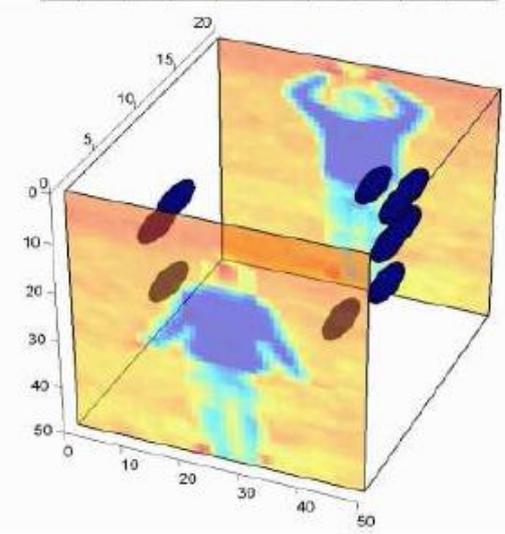
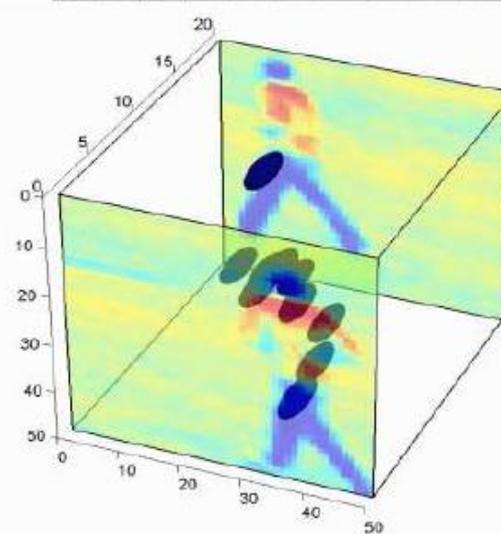
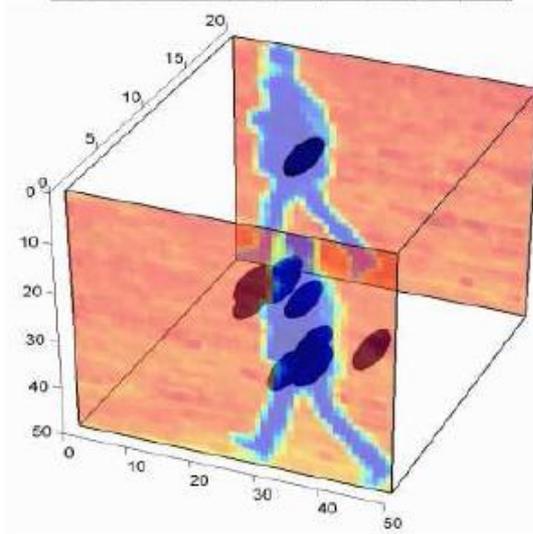
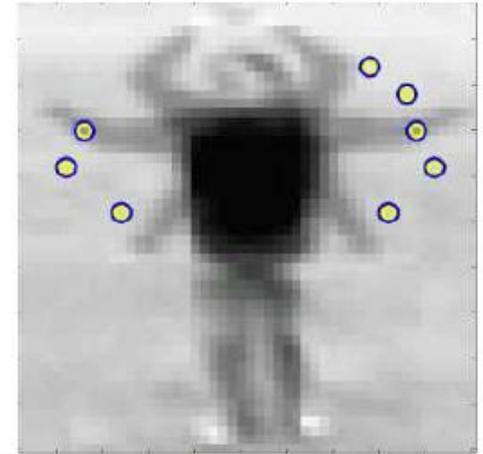
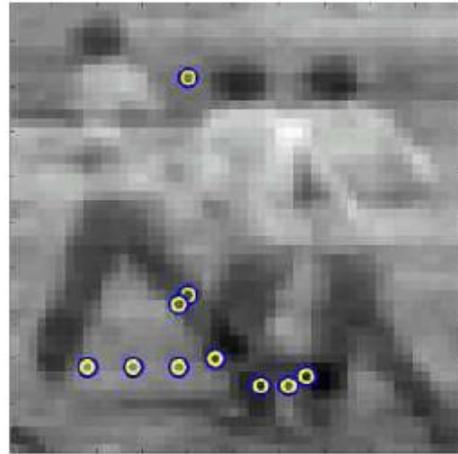
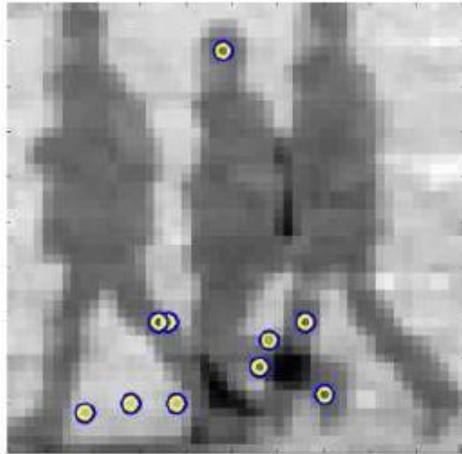
Bernard:



Frank:



# Action recognition



# Action recognition

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Boxing



Hand clapping



Hand waving



Jogging



Running



Walking

# Summary

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- Image registration and matching
- 3D shape from uncalibrated images.
- Object detection and tracking