# Introduction to Computer Vision

#### Michael J. Black

#### **Project Ideas**

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#### Dates

#### 11/13 Proposals due

- 1 page write-up
- summary and goals (problem/approach)
- what are the key references
- where will you get data and how will you evaluate your method?

#### 12/14 Projects due

# Eigen faces



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#### Detect faces





Mahalanobis distance

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Masked by skin detections

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*Times the image* 

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# If you work on face detection

- You must use color in some way (e.g. color eigenspace)
- You must search across a range of scales.
- You must try it on some image data that was not collected in class.
- Use the Moghaddam and Pentland paper.
- Extra: try gender recognition.





a) original photo

b) silhouette

c) reconstruction

- Kutulakos, K. and Seitz, S. 2000. Theory of Shape by Space Carving. *International Journal of Computer Vision*,
- 38(3):199-218.
- Images from ETH-80 database.

## Sources of data

#### http://www.vision.ethz.ch/projects/categorization/eth80-db.html



#### Image segmentation



#### http://www-2.cs.cmu.edu/~jshi/Grouping/

## Face gender classification?



http://vis-www.cs.umass.edu/lfw/

13233 labeled faces of 5749 people found with Viola-Jones detector (Adaboost).

Use names to get gender: (http://www.gpeters.com/names/ baby-names.php?)

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## More face images

# There are many other face databases on the web. Eg:

http://titan.cog.brown.edu:8080/TarrLab/face-place

### Kalman Filter Tracker

• Real-time tracker using a PC camera.



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### Mean-shift tracking



http://www.caip.rutgers.edu/~comanici/Papers/KernelTracking.pdf

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# Facial expressions

- Affine head tracking
- analysis of motions of facial features



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# **EigenTracking** (Black and Jepson)

Combines affine motion estimation with PCA representation to allow tracking of deforming objects.

Data:

http://www.cs.brown.edu/ ~black/images.html



http://www.cs.brown.edu/~black/eigenTrack.html

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EigenTracking: Robust Affine Matching Using a View-Based Representation

Submitted to ECCV'96

**Confidential: For Review Only** 

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http://www.robots.ox.ac.uk/~cbibby/research\_pwp.shtml

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#### Tadpole tracker



• Contact black or moldovan for data.

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(a) Example image with predicted position plotted





(c) Pixel intensity distribution



(d) Combined distribution



(e) Example image with 200 particles



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#### Rodent tracker



Contact: Prof. Russ Church (Russell\_Church@brown.edu)

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#### Active Shape/Appearance Models



*http://www.isbe.man.ac.uk/~bim/* Lot's of support code and data on web.

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#### Stereo



#### http://vision.middlebury.edu/stereo/

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#### Panoramic Mosaics



http://www.cs.washington.edu/education/courses/455/03wi/projects/project2/project2.htm

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# Dense Optical Flow or Stereo

Data and ground truth flow:

http://vision.middlebury.edu/flow/

Same for stereo

http://vision.middlebury.edu/stereo/



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# Pedestrian Detectors

#### http://lear.inrialpes.fr/people/triggs/pubs/Dalal-cvpr05.pdf



Figure 6. Our HOG detectors cue mainly on silhouette contours (especially the head, shoulders and feet). The most active blocks are centred on the image background just *outside* the contour. (a) The average gradient image over the training examples. (b) Each "pixel" shows the maximum positive SVM weight in the block centred on the pixel. (c) Likewise for the negative SVM weights. (d) A test image. (e) It's computed R-HOG descriptor. (f,g) The R-HOG descriptor weighted by respectively the positive and the negative SVM weights.

#### Non-linear diffusion

#### Image denoising.

Original Image



Edge Enhancing Diffusion



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http://www-mount.ee.umn.edu/~guille/inpainting.htm

-this turns out to be too hard and the best thing to do is a non-linear diffusion method.

## Image Inpainting



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#### Colorization



#### http://www.cs.huji.ac.il/~yweiss/Colorization/

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### Super-resolution



See Michal Irani and Shmuel Peleg.Super Resolution From Image Sequences.IEEE, 1990.

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Forensic Computer Vision April 2006



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## Super-resolution



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# More Project "Ideas"

Temporal model of mouth motions (HMM) for recognition. More advanced machine learning method for mouth or person detection

- AdaBoost
- support vector machines
- Bayesian image denoising

Stereo

Space carving from silhouettes

Grab-cut

Moghaddam mixture model for face/mouth recognition.