

# Lecture 17

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## Final Review

**All good things must...**

# Course Grading

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## Programming Projects (80%)

- Image scissors (20%) - **DONE!**
- Panoramas (20%) - **DONE!**
- Content-based image retrieval (20%) - **DONE!**
- Face recognition & detection (20%) – **Due tomorrow (11:59pm)**

## Final (20%)

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Everything you wanted to know about the final  
but were afraid to ask...

# The Final Exam

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- When & Where: Thursday March 19 10:30am-12:20pm here (EE 037)
- Closed book, closed notes except for one 8 ½” x 11” sheet of notes
- Format: Similar to [sample exam](#) on class website (under Calendar)
  - Five questions
  - Short answer questions as well as longer problems (see sample exam for examples)

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How do I ace the final?

# Option A: Memorize this

NAME: \_\_\_\_\_

Computer Tables Prof. David Wu  
The Ohio State Univ.  
March 2016

1. Example

Example (Final Exam)  
Based on problem 10 (10 points) in this, this and 10 (10)

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NAME: \_\_\_\_\_

Problem 1: Blank matrix

For each of the following, explain how it may be used to help solve a computer table problem (and give the problem).

- Example
- Principal components
- Linear regression
- Bayes' rule
- Neural networks
- Support vectors
- Decision trees
- Bayesian networks
- Hidden Markov models
- Markov chains
- Monte Carlo
- Genetic algorithms
- Evolutionary algorithms
- Game theory
- Reinforcement learning
- Deep learning
- Transfer learning
- Active learning
- Online learning
- Meta-learning

NAME: \_\_\_\_\_

Problem 2: Blank matrix

For each of the following, explain how it may be used to help solve a computer table problem (and give the problem).

- Example
- Principal components
- Linear regression
- Bayes' rule
- Neural networks
- Support vectors
- Decision trees
- Bayesian networks
- Hidden Markov models
- Markov chains
- Monte Carlo
- Genetic algorithms
- Evolutionary algorithms
- Game theory
- Reinforcement learning
- Deep learning
- Transfer learning
- Active learning
- Online learning
- Meta-learning

NAME: \_\_\_\_\_

Problem 3: Sparse Synthesis

Given the set of vectors  $\{v_1, \dots, v_n\}$  in  $\mathbb{R}^m$ , determine the smallest number of vectors that span  $\mathbb{R}^m$ .

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NAME: \_\_\_\_\_

Problem 4: Sparse Synthesis

Given the set of vectors  $\{v_1, \dots, v_n\}$  in  $\mathbb{R}^m$ , determine the smallest number of vectors that span  $\mathbb{R}^m$ .

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NAME: \_\_\_\_\_

Problem 5: Sparse Synthesis

Given the set of vectors  $\{v_1, \dots, v_n\}$  in  $\mathbb{R}^m$ , determine the smallest number of vectors that span  $\mathbb{R}^m$ .

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NAME: \_\_\_\_\_

Problem 6: Sparse Synthesis

Given the set of vectors  $\{v_1, \dots, v_n\}$  in  $\mathbb{R}^m$ , determine the smallest number of vectors that span  $\mathbb{R}^m$ .

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NAME: \_\_\_\_\_

Problem 7: Sparse Synthesis

Given the set of vectors  $\{v_1, \dots, v_n\}$  in  $\mathbb{R}^m$ , determine the smallest number of vectors that span  $\mathbb{R}^m$ .

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# Option B

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- Read the lecture slides
  - Only topics covered in the lecture slides will be the subject of questions in the final
- Consult the accompanying readings (on Calendar page) and projects for details where necessary
- Try the sample final

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# Course review and highlights



In the beginning...

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# Image Processing and Filtering

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- Image as a function
- Image formation: illumination, reflectance
- Domain transformations
  - Homogenous coordinates
  - Affine transformations
- Range transformations:

## Filtering

- Cross-correlation, template matching
- Convolution
- Filter kernel
- Mean filtering, Gaussian filtering
- Median filtering

# Highlight

## Noise reduction through mean filtering

Input image



Salt and pepper noise

Filtered Images



3 x 3



5 x 5



7 x 7

Kernel size

# Image Sampling and Pyramids

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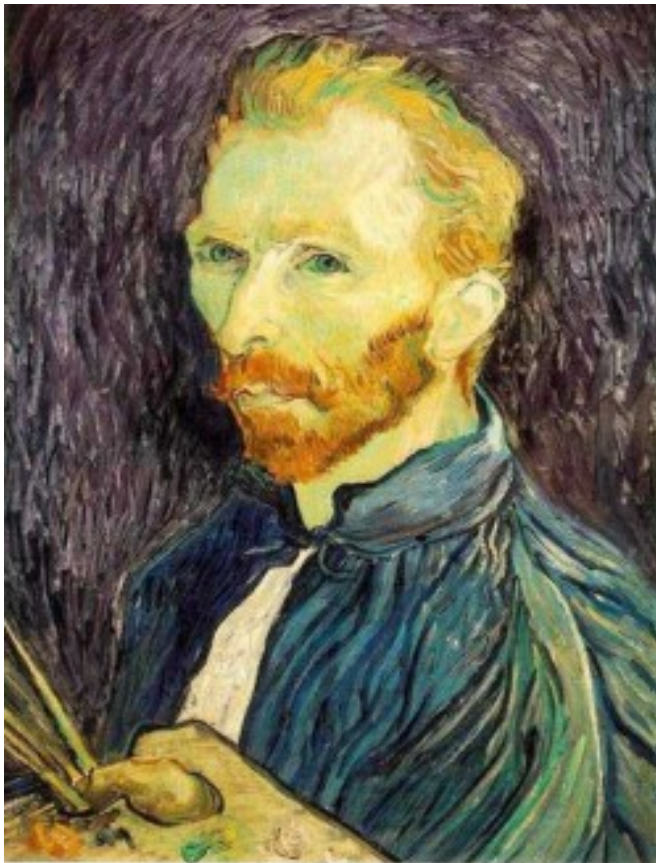
- Sub-sampling artifacts
- Aliasing and Nyquist rate
- Gaussian pre-filtering
- Gaussian pyramid
- Image resampling and interpolation
  - Linear and bilinear

# Highlight

## Gaussian Pyramid

Blur the image (low pass filter) the image, *then* subsample

- Blur using Gaussian filter



Gaussian 1/2



G 1/4



G 1/8

# Edge Detection

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- Image gradient
  - Magnitude and direction
  - Discrete gradient, Sobel operator
- Derivative theorem of convolution
- Edge detection filters
  - Derivative of Gaussian
  - Laplacian, Laplacian of Gaussian (LoG)
- Canny edge detector (5 steps)

**Highlight**

Edge detection using LoGs

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original image (Lena)



LoG followed by zero crossing detection

# Intelligent Scissors

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- Problem being solved?
  - Semi-automatic segmentation
- Computing lowest cost paths from seed to all pixels
- Image as a graph
  - Cost of links computed from cross-correlation filter
  - Dijkstra's shortest path algorithm
- Creating composite images



# Highlight

# Your creativity at work!



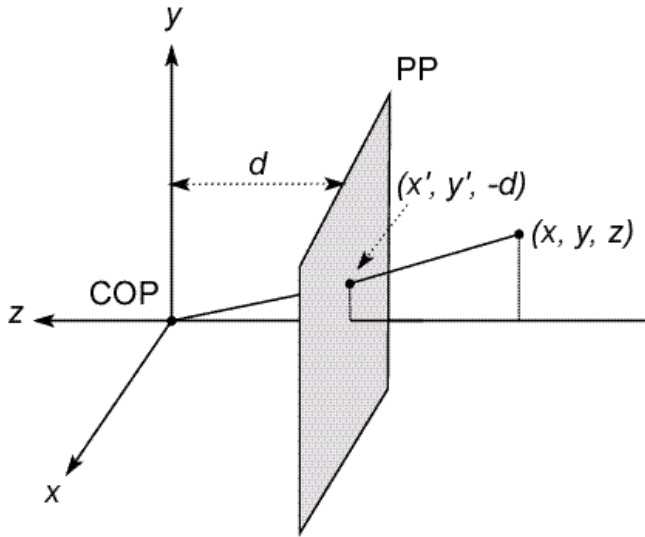
# Cameras and Projection

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- Pinhole camera
  - Aperture and effects of its size
- Lenses
  - Thin lens equation
  - Magnification by a lens
- Modeling projection
  - COP, PP
  - Projection equations (derive using similar triangles)
  - Homogenous coordinates
  - Perspective projection using homogenous coordinates
  - Vanishing points

# Highlight

# Perspective Projection



# Other Highlights



Your work!



# Cameras and Projection (cont.)

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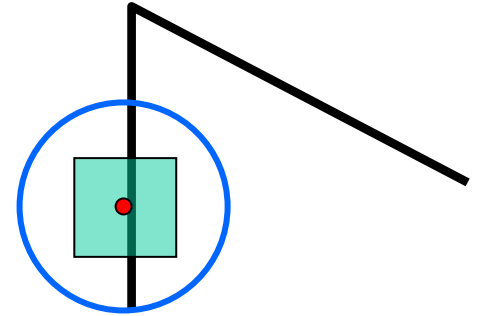
- Simplified projection models
  - Weak perspective
  - Orthographic
- Radial distortion modeling
- Extrinsic vs. intrinsic camera parameters
- World coordinates to pixel coordinates transform
- Camera calibration

# Features and Image Matching

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- Feature Detection
  - Finding interest points

$$E(u, v) \approx [u \ v] \underbrace{\left( \sum_{(x,y) \in W} \begin{bmatrix} I_x^2 & I_x I_y \\ I_x I_y & I_y^2 \end{bmatrix} \right)}_H \begin{bmatrix} u \\ v \end{bmatrix}$$

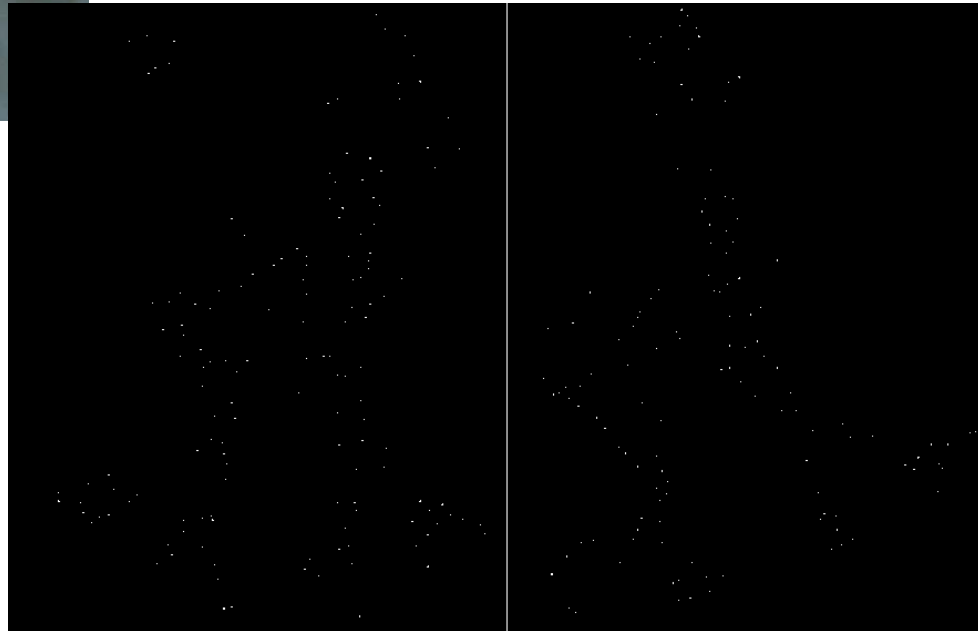


- Eigenvalues and eigenvectors of H
- Choose points where smaller eigenvalue is a local maximum as interest points
- Rotation and brightness invariance
- Harris corner detector
- Scale invariance using a pyramid

# Highlight



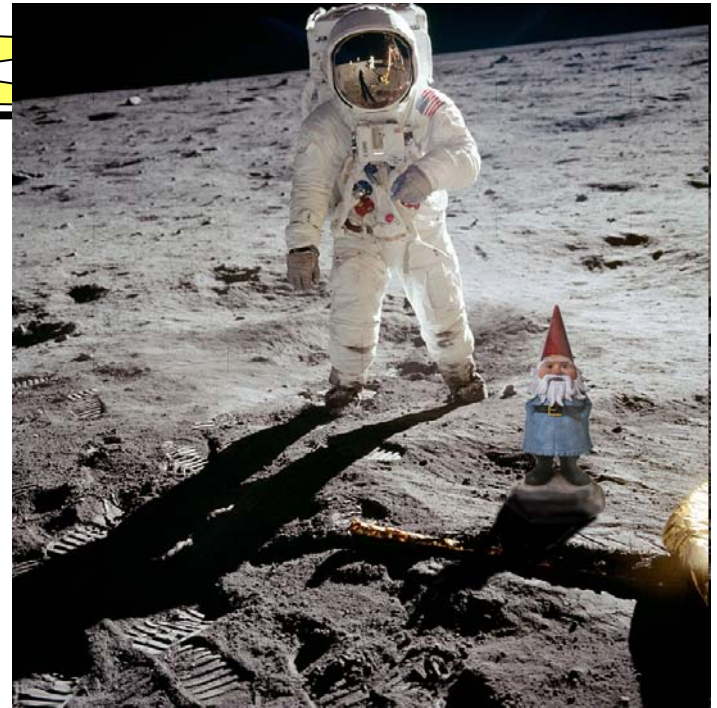
Harris  
detector  
example



# Other Highlights



Your work!





# Features and Image Matching

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- Feature Descriptors
  - MOPS
  - SIFT (just the basic idea of orientation histogram)
- Feature matching
  - SSD
  - Ratio of SSDs
- True positives and false positives
- ROC curve
  - How is it computed?
  - How is it used?

# Image Stitching and Panoramic Mosaics

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- Go back and review the steps you used in your project 2 (Panorama)
  - Warp to spherical coordinates
  - Feature extraction and matching
    - Note: The correct definition for the original Harris detector is the one in the Feature detection lecture
  - Aligning using RANSAC
  - Correcting for drift and blending

Highlight

Your Panoramas!



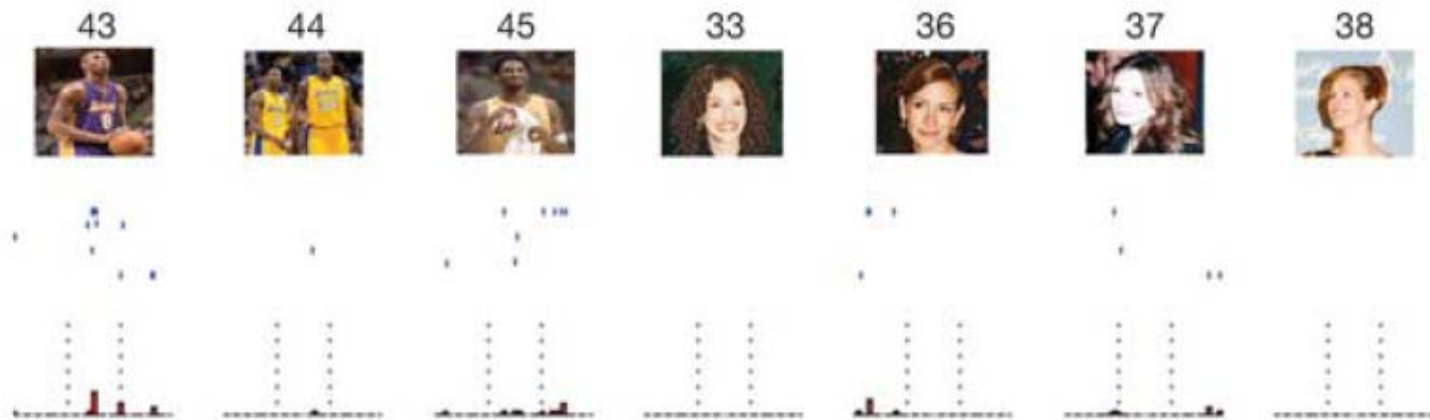
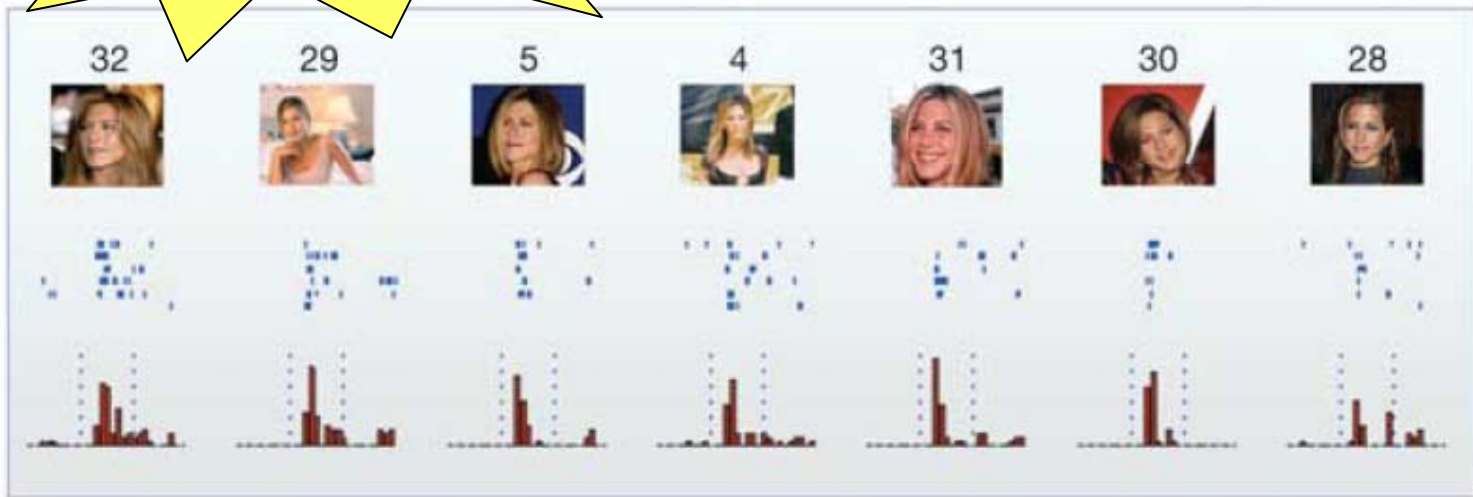
# Human Vision

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- Visible spectrum
- Retina
  - Rods, cones
  - Space-time-color filtering in the retina
- Eye movements
- Visual pathway
  - Simple, complex receptive fields
  - Orientation, direction, depth selectivity
  - Hierarchical organization
  - “What” and “Where”
  - Face selective cells

Highlight

“Jennifer Aniston cell”



Cell in medial temporal lobe responded selectively to different images of JA  
Other cells were found that responded to Bill Clinton, Halle Berry...

(Quiroga et al., 2005)

# Human vision relies heavily on context

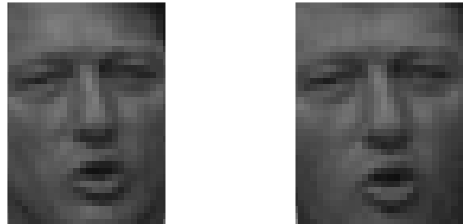
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Sinha and Poggio, *Nature*, 1996

# Human vision relies heavily on context

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Sinha and Poggio, *Nature*, 1996

# Change Blindness

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Something big is changing in this scene – what is it?

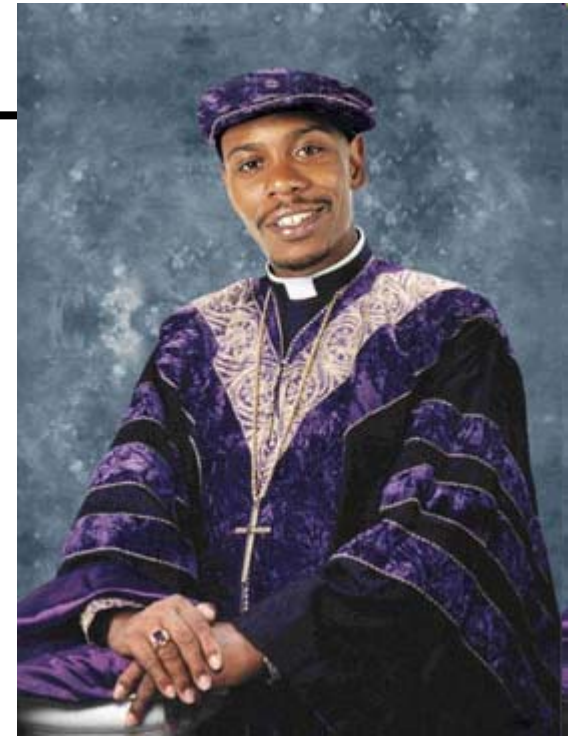


<http://www.psych.ubc.ca/~rensink/flicker/download/>

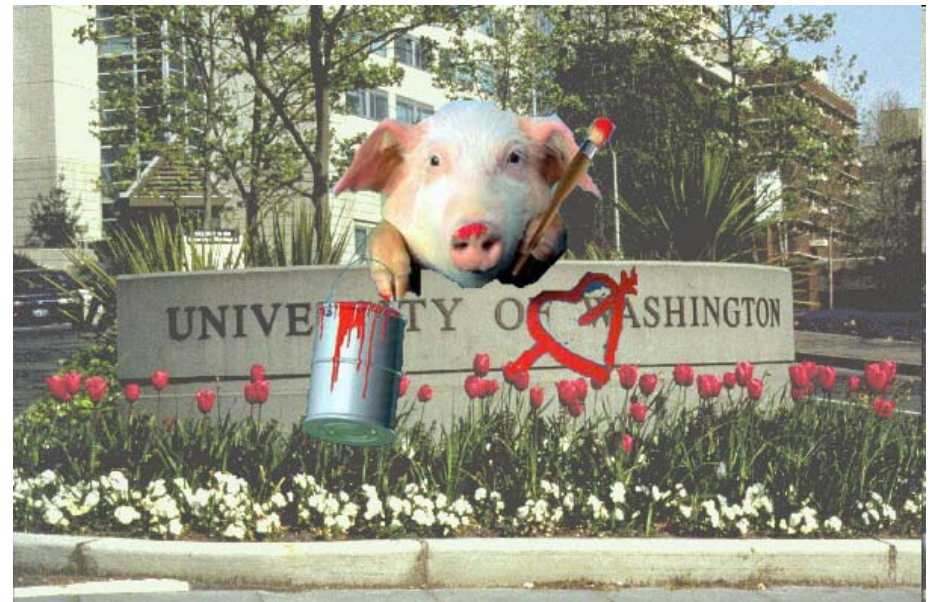
Attention is needed to perceive changes in scenes



# Other Highlights



Your work!



# Pattern Recognition and Learning

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- Decision tree learning
- Classification using nearest class mean
- Nearest neighbors and K-NN
- Perceptrons
- Multilayer Sigmoidal networks
  - Backpropagation learning (no need to derive)
- Max margin classification and SVMs
  - No need to derive but know the basic idea and intuition
  - Kernel trick
- Probabilistic (Bayesian) approach (skin classification example)
  - Histogram-based vs. parametric (Gaussian) PDFs
  - Bayes rule: prior, likelihood, and posterior
  - MAP vs. ML estimation

# Color

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- Color perception
- Color coding
  - RGB, color triangle, HSI, YIQ
- Color histograms for image matching
  - Histogram intersection
- Color constant color indexing

# Texture

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- Texture analysis
  - Edginess, histograms, Laws texture energy features
  - Autocorrelation and Fourier power spectrum
- Texture synthesis
  - Markov random fields (MRFs)
  - Synthesizing one pixel at time (read Efros & Leung paper)
  - Block-based synthesis (read Efros & Freeman paper for basic idea)
  - Other methods: Graphcut textures, image analogies (know the basic idea, details not necessary)

Highlight

# Texture transfer

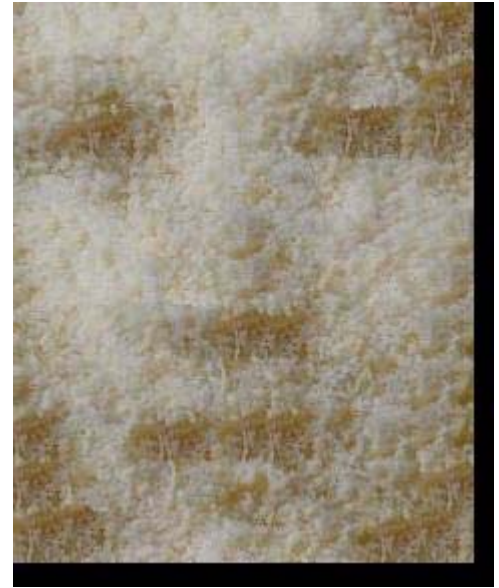


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# Highlight

## Image Analogies: Artistic Filters



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B



B'

Highlight

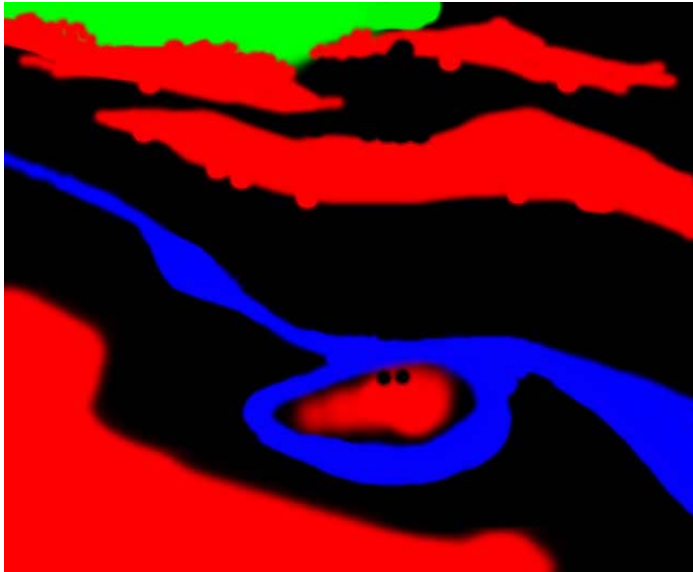
# Texture-by-numbers



A



A'



B



B'

# Segmentation

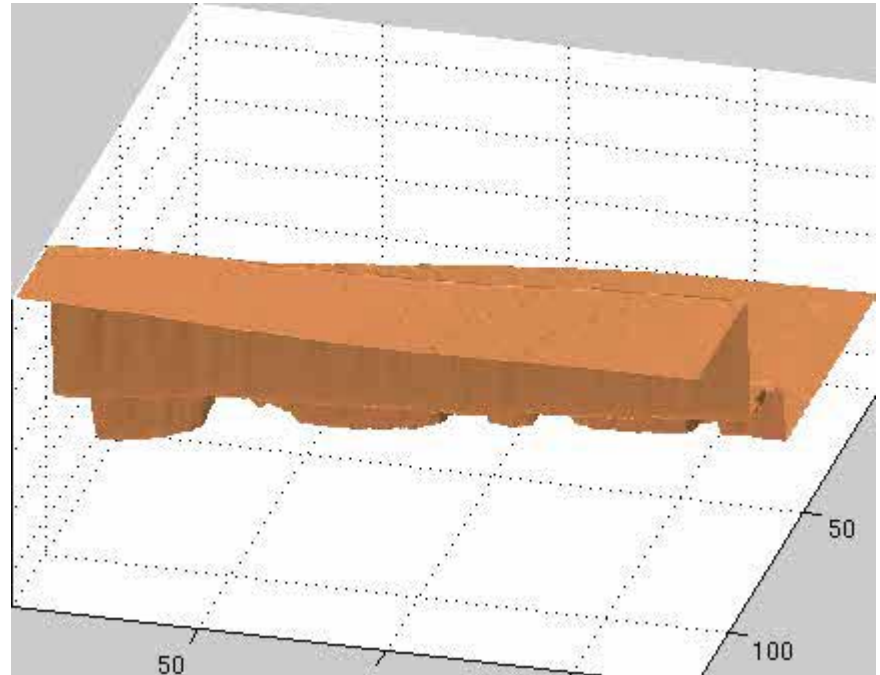
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- Histogram-based segmentation
- K-means clustering
  - EM algorithm
- Morphological operators
  - Dilation, erosion, opening, closing
- Graph-cut based segmentation
  - Normalized cuts (read Shi and Malik paper for basic idea and problem formulation, no need to know all the details)



# Highlight

## Graph-cut based segmentation



# Other Highlights



Your work!



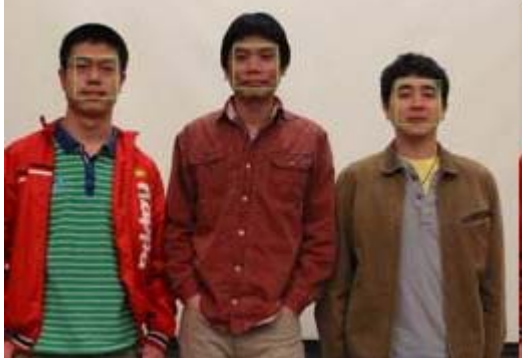
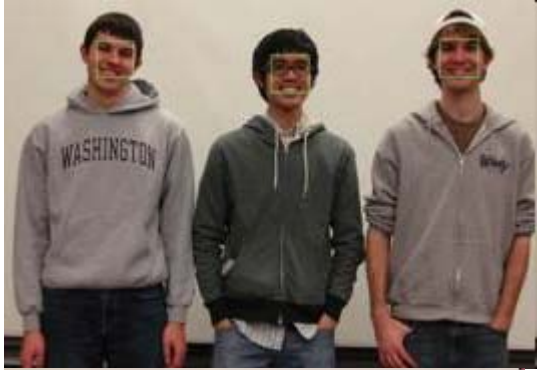
# Object Recognition

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- Principal Component Analysis (PCA) for Recognition
  - Linear subspaces and dimensionality reduction
  - Eigenvectors and eigenvalues of data covariance matrix
  - Eigenfaces, reconstruction using eigenfaces
  - Recognition and detection with eigenfaces
- Object class recognition by parts (Fergus et al. paper)
  - Constellation model
  - Know the general idea: Generative probabilistic model that includes Location, Scale, and Appearance of Parts; EM for learning
  - Browse Fergus et al. paper (no need to know the details or math derivations, just understand their basic generative model)

# Highlight

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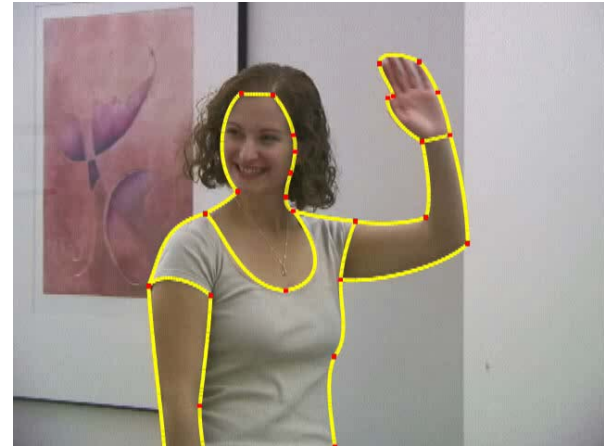
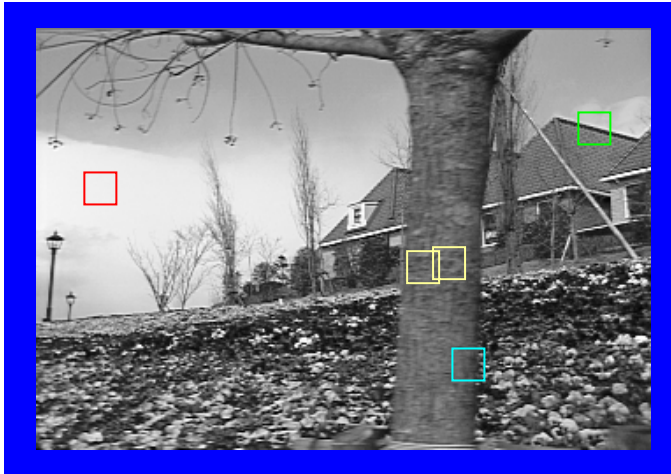
Your  
eigenfaces  
results  
here!

# Motion

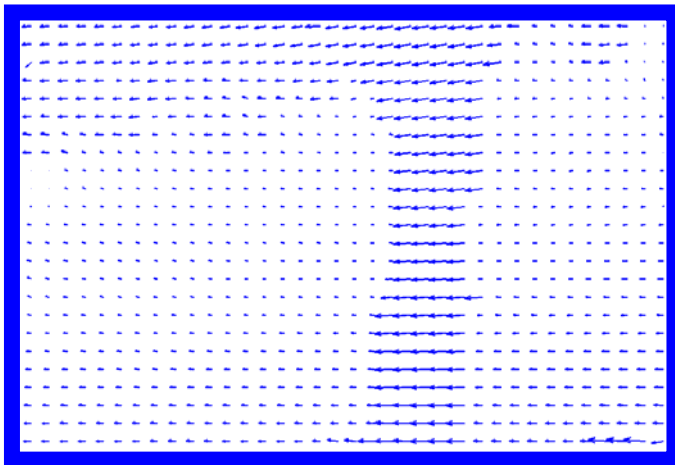
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- Optical flow problem
  - Brightness constancy and small motion assumptions
  - Aperture problem
  - Lucas-Kanade technique
  - Using higher-order terms
  - Iterative L-K technique
- Coarse-to-fine optical flow using a pyramid
- Applications: Structure from motion, MPEG
- Feature tracking
  - Example application: Rotoscoping (know the basic idea, no need to know the details)

# Highlights



Feature tracking and  
Rotoscoping



Optical flow

# Other Highlights



Your work!



# Stereo

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- Disparity and perception of depth
  - Stereoscopes, anaglyphs
- 3D from stereo: Triangulation method
  - Depth from disparity (derive from projection equations)
- Correspondence problem
  - Epipolar constraint
  - Stereo image rectification (know what it is, no need to know details from Loop and Zhang paper)
  - Window-based matching using SSD
  - Stereo correspondence as energy minimization
    - Know the two costs, no need to know details of actual algorithm
  - Sources of errors
- 3D reconstruction using structured light



# Highlights

## Video View Interpolation



(from "The Matrix")

# Applications

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## Content-Based Image Retrieval (CBIR)

- Browse through Prof. Shapiro's lecture slides
- Know how features and object recognition techniques can be applied to CBIR
- Understand the use of key images to speed up retrieval

## Tactile Graphics

- What is tactile graphics?
- How can computer vision techniques be used to solve problems in tactile graphics?
- Browse through Prof. Ladner's lecture slides

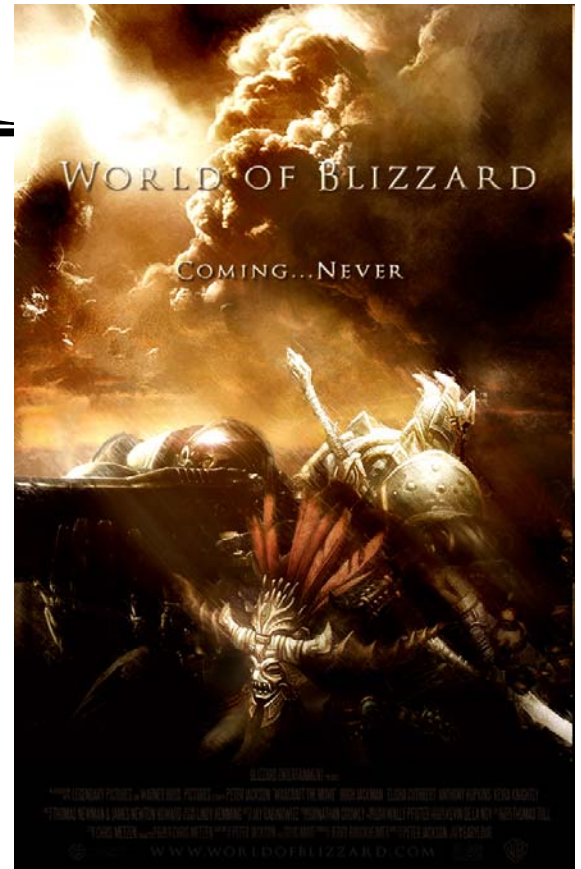
# Other Highlights



# Other Highlights



# Other Highlights



# ~~Other Highlights~~

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# Other Highlights



**Good luck on the final exam!**

