# ECS 174: Computer Vision, Final exam study-guide

#### 1. Linear filters

Image formation

- Light to discrete pixel arrays
- Images in Matlab

Image noise (as motivation for linear filtering)

• Types of noise

Correlation/convolution filtering with linear filters

- 1D and 2D examples
- Definition of correlation
- Implementation detail: handling boundaries
- Smoothing filters
  - Averaging/box filter
  - o Gaussian filter
  - o Impact of filter width in smoothing
  - o **Properties**
- Definition of convolution
- Examples: predicting linear filter outputs
- Properties of convolution
- Separability

Median filter, an example of a non-linear smoothing filter

### 2. Edge detection and image gradients

Computing image gradients with convolution

- Partial derivatives in x and y via finite differences
- Other popular finite difference derivative masks
- Properties of the gradient: direction, magnitude
- Impact of noise, counteracting with smoothing
  - o Smoothing and differentiating with derivative of Gaussian filters
  - Laplacian of Gaussian filter for edges
  - o Impact of smoothing scale on edges found
- Compare: mask properties for smoothing vs. derivatives
- Application with image gradients: seam carving

- Energy function definition
- o Greedy solution
- Optimal solution with dynamic programming

### 3. Edges, contours, and binary image analysis

Edge detection

- Basic pipeline: smooth, enhance, localize
- Thresholding a gradient image for edges
- Canny edge detector
  - o Non-maximum suppression
  - o Hysteresis
- Low-level edges vs. perceived contours

Chamfer matching: comparing shapes from edges

- Definition of the Chamfer distance
- Computing Chamfer efficiently with the distance transform
  - Definition of the distance transform
  - Efficiently computing the distance transform
  - Using distance transform to get Chamfer distances against template
- Properties of Chamfer matching

Binary image analysis

- Thresholding to create a binary image
- Morphological operators
  - o Dilation
  - o Erosion

### <u>4. Texture</u>

What is texture? What is it useful for?

Psychophysics of texture, "Textons" as primitive units of texture Texture representation

- Detect local patterns
- Describe their statistics
  - Example with gradients
- Compute distances in texture descriptor space
- Scale of a texture pattern

- Filter banks
  - o Example application
  - Name that filter bank response
- Applications of texture representations

Non-parametric texture synthesis

- Markov chains, Markov random fields
- Text synthesis example
- Analogy for texture generation in images
- Image quilting extension
  - o Minimum error boundary computation

#### 5. Segmentation and grouping

Grouping problems in vision Inspiration from human perception, Gestalt properties

Bottom-up segmentation via clustering

- Mode finding and mean shift
  - o k-means:
    - Algorithm definition
    - Pros and cons
    - Examples of feature spaces: color, intensity,...
    - Texton histograms; clustering for feature space quantization
  - o Mean-shift:
    - Algorithm sketch
    - Pros and cons
- Graph-based:
  - Images as graphs
  - Measuring affinity between nodes
  - Minimum cut
  - Normalized cuts; objective definition, pros and cons

#### Segmentation by cutting graphs

Graph definition

• Visualizing affinity matrices

Min cut and Normalized cut

- Objectives
- Pros and cons

#### 6. Fitting

Hough Transform for lines

- Image space vs. parameter space and concept of voting
- Complete Hough Transform algorithm for lines
- Examples: interpreting Hough space, strengths and weaknesses
- Extensions to the basic approach: esp., using gradient direction
- Computational complexity

Hough Transform for circles

- Parameterization
- Variants depending on (un)known gradient direction, radius
- Voting/Hough recap
- Strengths and weaknesses of the approach

**Generalized Hough Transform** 

- Offline training for arbitrary shape model
- Detection of shapes in novel images
- Applicability beyond shape fitting

#### 7. Fitting with deformable contours

Motivation for deformable contours:

- Beyond explicit models, need to fit unpredictable shapes
- Want to express loose shape priors
- Non-rigid, deformable objects; video
- Compare/contrast with Hough

Defining the energy functions

- External term to match image data: definition, example
- Internal term to enforce generic priors (smoothness, low curvature): definition, example
- Role of the weights in the energy function

Minimizing the energy function

- Greedy search
- Dynamic programming Viterbi algorithm (for a chain)

Summary of pros and cons

### 8. Alignment and fitting

- Feature-based alignment
  - Definition of alignment, place in fitting
  - 2D transformations
    - Parametric transformations

- Representing with matrices, homogeneous coordinates
- Affine fit
  - Fitting with least squares given correspondences
  - What are the correspondences?
  - Number of correspondences needed?
  - Recognition by alignment
  - Problem: outliers
- RANSAC
  - Algorithm definition
  - RANSAC for line fitting (example)
  - Pros and cons
- Homography and image reprojection
  - Perspective projection intuition
  - Connection to image mosaics
  - Solving for homographies
  - Number of correspondences needed?
  - Applying with 2d image warping (forward, inverse)
  - Using homographies to manipulate images: rectification, stitching

#### 9. Local invariant feature detection and description

Overview of main components:

- 1. Detection: find the interest points
- 2. Description: extract a descriptor for each one
- 3. Matching: determine correspondence

Desirable properties: repeatability during detection, distinctiveness during description

#### **Detection**

Rotation invariant interest point detection

• Harris corner detection: review of measure, main steps, properties

Scale invariant interest point detection

- Intuition behind automatic scale selection
- Laplacian of Gaussian filter: 1d, 2d
- Characteristic scale selection
- Interest point detection across an image
- Laplacian approximated as DoG in practice

#### **Description**

Desired invariance properties: geometric and photometric transformations Simplest solution: SSD on patch intensity SIFT descriptor

- Definition
- Rotation invariance

### **Matching**

- Generating candidate matches
- Eliminating ambiguous matches
- Robust alignment pipeline using detection, description, and matching.
- Applications

### 10. Indexing local features and applications to visual search

- Visual words: Quantizing local descriptor space
  - o Connection to textons
  - Issues in visual vocabulary formation
- Bag of words representation and matching
  - Comparing bags of words
  - Tf-idf weighting
  - o Scoring retrieval results with precision recall
  - o Pros and cons

### **11. Instance recognition**

- Motivation: visual search
- Visual words
  - Feature space quantization (recap)
  - Indexing with inverted file
  - Issues in visual vocabulary formation
  - o Bag of words representation and comparisons
  - Vocabulary trees for large vocabularies
  - Pros and cons of visual words
- Spatial verification strategies, using affine transformation
  - o RANSAC
  - o Example applications, evaluating retrieval results with precision recall
  - o Generalized Hough transform
  - Tradeoffs between RANSAC, GHT
- More text retrieval influence
  - Tf-idf weighting
  - o Query expansion

### **12. Generic category recognition**

Introduction to the problem: applications, challenges, state-of-the-art Basic pipeline of category recognition techniques Supervised classification

- Definition
- Example with skin color detection

### 13. Window-based models for generic object detection

Basic pipeline for generic category recognition with a window-based representation

- Representation choice: pixels, gradients, colors
- Discriminative classifier choice
- Sliding windows to generate candidates
- Classifier to score candidates

Boosting classification algorithm

- Intuition
- Weights on training samples
- Training process
- Pros and cons

Viola-Jones face detector: a prime example of window-based object detection

- Overview
- Feature definition: rectangular features, integral images
- Selecting discriminative features among all candidates with Adaboost
- Attentional cascade of classifiers
- Summary of detector pipeline
- Example results

Strengths and weaknesses of window-based detection paradigm

### 14. Discriminative classifiers for image recognition

- Nearest neighbors and k-nearest neighbors
  - Definition of algorithm
  - Example for scene/location recognition, global "Gist" descriptor
  - Pros and cons
- Support vector machines (SVM)
  - Large margin motivation
  - o Definition of algorithm for linear case, solution

- o Non-linear SVMs and the kernel trick
- o Multi-class classification via binary SVMs
- Example for gender classification
- o Pros and cons

## 15. Deep learning

- Linear perceptron
- Simple backpropagation example
- Multi-layer neural network
  - Neural network properties
  - o Detection (testing) process
- Convolutional neural network
  - o Convolution, Non-linearity, spatial pooling, feature maps