

Image Processing for Object Recognition

Integrated Seminar Intelligent Robotics

Daniel Ahlers
2ahlers@informatik.uni-hamburg.de

University of Hamburg
MIN Faculty
Department Informatics

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Table of Contents

1 Introduction

2 Edge Detection

Canny Edge Detector

3 Object Recognition

SIFT

Feature Detection

Feature Description

Feature Matching

SURF

Feature Detection

Feature Description

Feature Matching

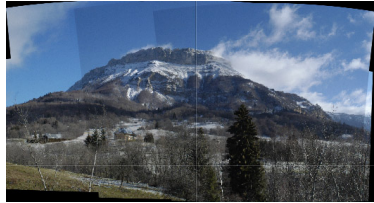
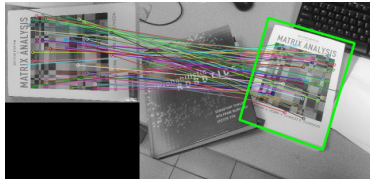


Object Recognition

- Identification of objects
- By sound, touching or image processing
- Faces, pedestrians or objects



Object Recognition



Problem

- Identifying object by pixels is not very useful
 - Different lighting
 - Different color
 - Other perspective
 - Rotated
 - Different scaling
 - ...



What is an Edge?

- An edge is a line
- Change in color, brightness or structure
- Straight or curved



Canny Edge Detector

- Algorithm to detect edges in 2D-images
- By John F. Canny in 1986 [Canny, 1986]
- Can only handle grayscale pictures



Canny Edge Detector

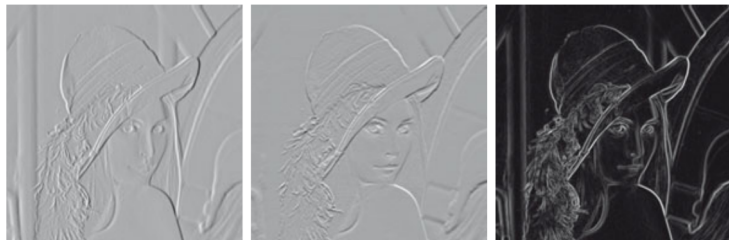
1 Apply Gaussian filter



[Nischwitz, 2011]

Canny Edge Detector

② Find intensity gradients



[Nischwitz, 2011]

Canny Edge Detector

- 3 Apply non-maximum suppression
- 4 Apply double threshold
- 5 Track edges by hysteresis



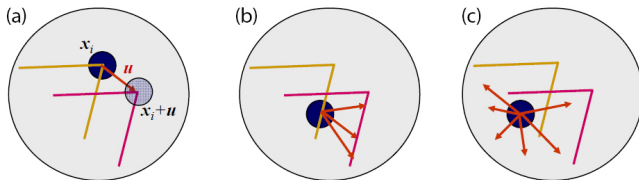
[Nischwitz, 2011]

Object Recognition by Edges

- Not useful
- Other perspective
- Rotated
- Different objects with same edges



Why not use Edges?



[Szeliski, 2010]

Features

- A feature is a point to describe the object
- Corners
- Crossing of edges
- Regions with constant properties
- Also called interest points



Object Recognition

- Feature detection
- Feature description
- Feature matching



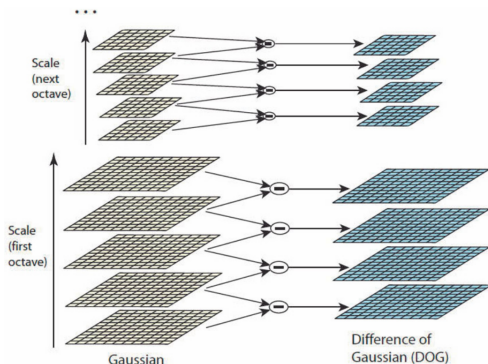
SIFT

- Scale Invariant Feature Transform
- By David Lowe in 2004 [Lowe, 2004]
- Can handle:
 - Different scales
 - Changes in viewpoint
 - Rotation
 - Noise
 - Different illumination



Feature Detection

- SIFT uses the difference of Gaussian (DoG)



[Lowe, 2004]

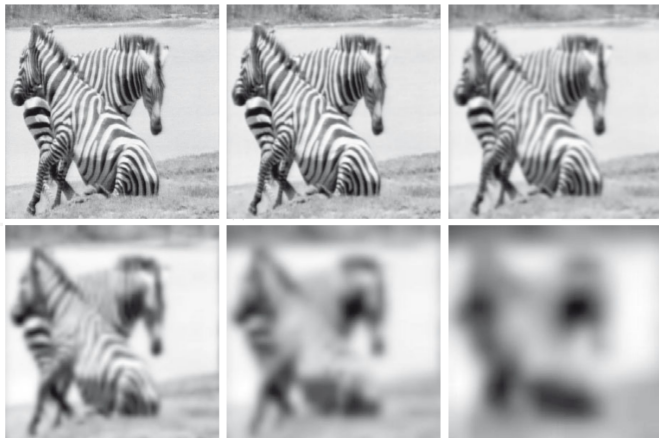


Feature Detection



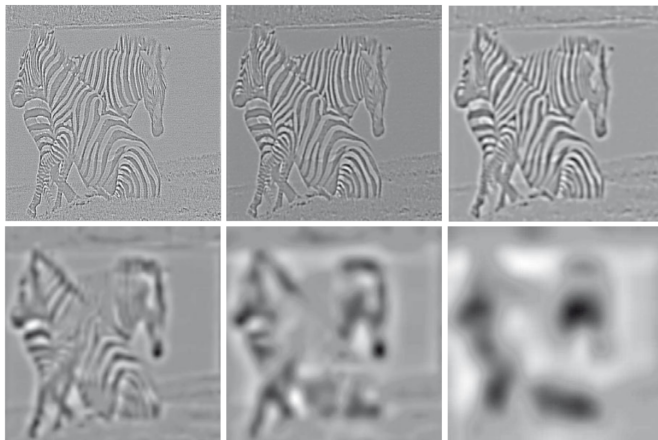
[Nischwitz, 2011]

Feature Detection



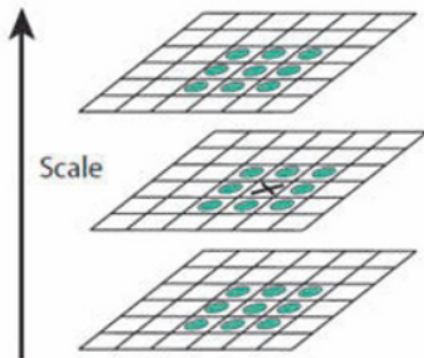
[Nischwitz, 2011]

Feature Detection



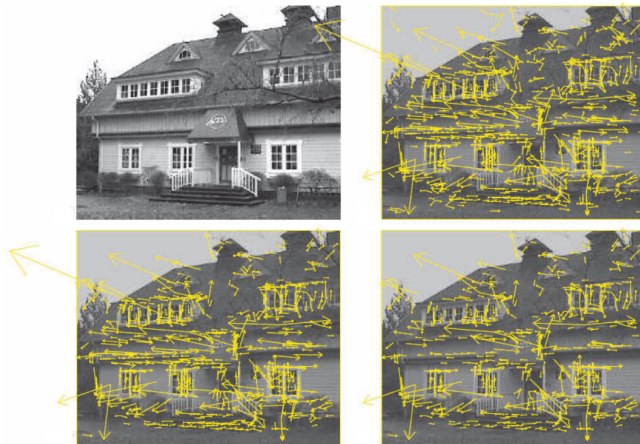
[Nischwitz, 2011]

Feature Detection



[Lowe, 2004]

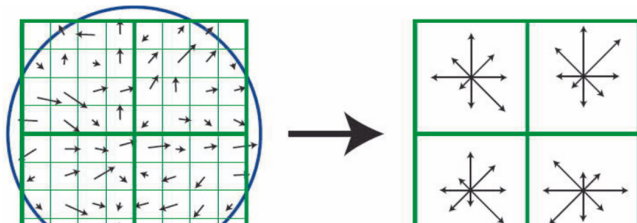
Feature Detection



[Nischwitz, 2011]

Feature Description

- 16x16 pixels around keypoint
- 4x4 groups with 4x4 pixels
- For each pixel: gradient with 36 directions
- Grouped with 8 directions
- Normalized and saved in 128 dimensional vector



Feature Matching

- Compared by Euclidean distance
- Second closest at least 20% away



[Lowe 2004]

SURF

- Speed Up Robust Features
- By Herbert Bay, et al. in 2006
[Bay et al., 2006]
- Can handle:
 - Different scales
 - Changes in viewpoint
 - Rotation
 - Noise
 - Different illuminations



Feature Detection

- SIFT uses the determinant of Hessian(DoH)
- ① Calculate an integral image

$$I_{\Sigma}(x, y) = \sum_{i=0}^{x-1} \sum_{j=0}^{y-1} I(i, j)$$

Original

5	2	3	4	1
1	5	4	2	3
2	2	1	3	4
3	5	6	4	5
4	1	3	2	6

Integral

5	7	10	14	15
6	13	20	26	30
8	17	25	34	42
11	25	39	52	65
15	30	47	62	81

Feature Detection

- 2 Calculate Hessian matrix

$$H(x, \sigma) = \begin{pmatrix} L_{xx}(x, \sigma) & L_{xy}(x, \sigma) \\ L_{xy}(x, \sigma) & L_{yy}(x, \sigma) \end{pmatrix}$$

- The determinant can measure local change
- Points are chosen when the determinant is maximal



Feature Detection

- Scale is implemented by box filters
- Sizes: 9x9, 15x15, 21x21, 27x27 ...



Feature Description

Orientation

- circular area around the feature point
- Oriented with a Haar wavelet responses that is weighed by a Gaussian function
- The longest Vector defines the orientation of the feature



Feature Description

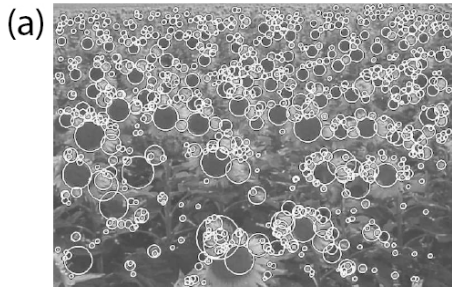
Description

- Square region around the feature
- Oriented along the orientation
- Split into 16 regions (4x4)
- Haar wavelet responses with 5x5 sample points for each region
- Summed up to a 4 dimensional vector
- All vectors combined to 64 dimensional vector weighed by a Gaussian function



Feature Description

Description



[Bay et al., 2006]

Feature Matching

- Compared by Euclidean distance
- Second closest at least 20% away



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