

Seminar: Processing Digital Camera Images Image Descriptors

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Image Descriptors

- 1 Introduction
- 2 Tiny Image Descriptor
- 3 Image Histograms
- 4 Gist Descriptor
- 5 Conclusion



Outline

1 Introduction

2 Tiny Image Descriptor

3 Image Histograms

4 Gist Descriptor

5 Conclusion



What are we going to be doing?

Searching through and analyzing **millions** of images to create a mosaic image, complete a scene, etc...



The inherent problems with large image databases

- Long access time to read all the image data
- Long processing time especially with a per-pixel analysis



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What are descriptors and why are they important?

Descriptor - A *piece* of stored information that is used to identify an item in an information storage and retrieval system.¹

- Descriptors save pertinent information, saving the processing time in future queries
- Descriptors can save image features that are essential for search and comparison

¹<http://wordnet.princeton.edu/perl/webwn?s=descriptor>



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How long would it take to read 1TB of data?

- In an ideal situation...
 - $\frac{1\text{Terabyte}}{100\text{Megabyte/s}} = 3 \text{ hours (Internal HD)}$
 - $\frac{1\text{Terabyte}}{100\text{Megabits/s}} = 23 \text{ hours (Ethernet)}$
 - $\frac{1\text{Terabyte}}{16\text{Megabits/s}} = 6 \text{ days (DSL-16000)}$
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One way to do it...

- ➊ generate small data-structures (descriptors) from the original images
 - *faster data access*
 - *reduced data (could be stored locally)*
- ➋ save the *interesting features* in the descriptor
 - *original would have to be transformed first*
 - *reduce future processing time*
- ➌ result: descriptor that can be compared with other descriptors
 - *calculate various distances (Euclidean, etc.)*
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What kinds of image descriptors are there?

- Meta-Information based: EXIF-Data, (Geo-)Tags... but are not interesting for this course
- Image based:
 - Pixel-based: Tiny-Images, Color Histograms, Texton Histograms
 - Structure-based: Fourier, Gradient, Geometric, Gist



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Tiny Image Source²



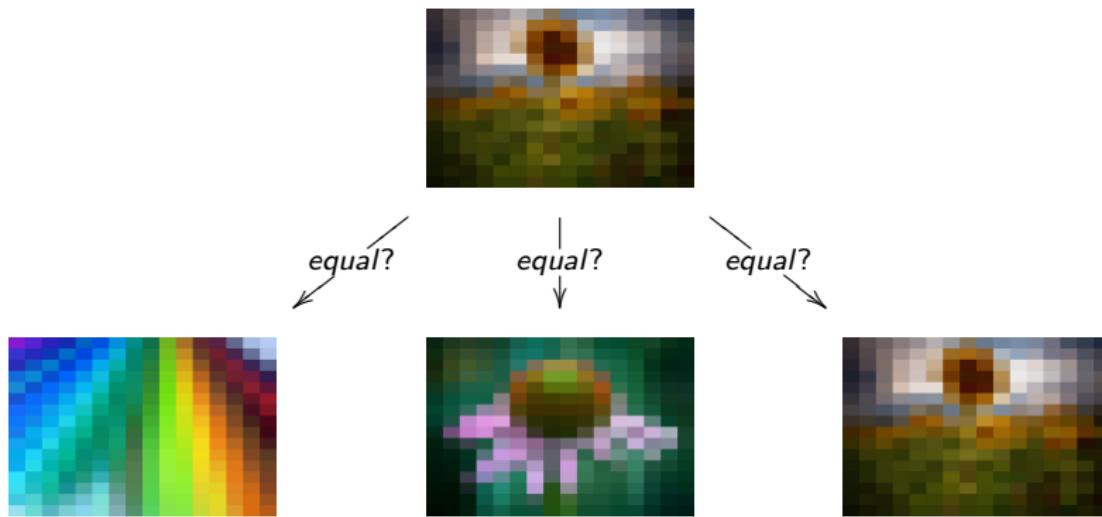
²from <http://www.flickr.com/photos/alphageek/2759567956/>



Tiny Image Descriptor



Comparison with database



³from <http://www.flickr.com/photos/40513596@N00/71762740>

⁴from <http://www.flickr.com/photos/alphageek/2993813155/>



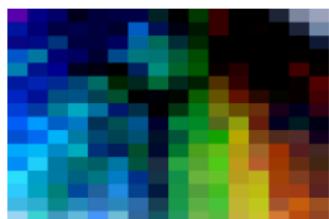
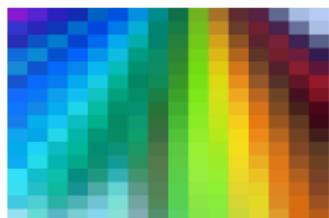
How to compare?



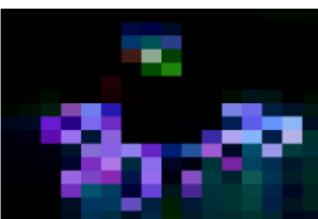
- ① Estimate value for each pixel
- ② Calculate difference to descriptor (using suitable colorspace)
- ③ Weigh higher errors by using square of the sum
- ④ Store the resulting difference
- ⑤ Sum up all differences
- ⑥ Take the smallest sum for best match



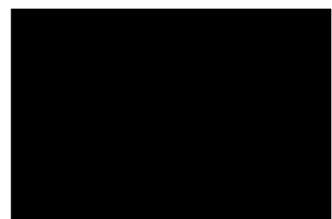
Comparision examples



14862080



437120



0



Tiny Images at a glance

Advantages

- simple
- good result on large datasets
- location dependent

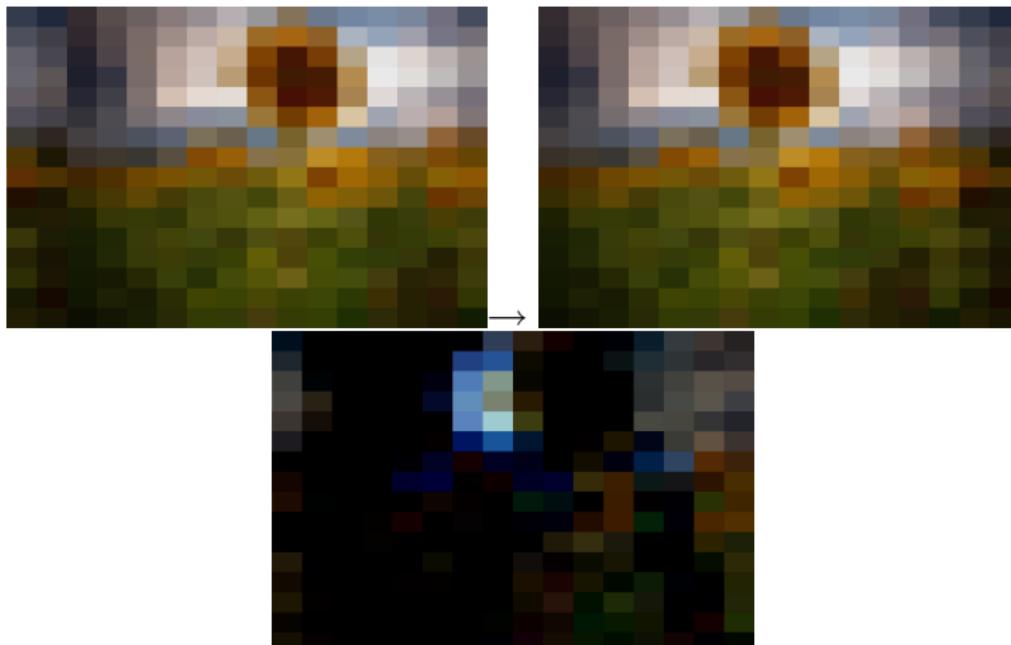
Disadvantages

- dependent on colorspace
- only useful for color information



Location dependency

- Shifting the descriptor 2 pixel to right (warping)
- completely different result



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Image Histograms

What are image histograms?

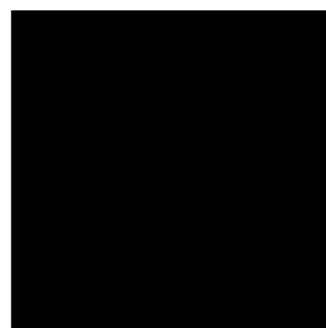


Several Types of Image Histograms

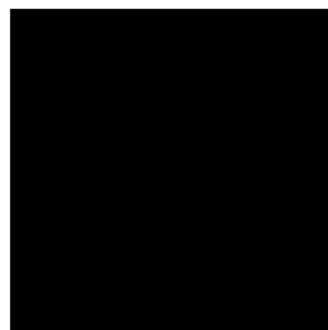
- Color histograms
- Texton histograms
- Line feature histograms
- ...



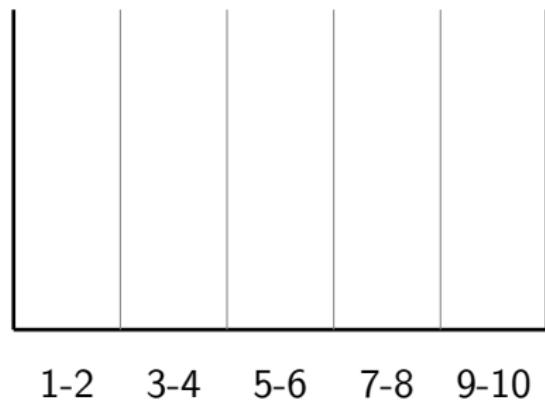
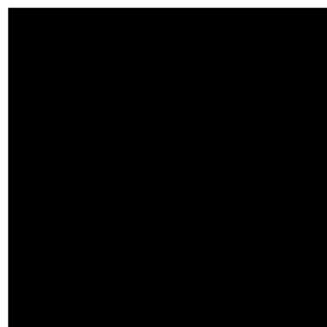
Definition of Color Histograms



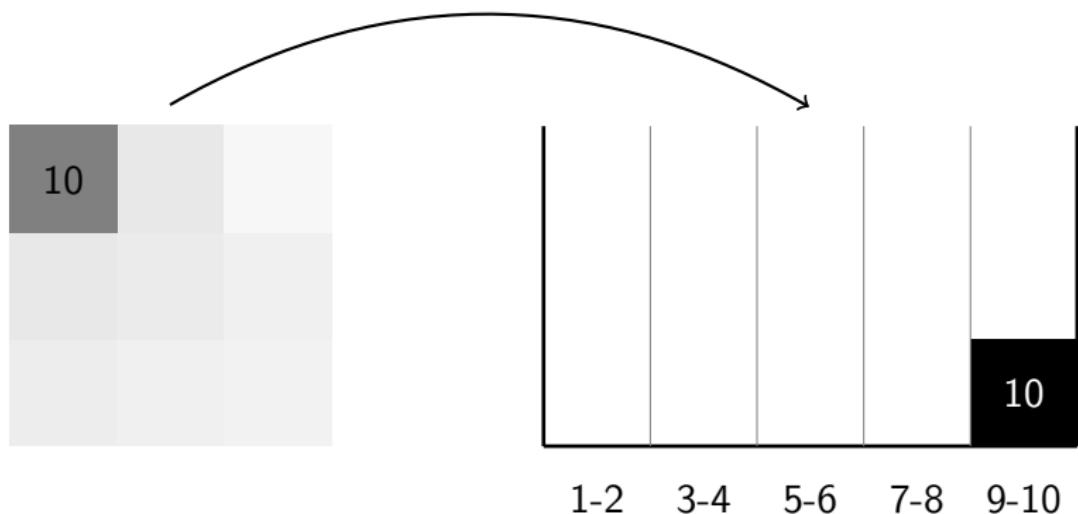
Definition of Color Histograms



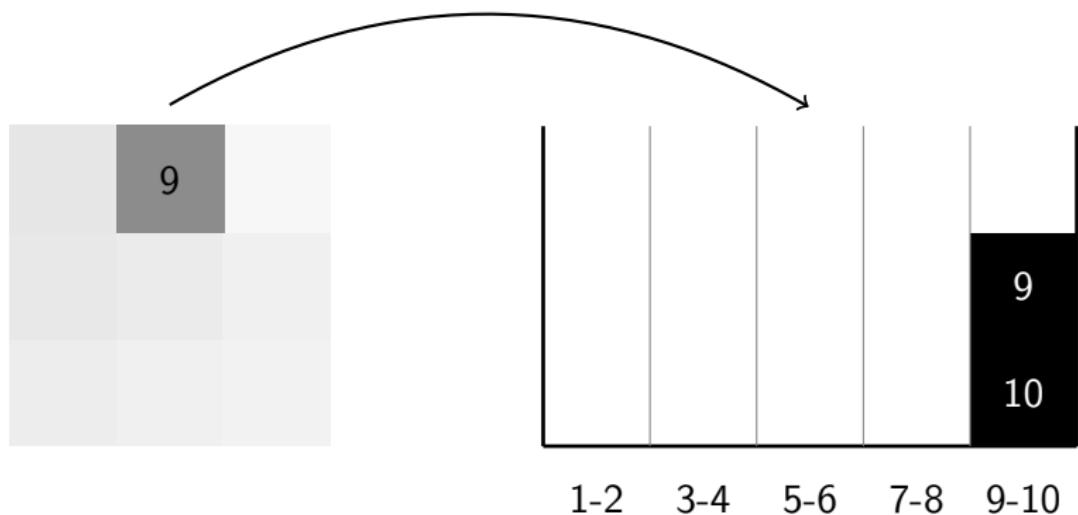
Definition of Color Histograms



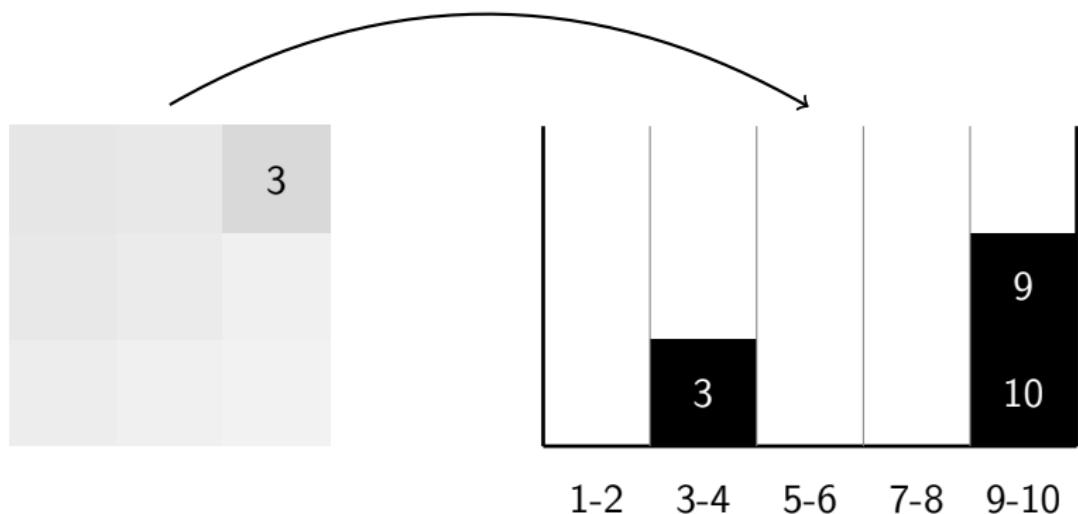
Definition of Color Histograms



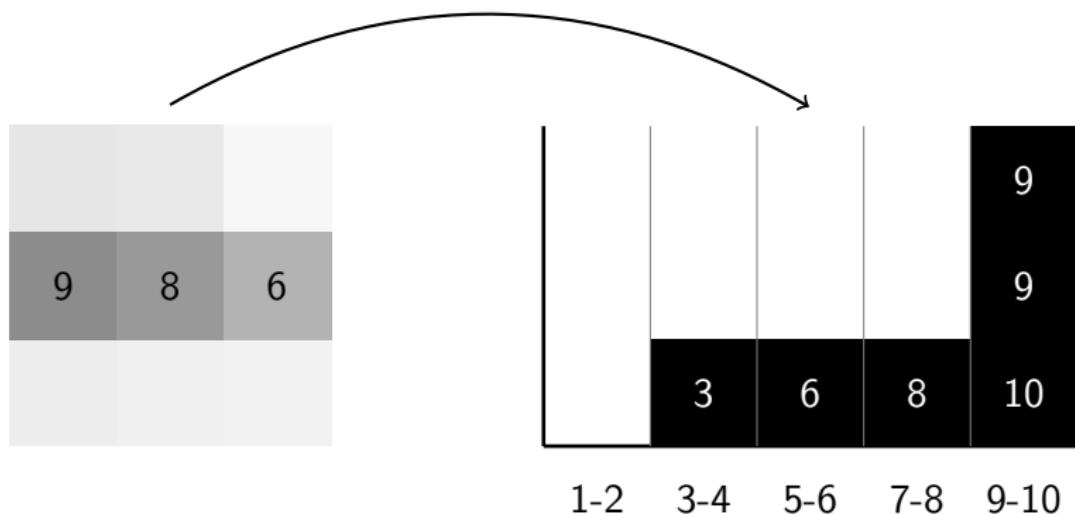
Definition of Color Histograms



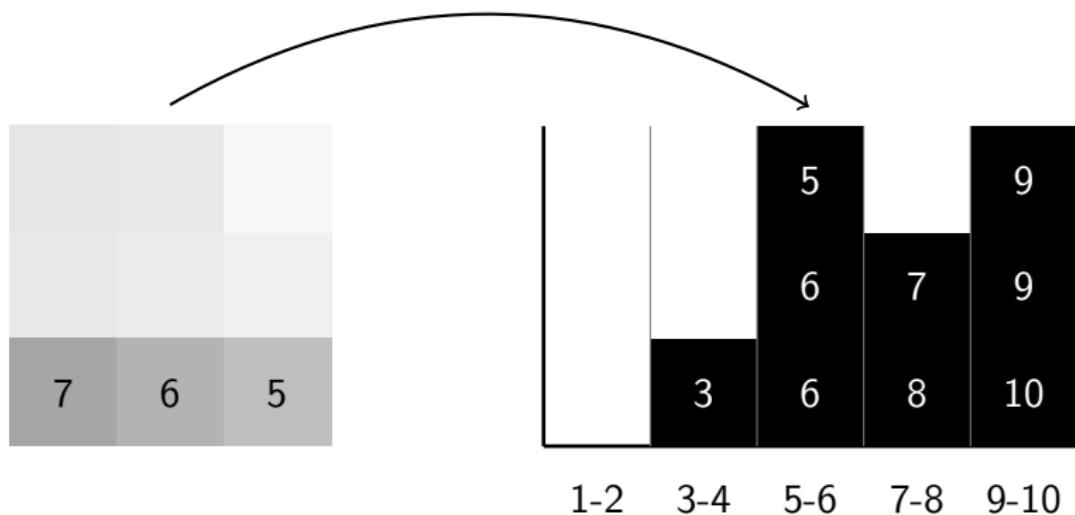
Definition of Color Histograms



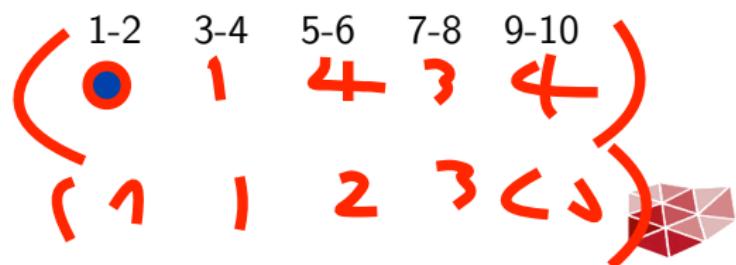
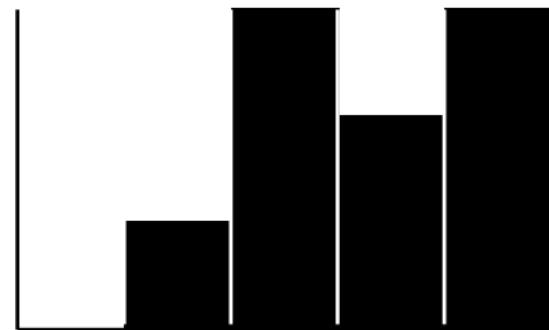
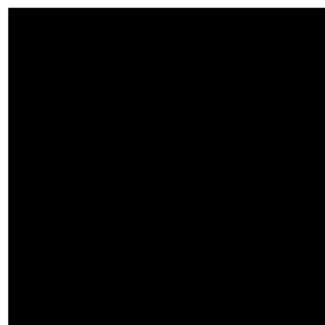
Definition of Color Histograms



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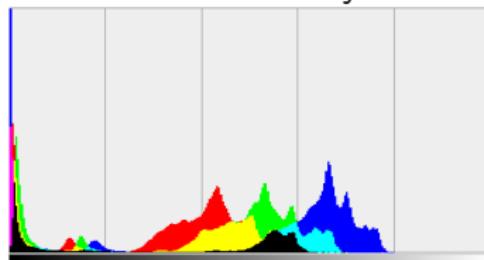
Definition of Color Histograms



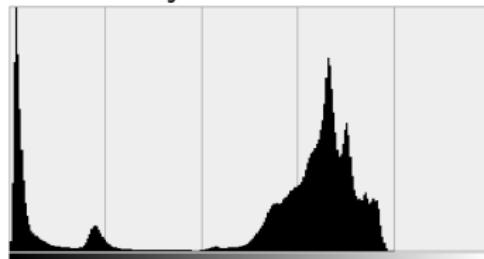
Examples



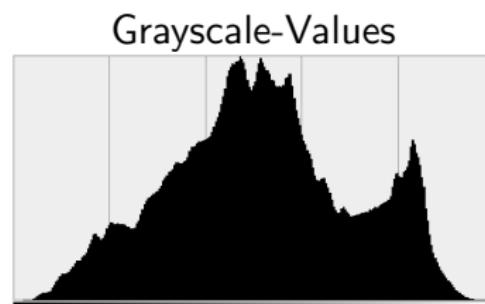
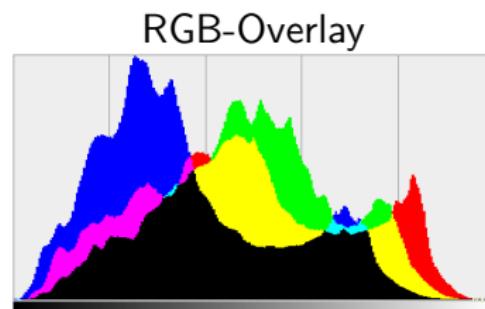
RGB-Overlay



Grayscale-Values



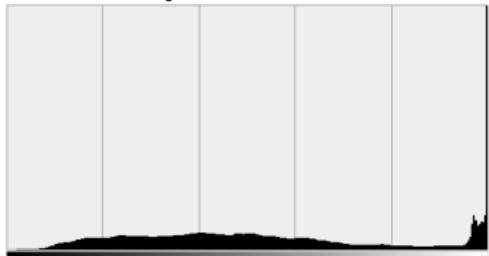
Examples



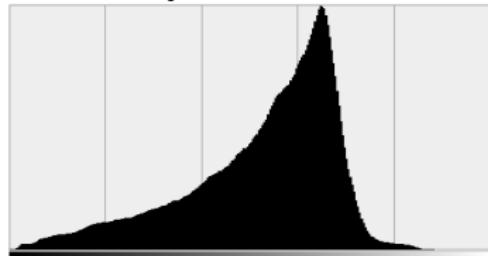
Examples



Grayscale-Values



Grayscale-Values



Advantages and Disadvantages

Advantages

- Translation and rotation invariant
- Colors often represent content

Disadvantages

- Translation and rotation invariant
- No consideration of features and structures

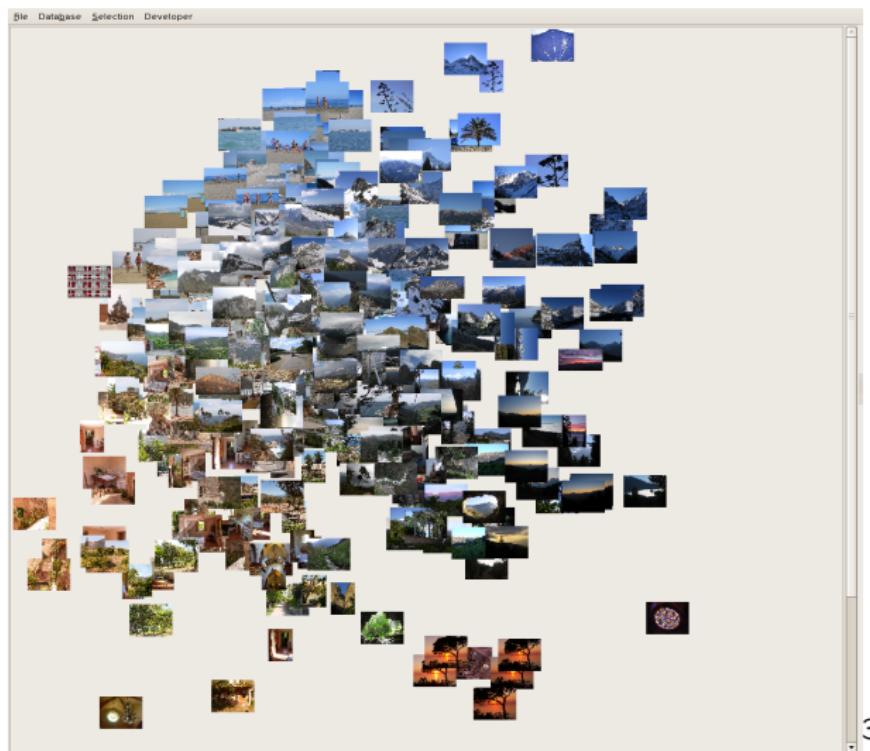


Distance between Color Histograms

- Simple approaches (SSD, ...)
- Multi-Dimensional-Scaling



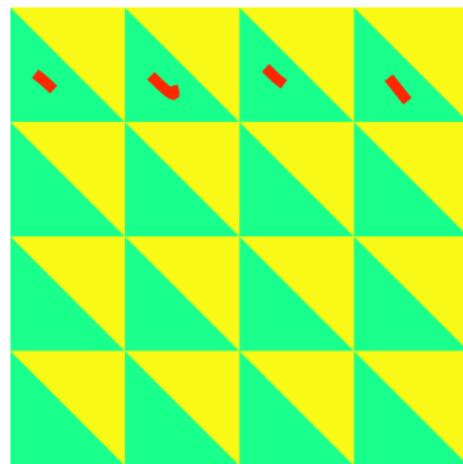
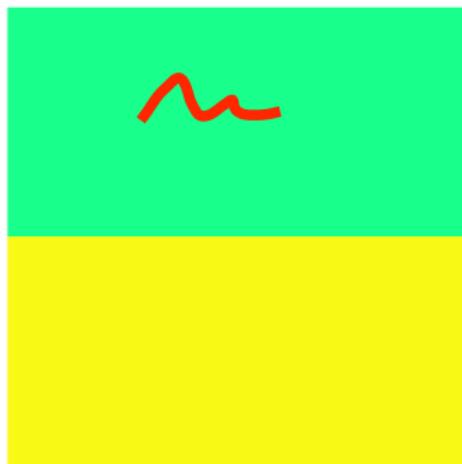
Use of Color Histograms



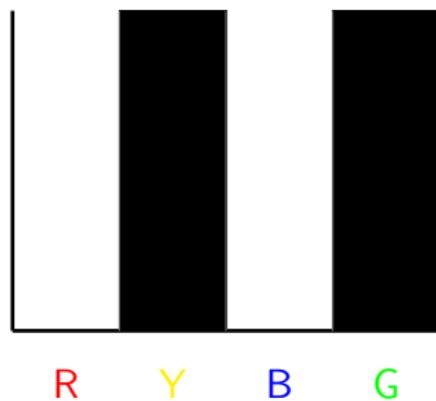
³<http://lear.inrialpes.fr/src/yorg/>



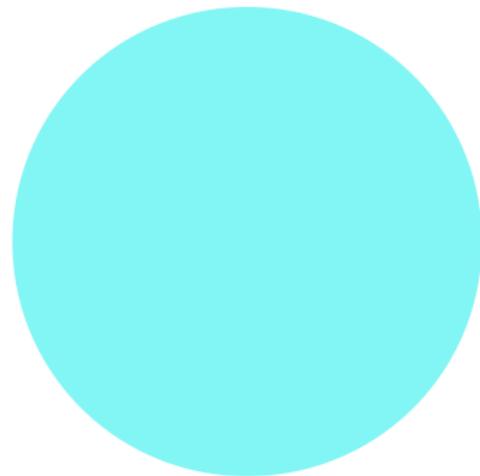
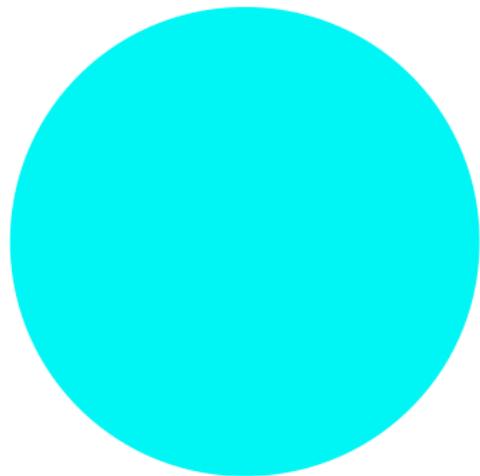
The Problems of Color Histograms



The Problems of Color Histograms



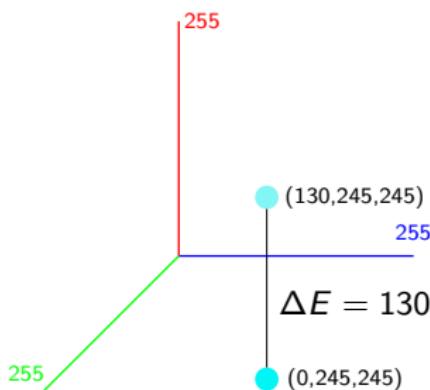
The Importance of Color Spaces



The Importance of Color Spaces

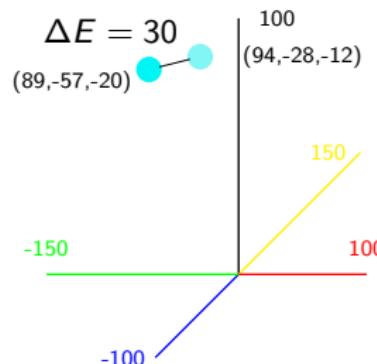
RGB

- Models the output on phys. devices
- Euclidean distances are pointless



Lab

- Designed to approximate human vision
- Euclidean d. represents human perception



$$\Delta E = \sqrt{\Delta L^2 + \Delta a^2 + \Delta b^2}$$



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Gist Descriptor

- **gist** = quintessence; general idea
- created for recognition of similar scenes,
like mountains, tall buildings, streets ...
hence the name
- structure, without color information



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Different Scenes



General Idea

Idea:

Partition the image to anchor the structure of the subimages to their location in the image.

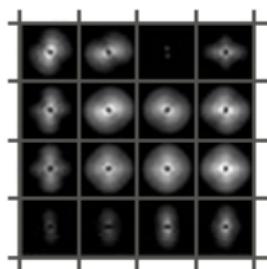




split



fourier
transformation



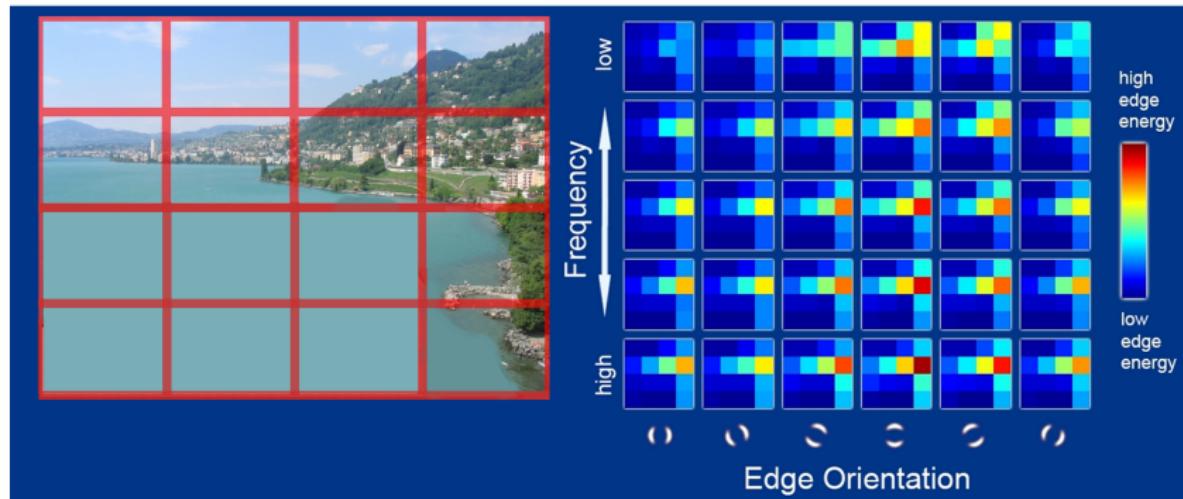
multiscale
oriented
filters

$n \times n \times k$ -vector
=

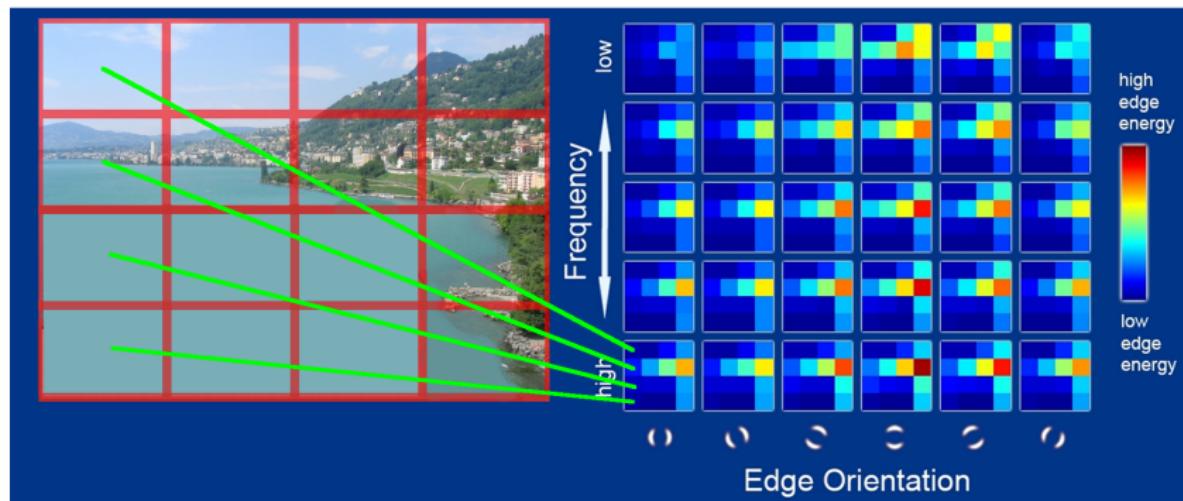
gist-descriptor

$$n \times n = \# \text{ partitions}$$
$$k = \# \text{scales} \times \# \text{orientations}$$

Gist Descriptor

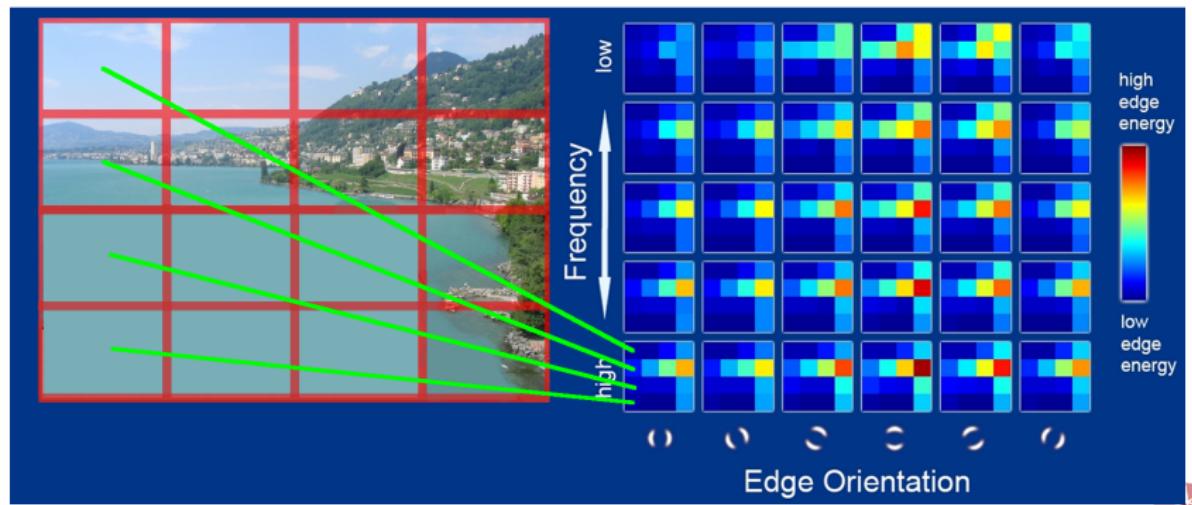


Gist Descriptor

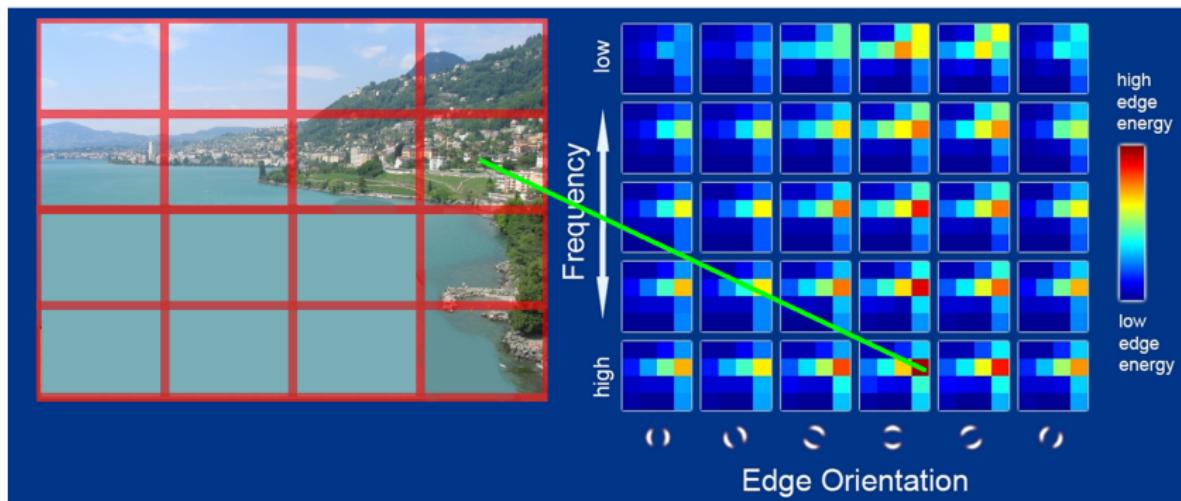


Gist Descriptor

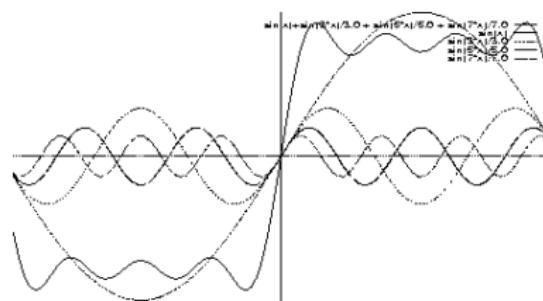
$n \times n \times k = 4 \times 4 \times 5 \text{ frequencies} \times 6 \text{ orientations}$
= 480-dim. Vector



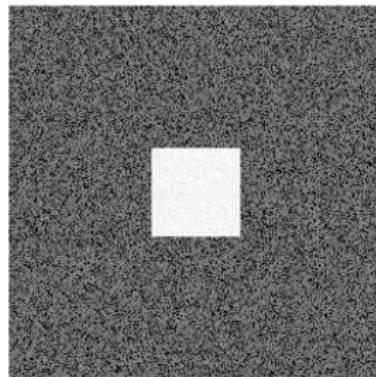
Gist Descriptor



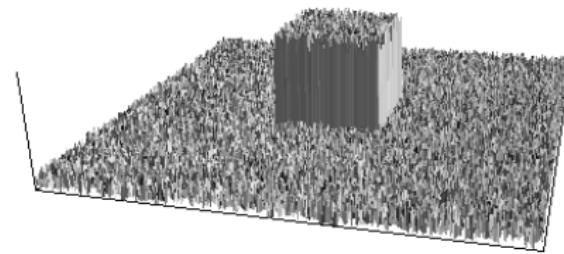
Fourier Transformation



rectangular pulse (with added noise)

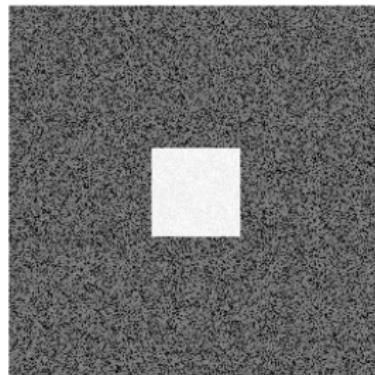


image

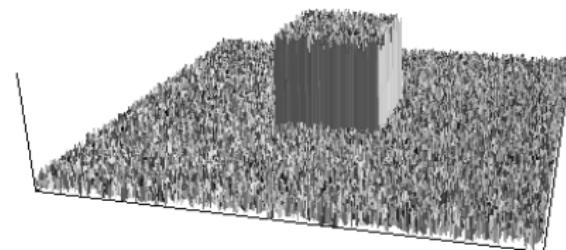


isometric view

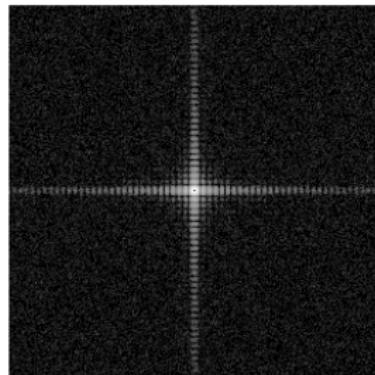
rectangular pulse (with added noise)



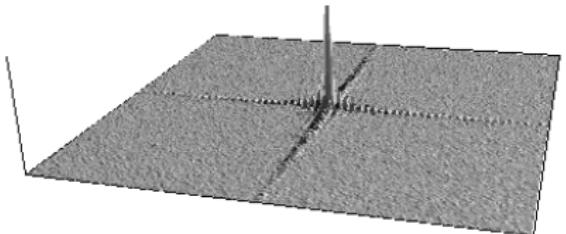
image



isometric view

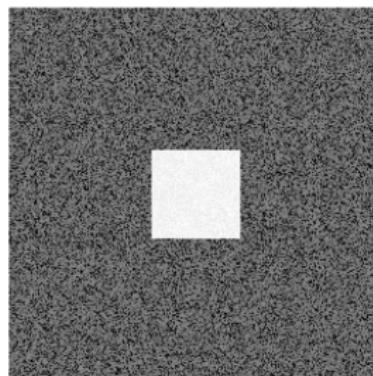


fourier transformation

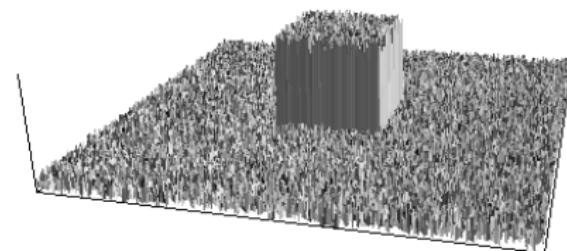


isometric view

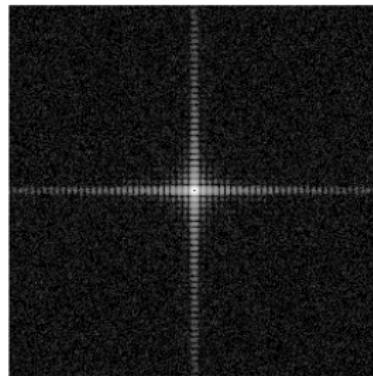
rectangular pulse (with added noise)



image

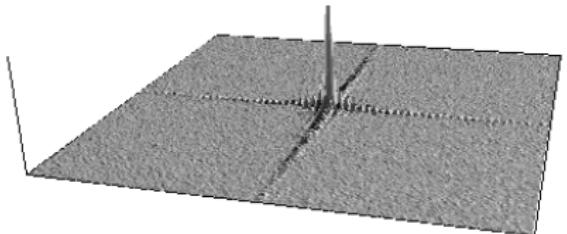


isometric view



fourier transformation

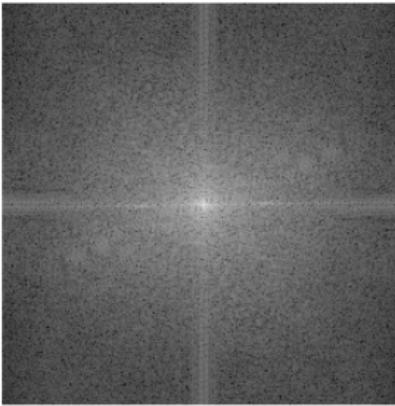
only frequency and amplitude!



isometric view



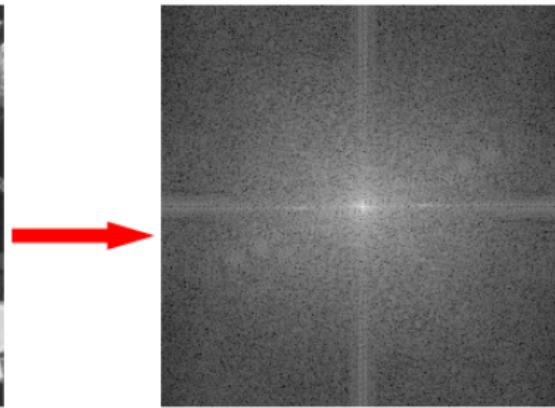
image



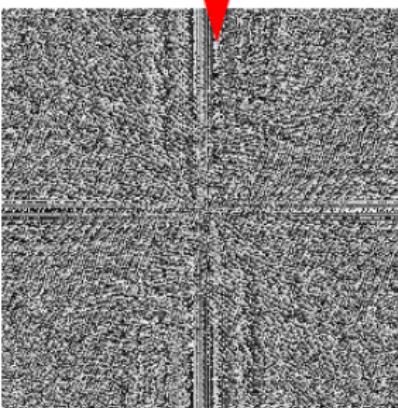
fourier transformation (amplitude over frequency)



image



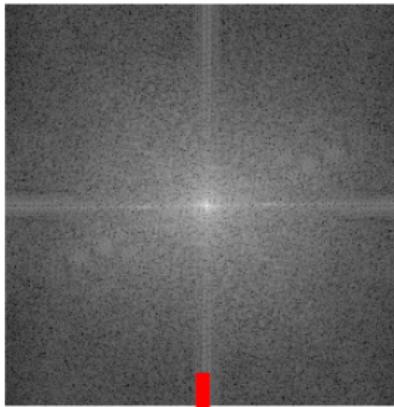
fourier transformation (amplitude over frequency)



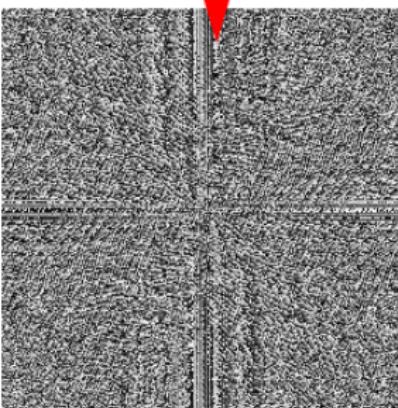
fourier transformation (phase over frequency)



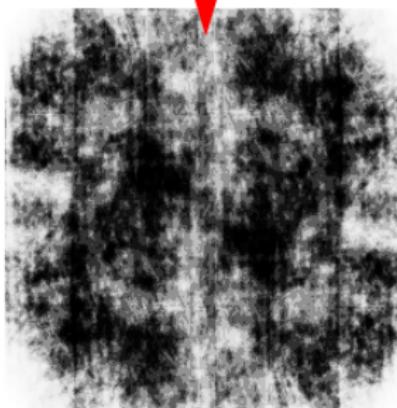
image



fourier transformation (amplitude over frequency)



fourier transformation (phase over frequency)



retransformation with phases = 0





fourier transformation



fourier transformation





fourier transformation



amplitudes



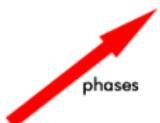
retransformation



fourier transformation



phases





fourier transformation



amplitudes

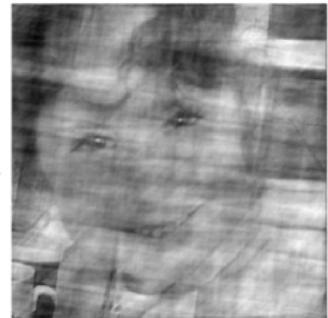
retransformation



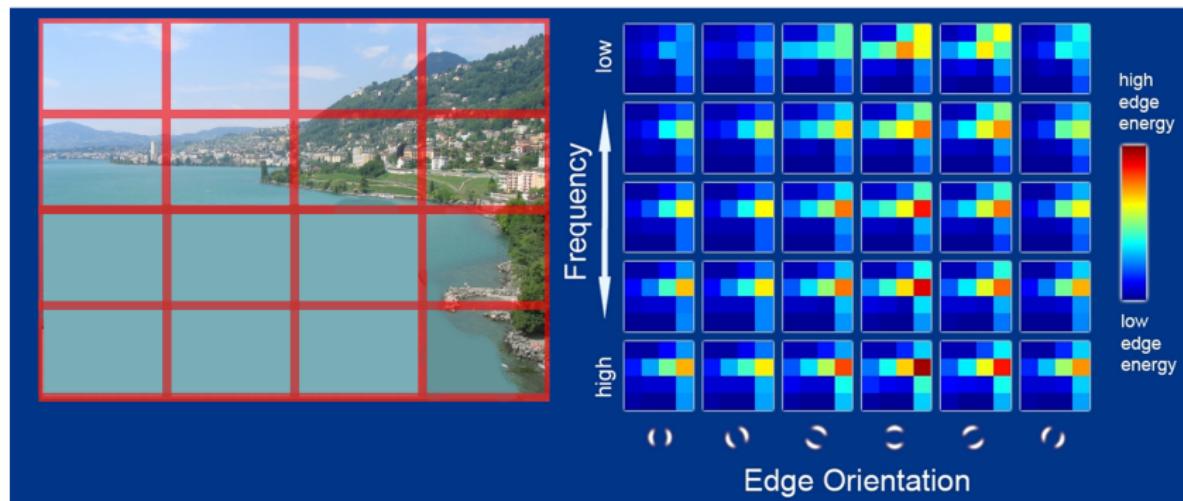
fourier transformation



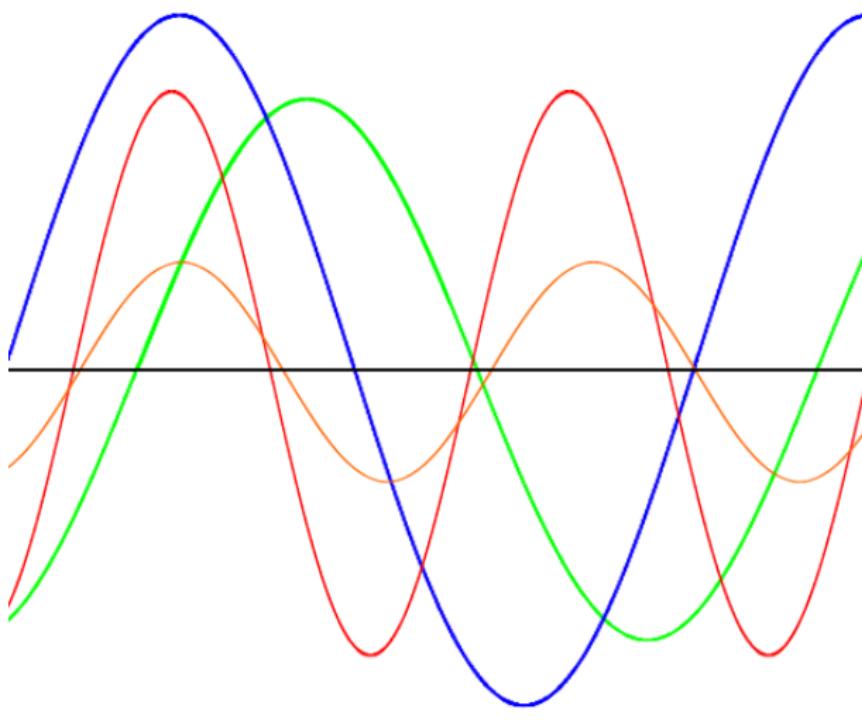
phases



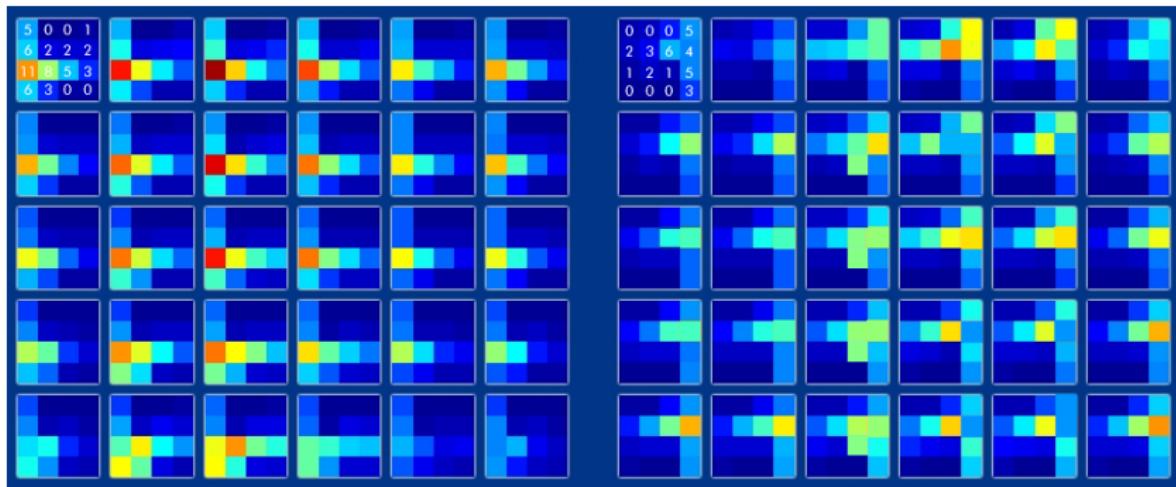
Gist Descriptor



How to cluster



SSD to compare gist descriptors











Advantages and Disadvantages

Advantages

- holistic approach, without segmentation or grouping operations
- get the gist of a scene in a small vector
- easy to store
- easy to handle with established algorithms
- fast

Disadvantages

- difficult implementation



Outline

- 1 Introduction
- 2 Tiny Image Descriptor
- 3 Image Histograms
- 4 Gist Descriptor
- 5 Conclusion



In Conclusion

- descriptors are very important for image retrieval
- there are different descriptors for different purposes
- descriptors are thus dependent on the problem



Comparison

Descriptor	Advantages	Disadvantages
Tiny images	Easy to calculate	No features
Histograms	Translation/rotation invariant	No local features
Gist	Preserves features and locales	Implementation

