

# CS 3570

# Final Project

# 2015

Ready for challenges ?

Part 1 (5/18)

# General Guideline



2~4 students form a team



Select a topic from our suggestion list, or think a new one yourself



1-page proposal on iLMS, before 5/28, including team name, members, project name, brief description and goal



If your proposal is not approved, you will know before 6/1  
Please hand in another one before 6/4

# General Guideline



Each topic difficulty is rated as stars



Team with more members are recommended  
to take a challenging topic



Hand in your report before presentation



Every team member needs to participate the presentation

# Project Requirement

- ▶ Realize the algorithms (Essential)
- ▶ Implementation (Essential)
  - Using open source code and toolbox are permitted
- ▶ Comparison and analysis (Suggested)
- ▶ Improvement (Quality, accuracy, speed, or others)(Optional)
- ▶ System for demonstration (Optional)

# Presentation in class



Afternoon in class



6/18 for the seniors

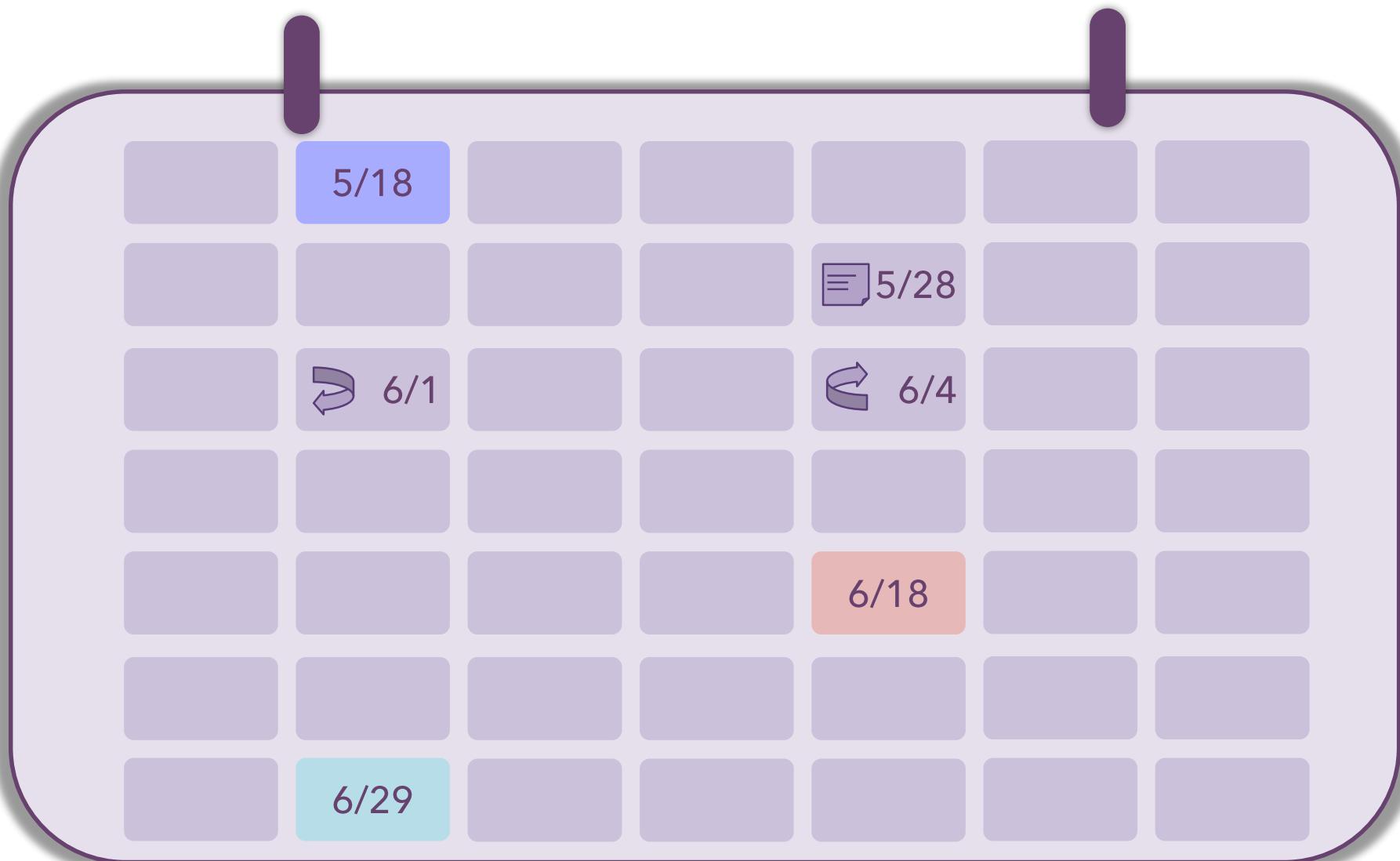


6/29 for the undergraduates



Best 3 teams get bonus points

# Vote for presentation day



# Project Categories

Image Processing

Machine Learning

Graphics

Audio Processing

Video Processing

# Image Processing

★ ★

Image denoising

★

Bilateral Filtering

★ ★ ★

Image matting

★ ★

Image inpainting

Poisson blending

★ ★

Seam carving

★ ★ ★

Special effects  
camera app

★ ★ ★

Rectangle rectification  
for document capture

★ ★

# Machine Learning

★ ★ ★

Face Recognition

★ ★ ★ ★

Content Based Image Retrieval

★ ★ ★ ★

Image Classification

# Graphics

★ ★ ★ ★

Vectorization

★ ★ ★ ★

OpenGL/Direct3D based game / application

# Audio Processing

★ ★

DTMF Recognition

★ ★ ★

Music genre classification

★ ★ ★

Chord Transcription

# Video Processing

★ ★ ★

Background subtraction

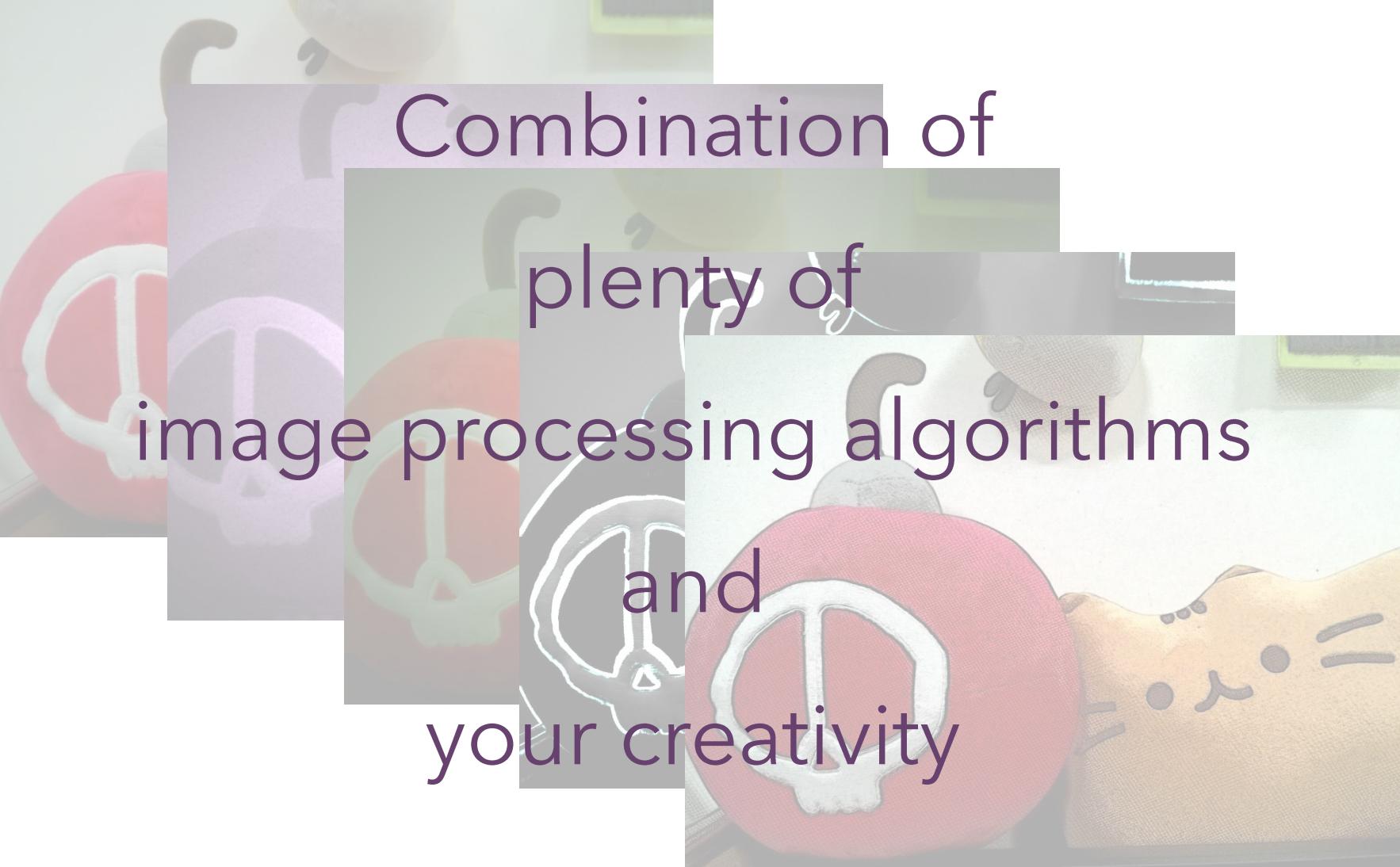
★ ★ ★

Video denoising

★ ★ ★

Video stabilization

# Special effects camera app



Combination of  
plenty of  
image processing algorithms  
and  
your creativity

# Special effects camera app



Original Photo



# Special effects camera app



Old time

Spotlight effect

Color Transfer

Salt-and-pepper  
noise



Powered by  
Camera 2



# Special effects camera app



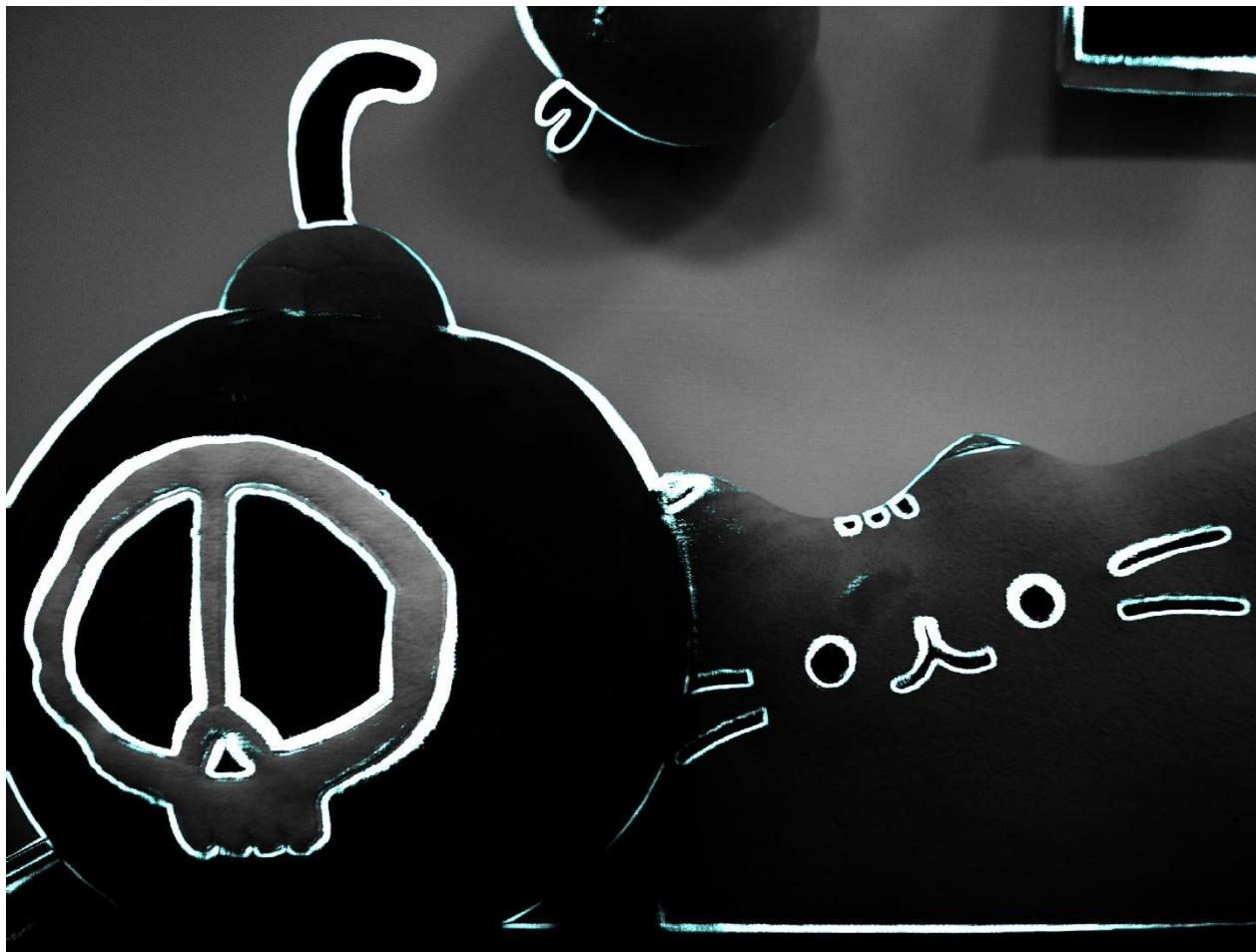
Lomo

Spotlight effect

Color Transfer



# Special effects camera app



Scientific

RGB to gray

Contrast  
Enhancing

Edge detection



# Special effects camera app



Comic  
(Non-photorealistic rendering)

Color Quantization

Add mesh texture  
in shadow regions

Edge detection



# Special effects camera app



Face detection

Face pose  
estimation

Render some  
template image  
onto the face



Powered by  
Hangouts

# Color transfer



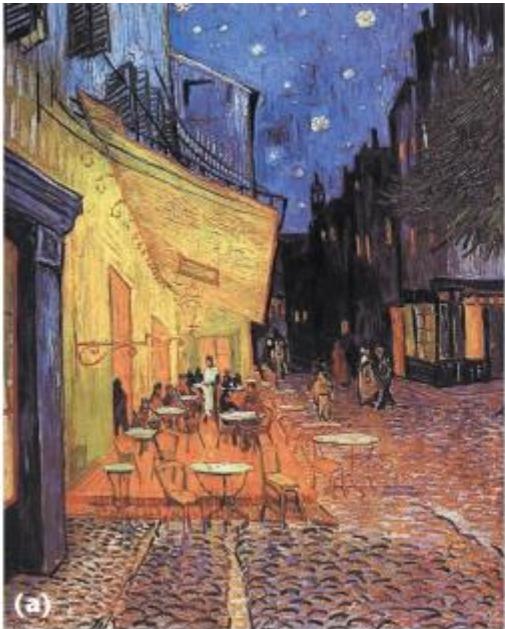
Target image



Reference image



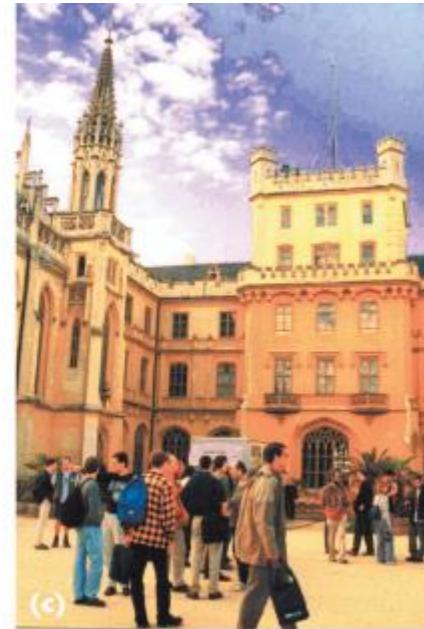
Result image



Reference image



Target image



Result image

# Color transfer

- Target image divided into RGB channels
- Reference image divided into RGB channels too.
- Use histogram equalization to let the histogram of target image be similar to reference image.

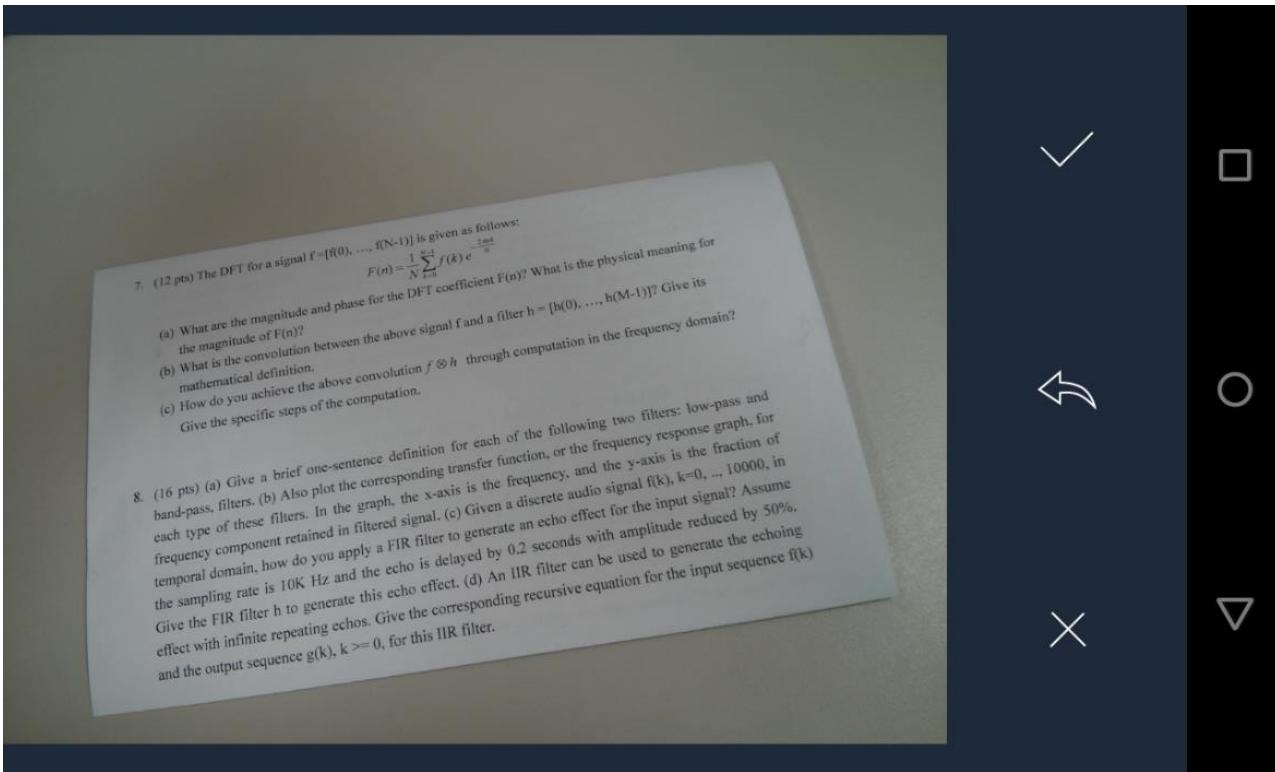
# Special effects camera app

An interactive app or application with user interface  
for live demonstration is necessary

Seek existing special-effect camera for inspiration

Realize your creativity

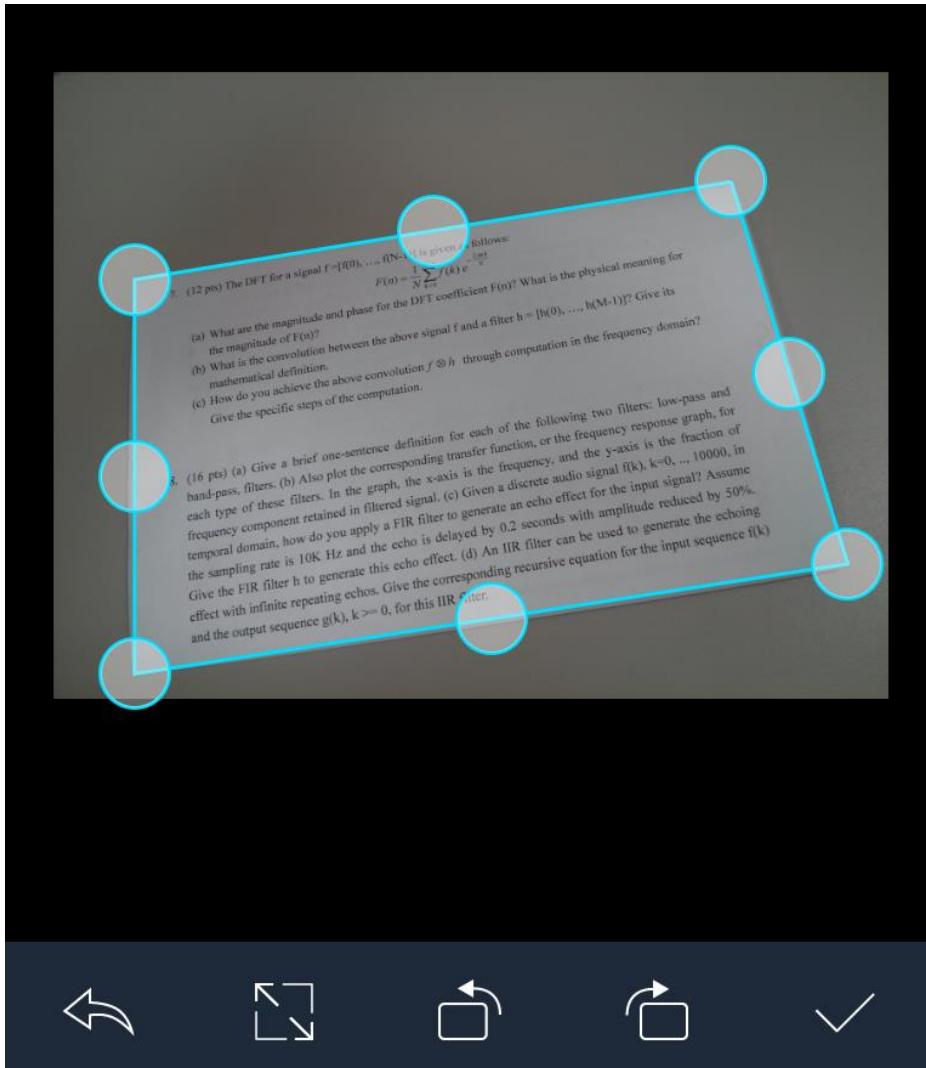
# Rectangle rectification for document capture



Take a photo of a document



# Rectangle rectification for document capture



Select 4 corners  
of this document



# Rectangle rectification for document capture

正在增強圖像...

7. (12 pts) The DFT for a signal  $f = [f(0), \dots, f(N-1)]$  is given as follows:

$$F(n) = \frac{1}{N} \sum_{k=0}^{N-1} f(k) e^{-\frac{j2\pi k n}{N}}$$

(a) What are the magnitude and phase for the DFT coefficient  $F(n)$ ? What is the physical meaning for the magnitude of  $F(n)$ ?  
(b) What is the convolution between the above signal  $f$  and a filter  $h = [h(0), \dots, h(M-1)]$ ? Give its mathematical definition.  
(c) How do you achieve the above convolution  $f \otimes h$  through computation in the frequency domain? Give the specific steps of the computation.

8. (16 pts) (a) Give a brief one-sentence definition for each of the following two filters: low-pass and band-pass, filters. (b) Also plot the corresponding transfer function, or the frequency response graph, for each type of these filters. In the graph, the x-axis is the frequency, and the y-axis is the fraction of frequency component retained in filtered signal. (c) Given a discrete audio signal  $f(k)$ ,  $k=0, \dots, 10000$ , in temporal domain, how do you apply a FIR filter to generate an echo effect for the input signal? Assume the sampling rate is 10K Hz and the echo is delayed by 0.2 seconds with amplitude reduced by 50%. Give the FIR filter  $h$  to generate this echo effect. (d) An IIR filter can be used to generate the echoing effect with infinite repeating echos. Give the corresponding recursive equation for the input sequence  $f(k)$  and the output sequence  $g(k)$ ,  $k \geq 0$ , for this IIR filter.

Apply Homography  
for rectification



Powered by  
CamScanner



# Rectangle rectification for document capture

7. (12 pts) The DFT for a signal  $f = [f(0), \dots, f(N-1)]$  is given as follows:

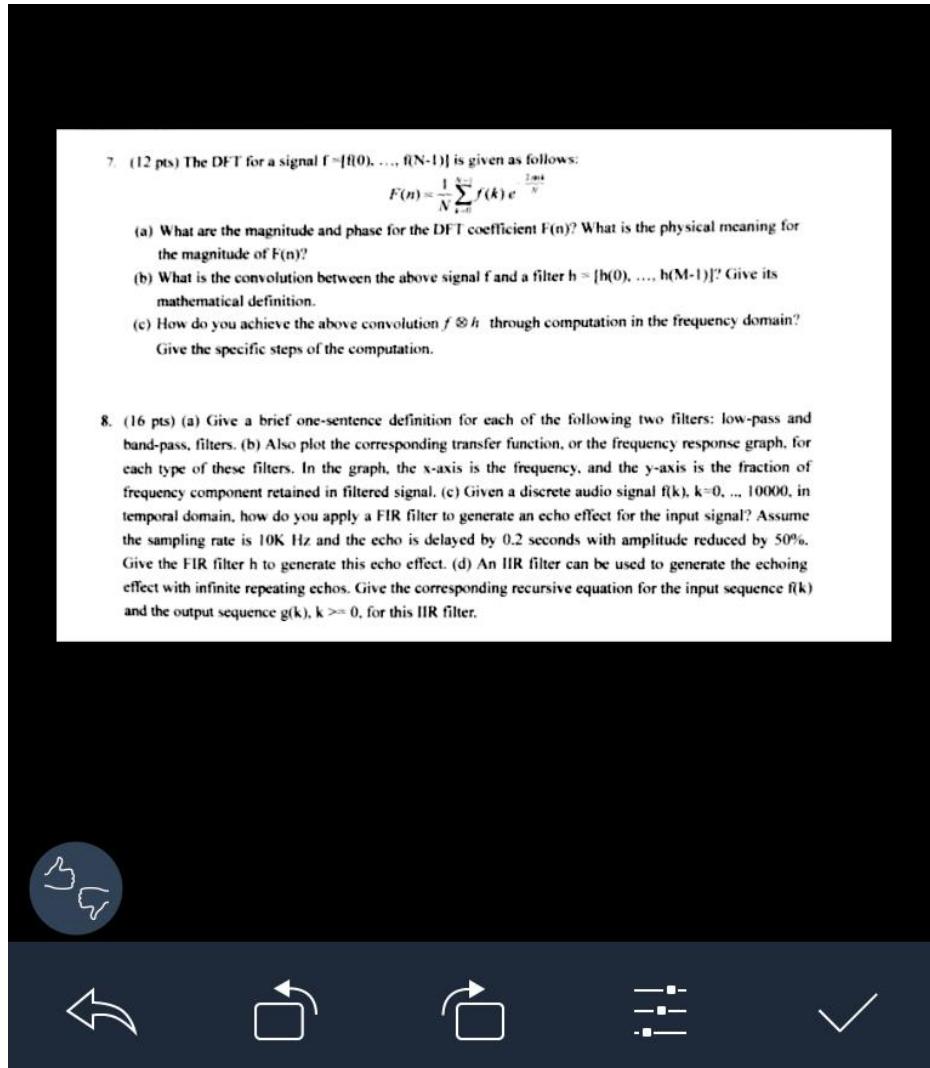
$$F(n) = \frac{1}{N} \sum_{k=0}^{N-1} f(k) e^{-\frac{j2\pi k n}{N}}$$

(a) What are the magnitude and phase for the DFT coefficient  $F(n)$ ? What is the physical meaning for the magnitude of  $F(n)$ ?

(b) What is the convolution between the above signal  $f$  and a filter  $h = [h(0), \dots, h(M-1)]$ ? Give its mathematical definition.

(c) How do you achieve the above convolution  $f \otimes h$  through computation in the frequency domain? Give the specific steps of the computation.

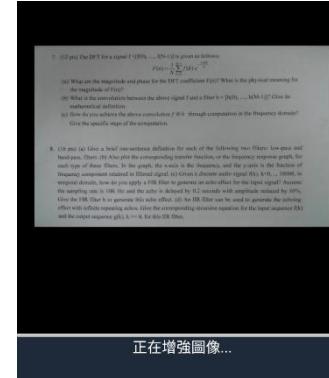
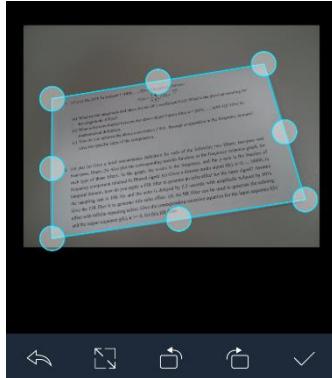
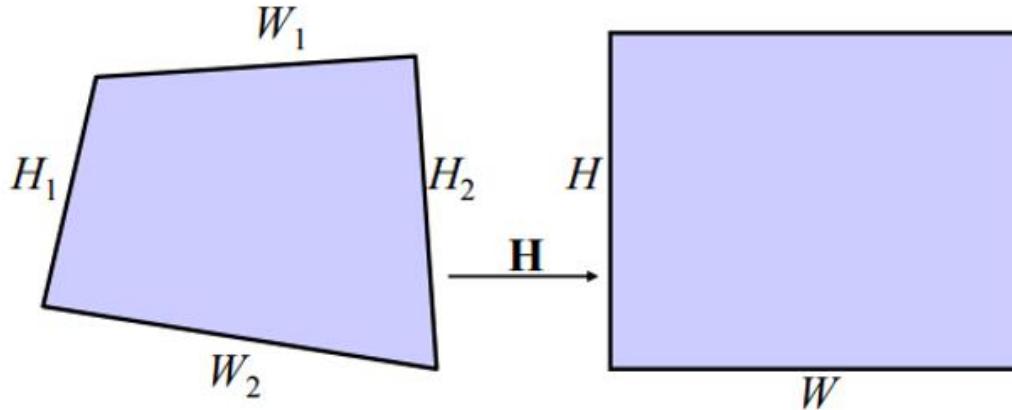
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Reconstructed  
document



# Homography



Left : original shape ; Right : rectified shape



# Homography example

$$\text{Left point} = H * \text{Right point}$$

Left points

Right points

$H =$

$$\begin{array}{ccc} 0.0084 & 0.0009 & 0.6518 \\ 0.0047 & 0.0052 & -0.7583 \\ 0.0000 & 0.0000 & 0.0042 \end{array}$$



Backward warping  
(bilinear interpolation)



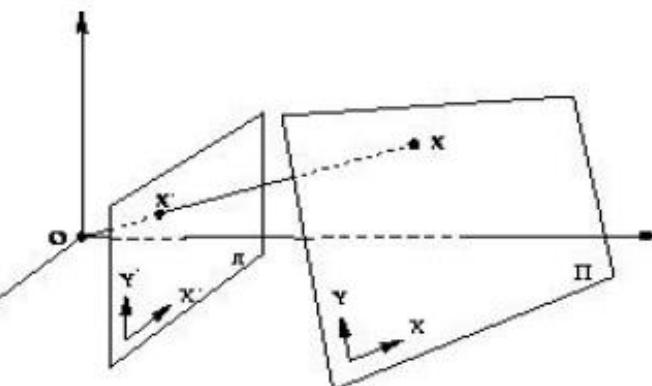
input



output



# Homography

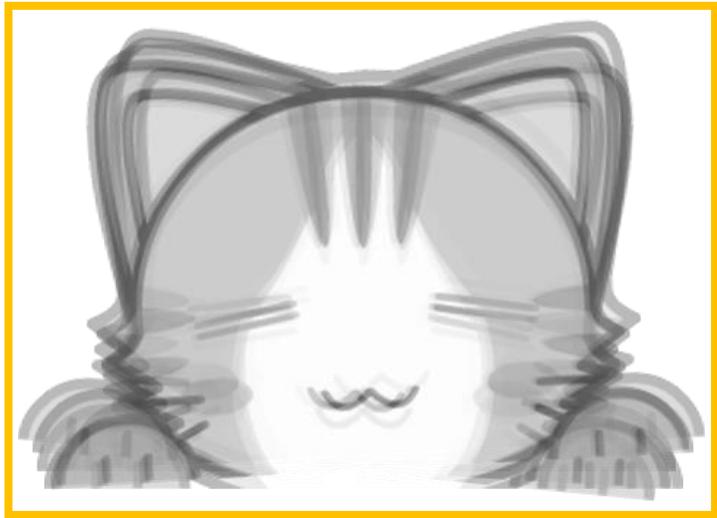
$$\begin{pmatrix} x'_1 \\ x'_2 \\ x'_3 \end{pmatrix} = \begin{bmatrix} h_{11} & h_{12} & h_{13} \\ h_{21} & h_{22} & h_{23} \\ h_{31} & h_{32} & h_{33} \end{bmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}$$


or  $\mathbf{x}' = \mathbf{H}\mathbf{x}$ , where  $\mathbf{H}$  is a  $3 \times 3$  non-singular homogeneous matrix.

- This is the most general transformation between the world and image plane under imaging by a perspective camera.
- It is often only the  $3 \times 3$  **form** of the matrix that is important in establishing properties of this transformation.
- A projective transformation is also called a “homography” and a “collineation”.
- $\mathbf{H}$  has 8 degrees of freedom.



# Video Stabilization



Shaky Video



Stabilized Video

# Video Stabilization

Why your video is shaky ?

Your camera is shaky

So we need to detect how camera shakes?

No, just how video shakes

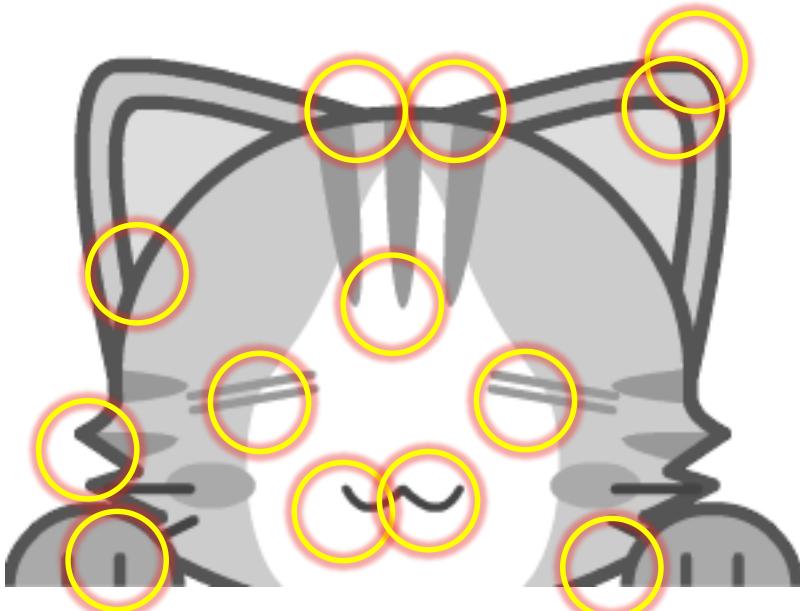
And compensate it

# Video Stabilization



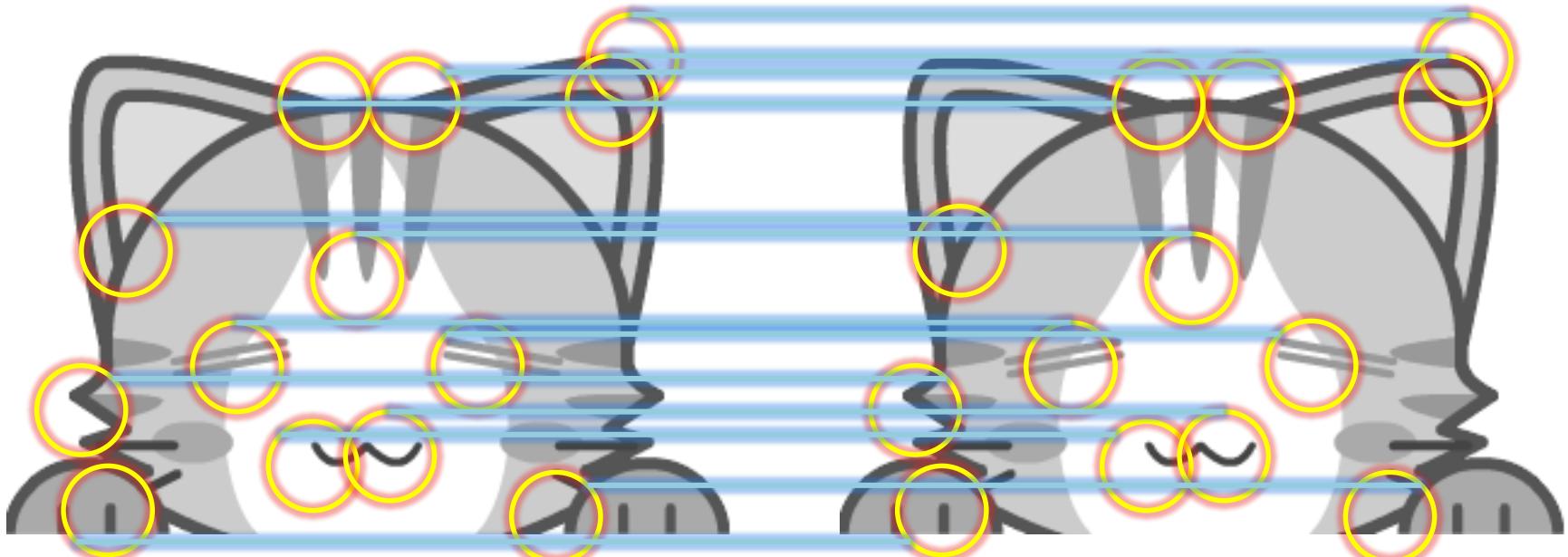
# Video Stabilization

## Local feature descriptor



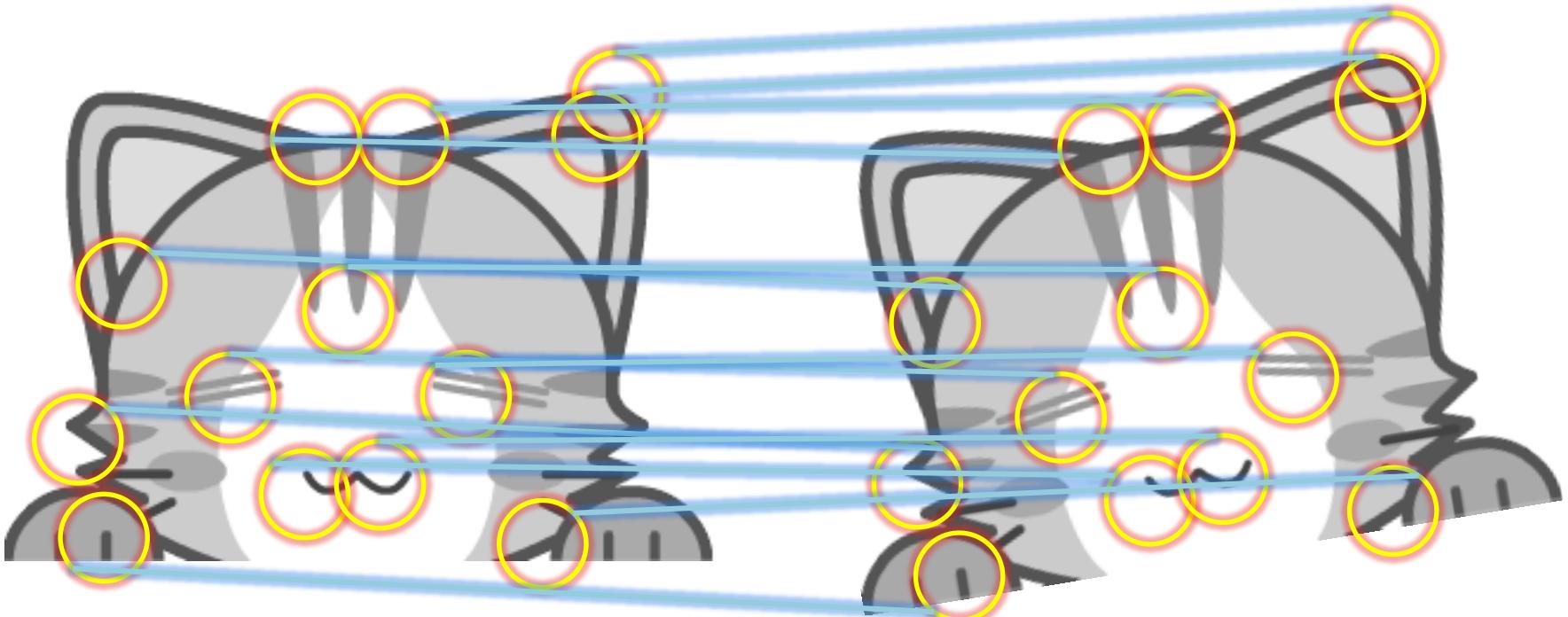
# Video Stabilization

Local feature descriptor  
Feature correspondence



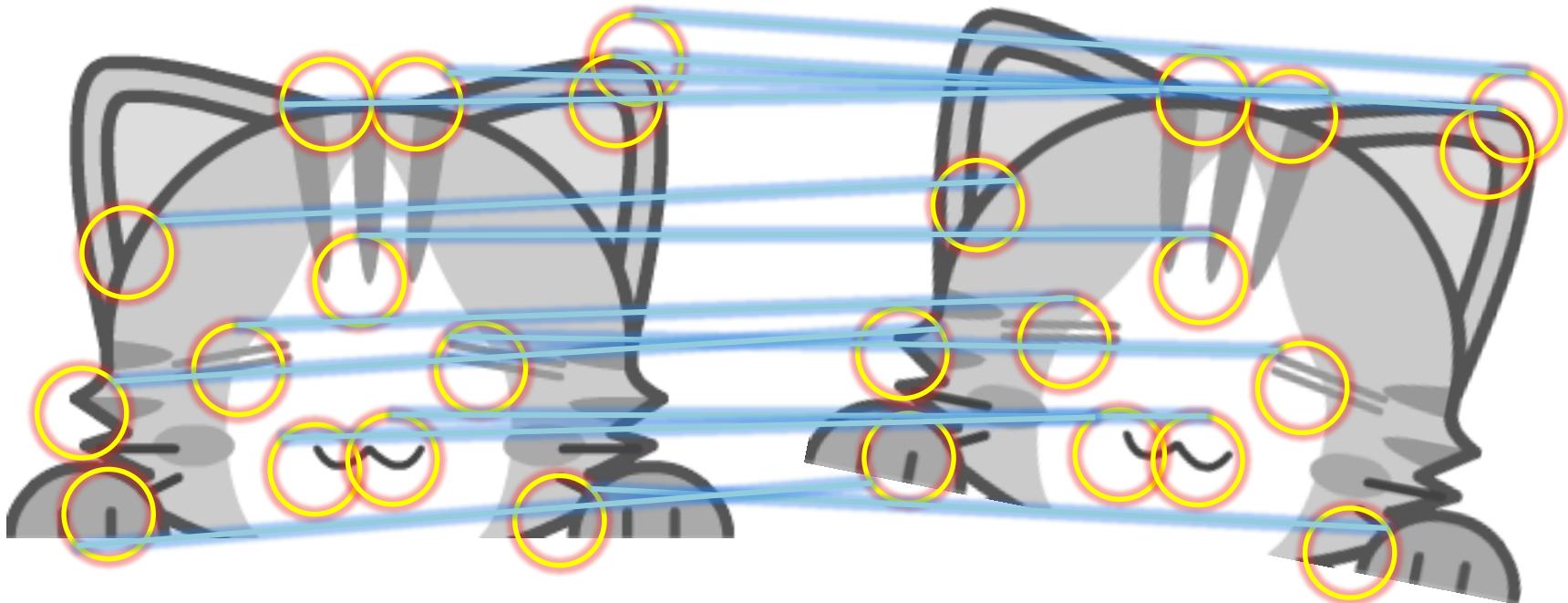
# Video Stabilization

Local feature descriptor  
Feature correspondence



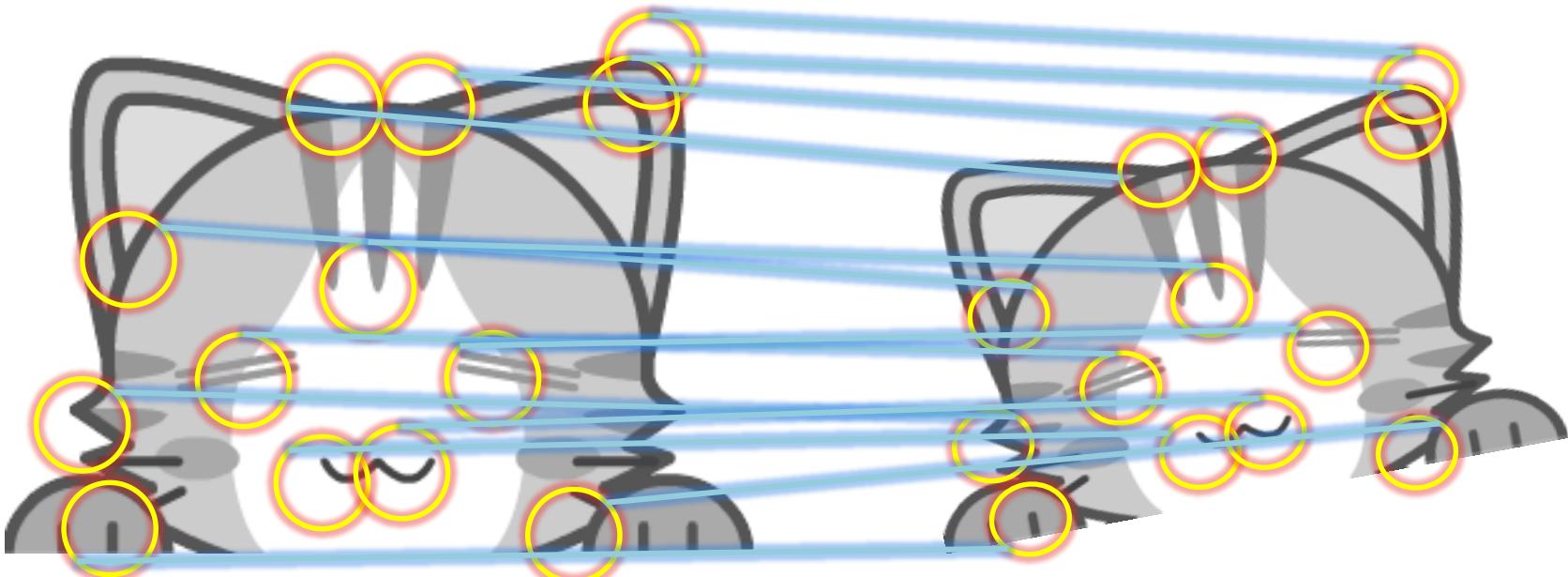
# Video Stabilization

Local feature descriptor  
Feature correspondence



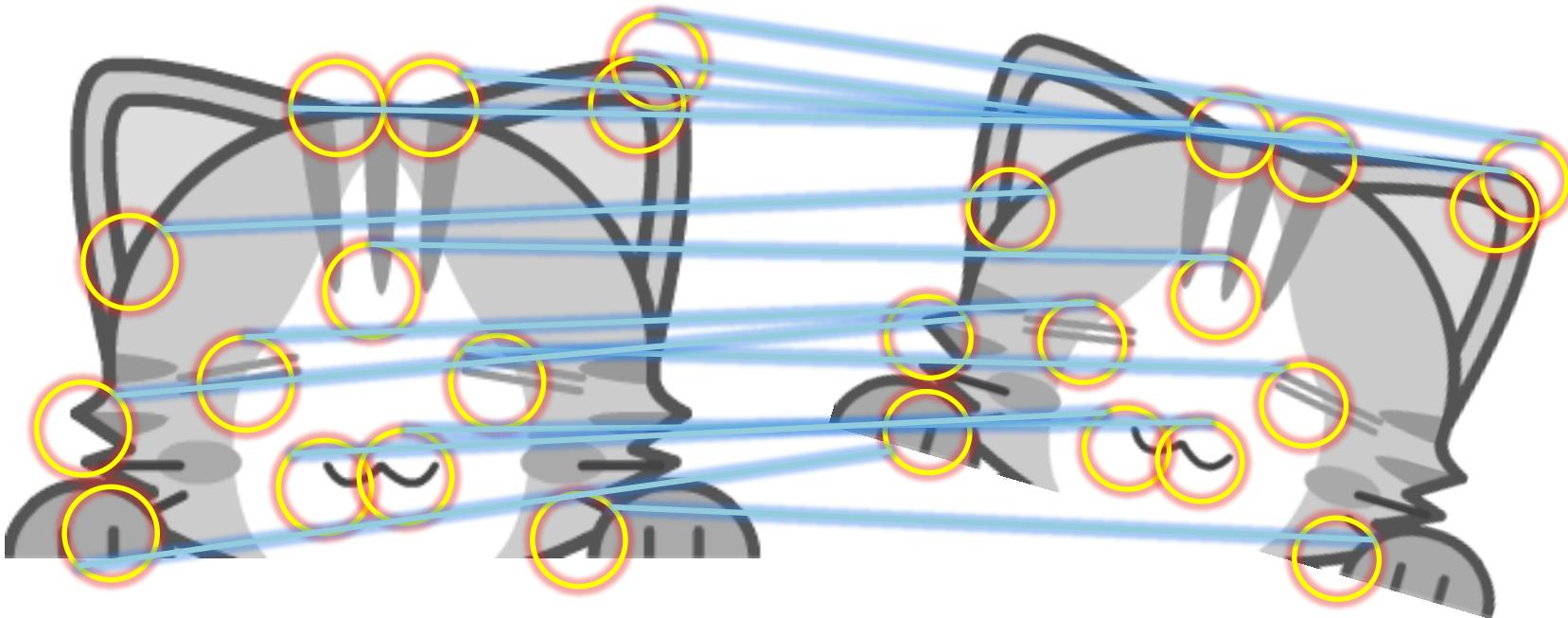
# Video Stabilization

Local feature descriptor  
Feature correspondence



# Video Stabilization

Local feature descriptor  
Feature correspondence



# Video Stabilization

Extract Local features

Match Features

Eliminate Outliers

Compute Homography

Warp Frames

MATLAB tutorial

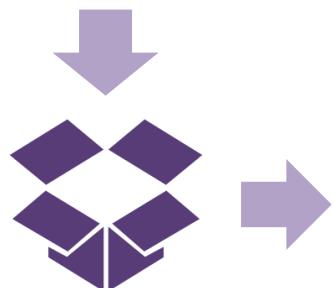
# Content-based Image Retrieval

Machine Learning

## Tag Based Image Retrieval

## Content Based Image Retrieval

“Angry Birds”



Retrieve with tags



Angry Birds



Angry Birds Rio



Angry Birds Space

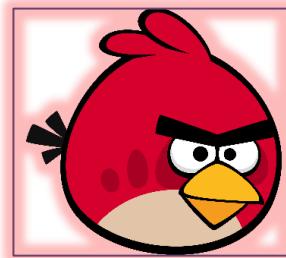


# Content-based Image Retrieval

Machine Learning

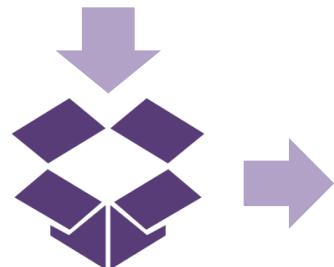
## Tag Based Image Retrieval

## Content Based Image Retrieval



Similar

Less  
Similar



Retrieve with features



# Content-based Image Retrieval

Machine Learning

What is “similarity” ?

Similarity between images

is

Similarity between features

# Content-based Image Retrieval

Machine Learning

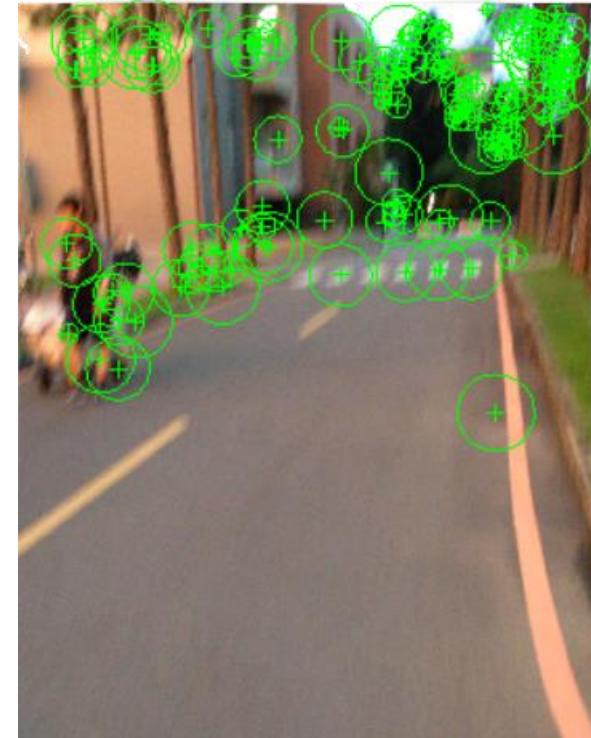
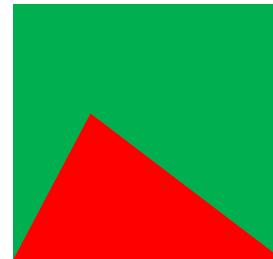
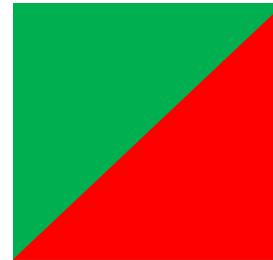
Each Image  
in Dataset



Detect Key-points  
(in gray scale)

Position

Scale



# Content-based Image Retrieval

Machine Learning

Position    Scale    Rotation

Describe these Key-points  
( For ex : using SIFT )



For example : SIFT

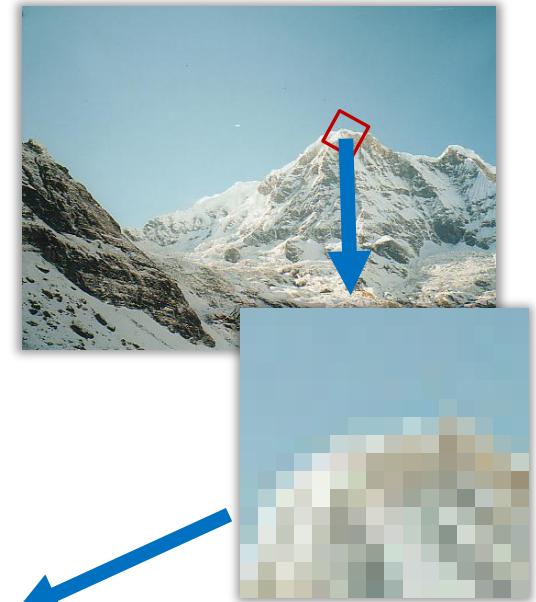
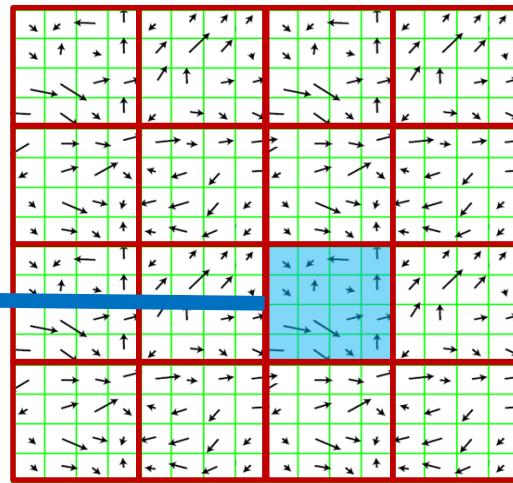
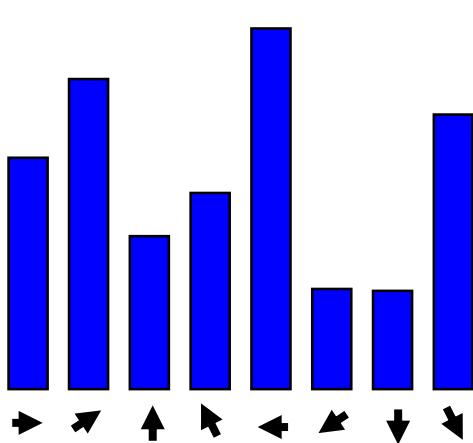
An SIFT feature is  
a normalized 128-D floating-point vector

# Content-based Image Retrieval

Machine Learning

Position    Scale    Rotation

Describe these Key-points  
( For ex : using SIFT )

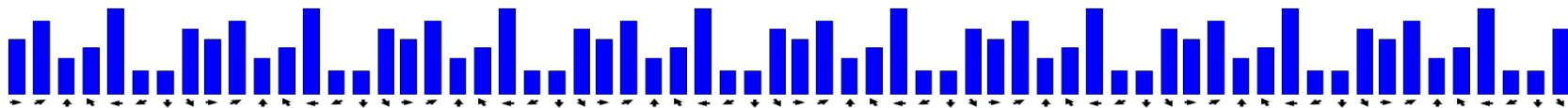


# Content-based Image Retrieval

Machine Learning

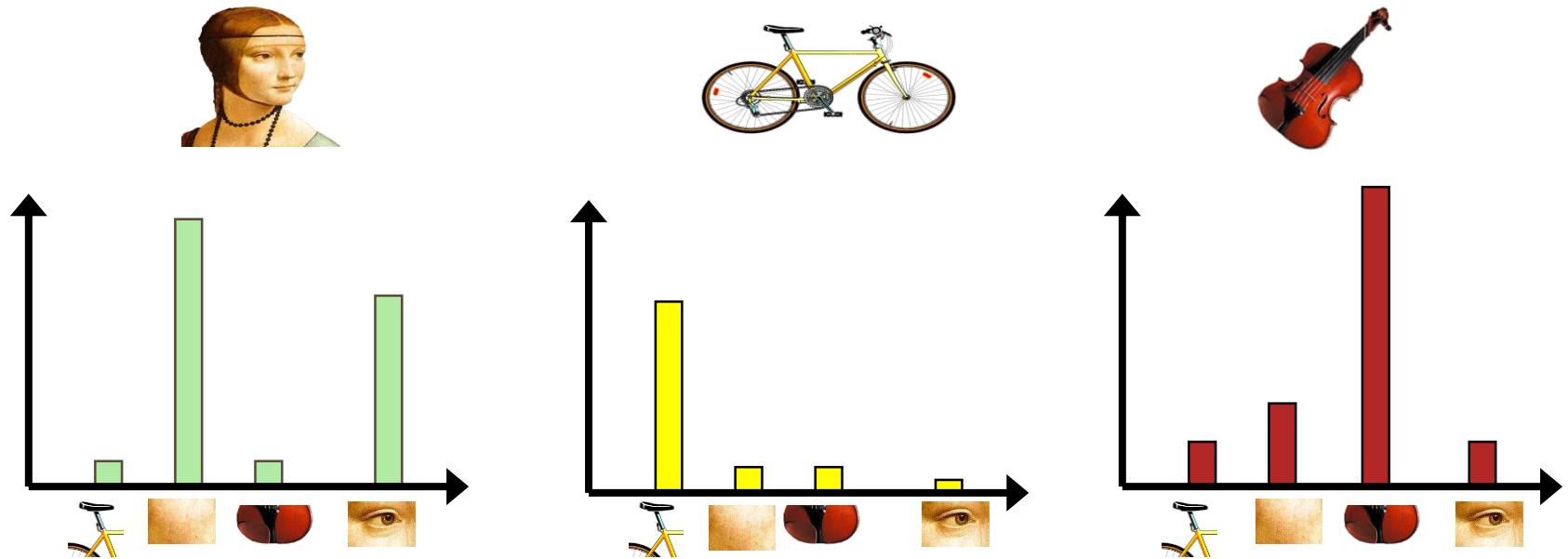
Position    Scale    Rotation

Describe these Key-points  
( For ex : using SIFT )



An SIFT feature is  
a normalized 128-D floating-point vector

# Bag of (visual) words

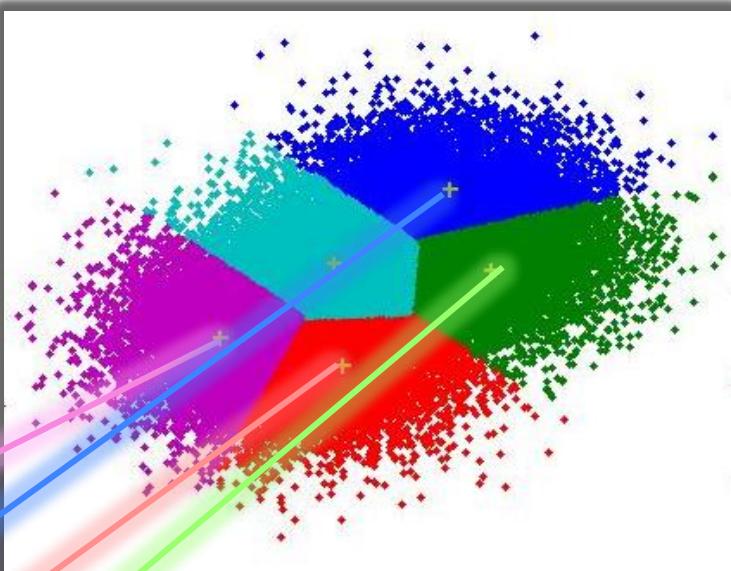
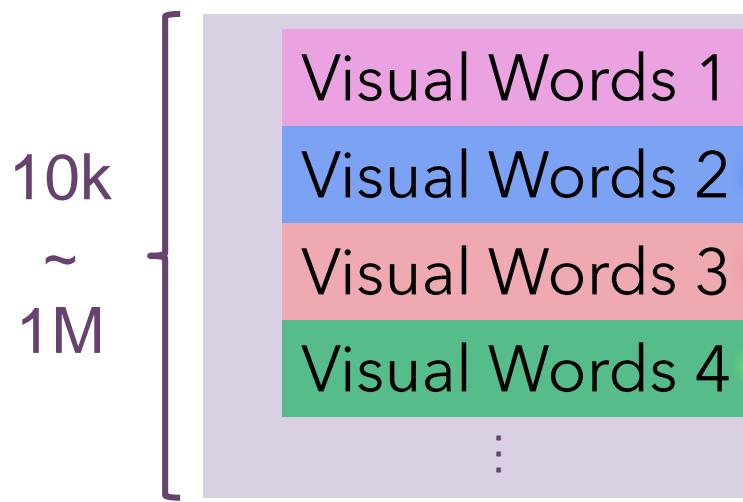


# Content-based Image Retrieval

Machine Learning

128-D Vector  $\equiv$  point in a 128-D space

Quantization  
( If using Bag of Words Model )



Dictionary



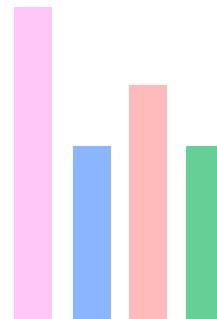
# Content-based Image Retrieval

Machine Learning

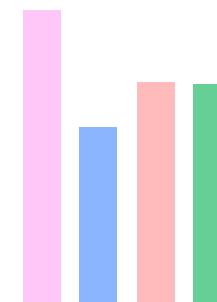
Compare and Compute Similarity

Euclidean Distance  
Taxicab Distance  
Hamming Distance  
Cosine Similarity

...



V	V	V	V
W	W	W	W
1	2	3	4



V	V	V	V
W	W	W	W
1	2	3	4

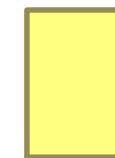
# Content-based Image Retrieval

Machine Learning

Sort Similarity  
Descending



query



# Content-based Image Retrieval

## Datasets

- INRIA Holidays
- Flickr 11k
- Caltech 101
- Caltech 256
- Bing Grand Challenge 2013

# Image Classification

- How to do “Classification” ?
  1. Preprocessing
  2. Feature extraction
  3. Modeling
    - Description of each class in mathematical form
  4. Classification
    - The classifier divides the feature space into class regions

# Image Classification

- Preprocessing

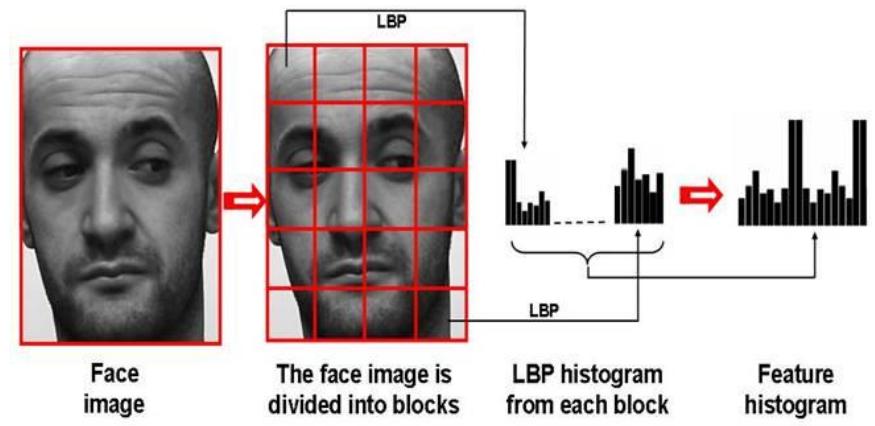
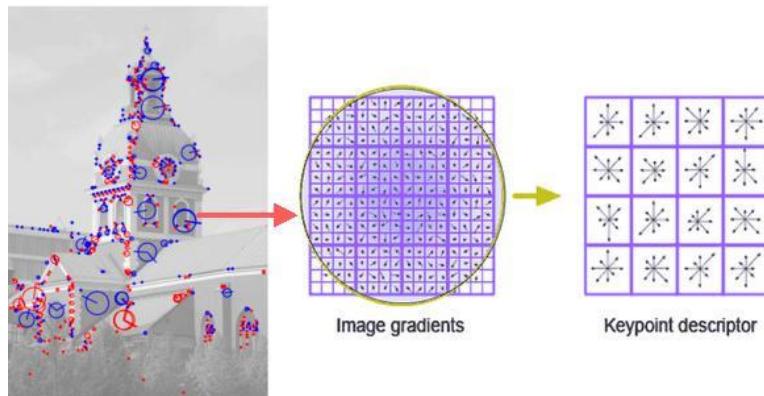
Cluster: K-means, Mean-shift...



Segmentation: Snake, Graph cut...

- Feature extraction

Feature: color intensity, SIFT, LBP...



# Image Classification

- Feature
  - A vector to describe a sample
  - e.g. Apply BoW to get a 10000-D feature
  - $\equiv$  a point in 10000-D space, called feature space
- Label
  - A scalar, indicate the class to which a sample belongs
  - e.g. for a 10-class classification problem, labels: 0, 1, ..., 9

# Image Classification

- Represent a sample as features and label
- Sample 1
  - Feature: 0, .3, 0, 0, 0, 0, 0, .1, 0, ..., .2, 0, 0
  - Labeled 4
- Sample 2
  - Feature: 0, 0, .1, .1, 0, 0, 0, 0, 0, ..., .2, 0, 0
  - Labeled 5
- ...

# Image Classification

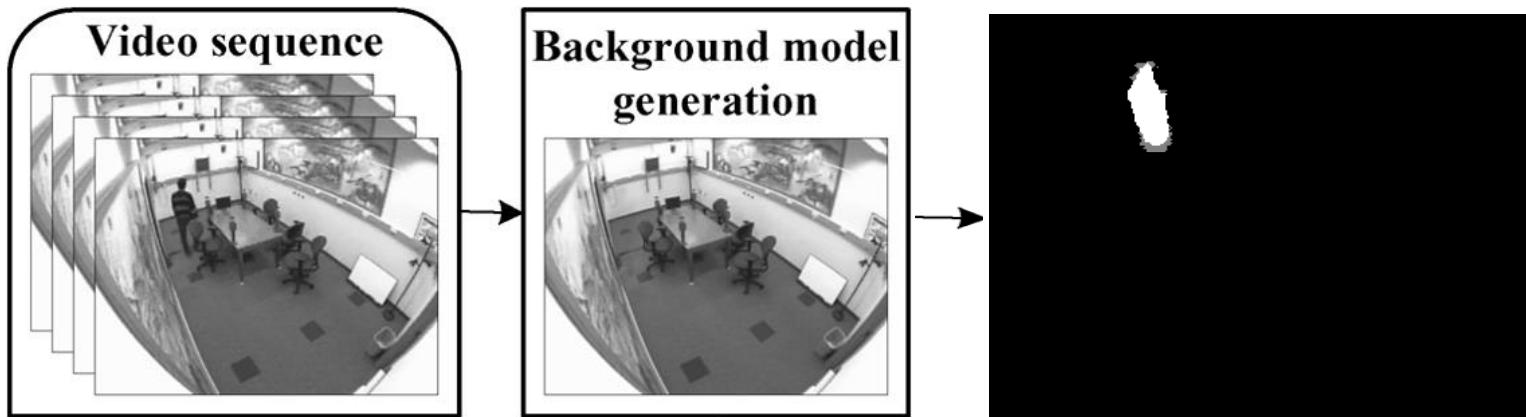
- Classification problem is
- Training on plenty of samples
  - with features and labels
- To get a model, such that
- The model can classify samples with features
  - into correct classes
- Not 100% correctness, but to maximize the
  - probability of correct classification

# Image Classification

- Classifiers
  - Probability based or Geometry based
- Bayesian classifier
- Support Vector Machine ([LibSVM](#))
- Kernel tricks (“plug-in” of classifiers)

# Background Subtraction

- **Background subtraction** is a technique in the fields of computer vision wherein an image's foreground is extracted for further processing
- BS is a widely used approach for detecting moving objects in videos from static cameras, e.g. surveillance tracking, human poses estimation...



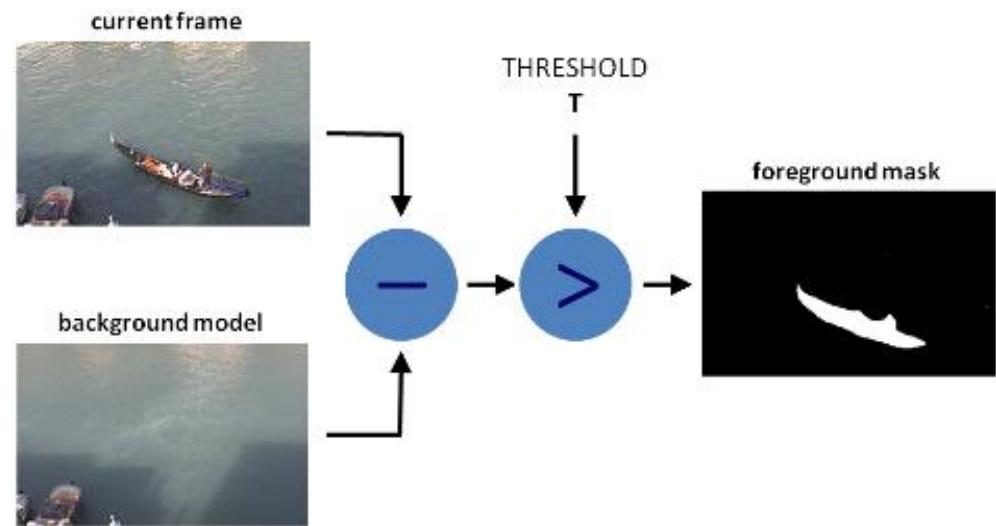
# Background subtraction

- The simplest way is that segment out the objects by using image subtraction technique
- For each pixels in  $I(t)$ , take the pixel value  $P[I(t)]$  and subtract it with the corresponding pixels at the same position on the background image  $P[B]$ .

$$P[F(t)] = P[I(t)] - P[B]$$

$$|P[F(t)] - P[F(t + 1)]|$$

> Threshold

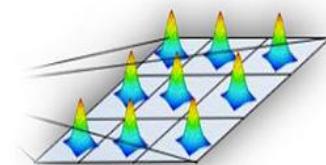


# Background subtraction

- Running Gaussian average adapt the background which changes over time (e.g. illumination changes or non-static background objects)
- mean  $\mu_t$  and variance  $\sigma_t$  at time t  
 $I_t$  is the value of the pixel's intensity at time t  
the parameter k is a free threshold

$$\frac{|(I_t - \mu_t)|}{\sigma_t} > k \rightarrow \text{Foreground}$$

$$\frac{|(I_t - \mu_t)|}{\sigma_t} \leq k \rightarrow \text{Background}$$



Distribucions gaussianes dels pixels

# Background subtraction - Example



# Background subtraction

- Better approach: BackgroundSubtractorMOG

[http://opencv-python-tutorials.readthedocs.org/en/latest/py\\_tutorials/py\\_video/py\\_bg\\_subtraction/py\\_bg\\_subtraction.html](http://opencv-python-tutorials.readthedocs.org/en/latest/py_tutorials/py_video/py_bg_subtraction/py_bg_subtraction.html)



- Related paper & book

<https://sites.google.com/site/backgroundsubtraction/overview>

- BS dataset

[http://bmc.iut-auvergne.com/?page\\_id=24](http://bmc.iut-auvergne.com/?page_id=24)



# To be continued

More topics to be reviewed in the next class