

# Bag-of-Words based Image Classification (with OpenCV-Python)

Heeseung Yun, Jaewoo Ahn



SEOUL NATIONAL UNIV.  
**VISION & LEARNING**

# 공지사항

- 질문은 비밀글로 올리시면 답변해드리지 않습니다
- 컴퓨터비전 1,2 실습은 아래의 3개로 구성되어 있습니다.  
숙제는 3번파일에 들어있는 미완성되어 있는 부분을 완성하여 **9/24 23:59**까지 **ETL**에 제출하면 됩니다.
  - 1\_getting\_started.ipynb (수업시간에 같이)
  - 2\_feature\_matching.ipynb (문제 없음)
  - 3\_classifictaion.ipynb (**ETL에 제출**)

# 실습코드 받는법

[https://github.com/bckim92/iab\\_practice\\_example](https://github.com/bckim92/iab_practice_example)

```
$ git clone https://github.com/bckim92/iab_practice_example.git
$ cd iab_practice_example/
$ pip install -r requirements.txt
$ jupyter notebook
```

The screenshot shows the GitHub repository page for `bckim92 / iab_practice_example`. The repository is private and has 1 commit, 1 branch, 0 releases, 1 contributor, and 303 KB of code. The main branch is `master`. The repository contains the following files:

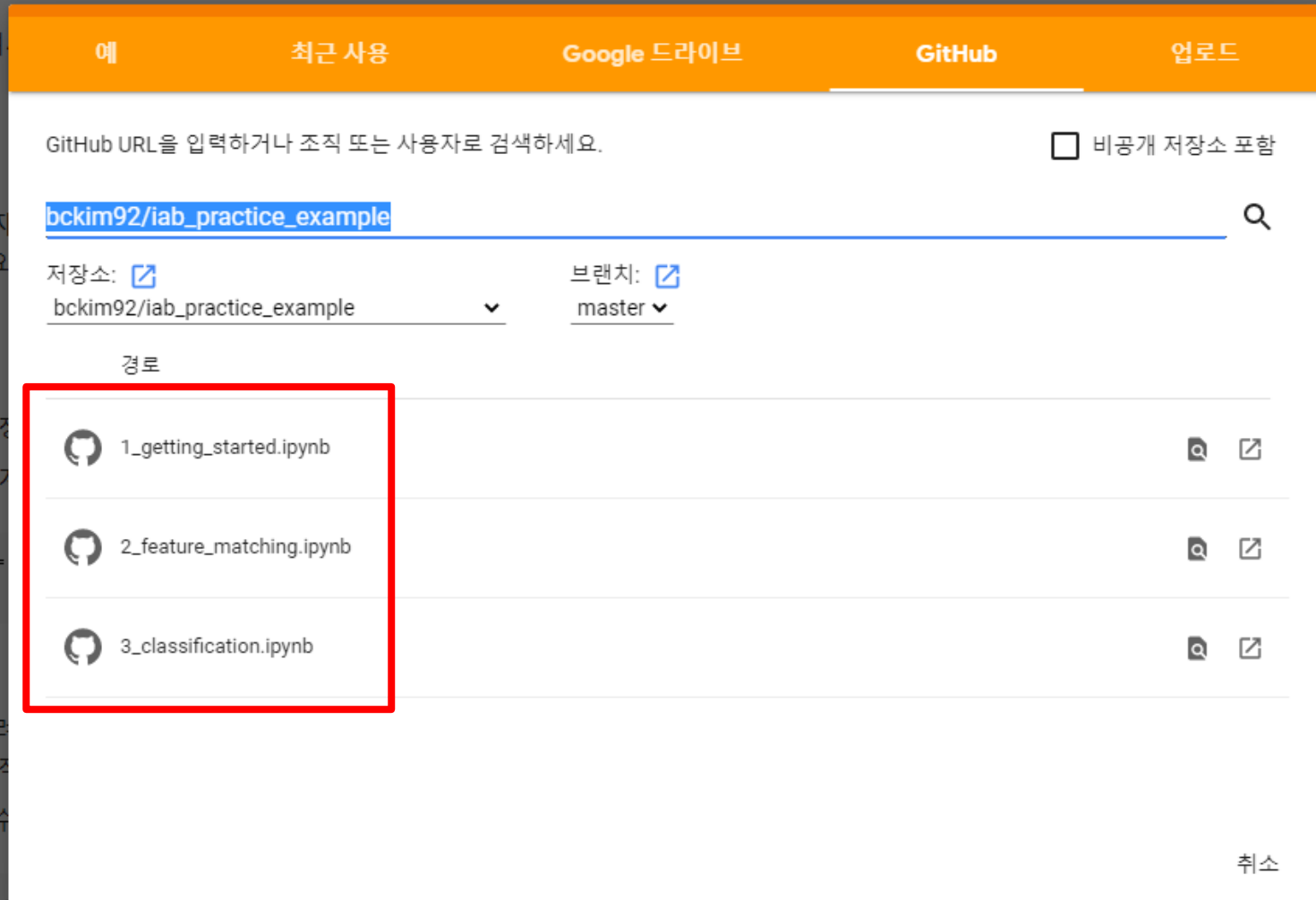
File	Commit	Size	Time
<code>images</code>	first commit		2 days ago
<code>1_getting_started.ipynb</code>	first commit	7.06 KB	2 days ago
<code>2_feature_matching.ipynb</code>	first commit	8.44 KB	2 days ago
<code>3_classification.ipynb</code>	first commit	15.29 KB	2 days ago
<code>requirements.txt</code>	first commit	166 B	2 days ago

At the bottom, there is a prompt to "Add a README" to help people understand the project.

# 실습환경 세팅 google colab

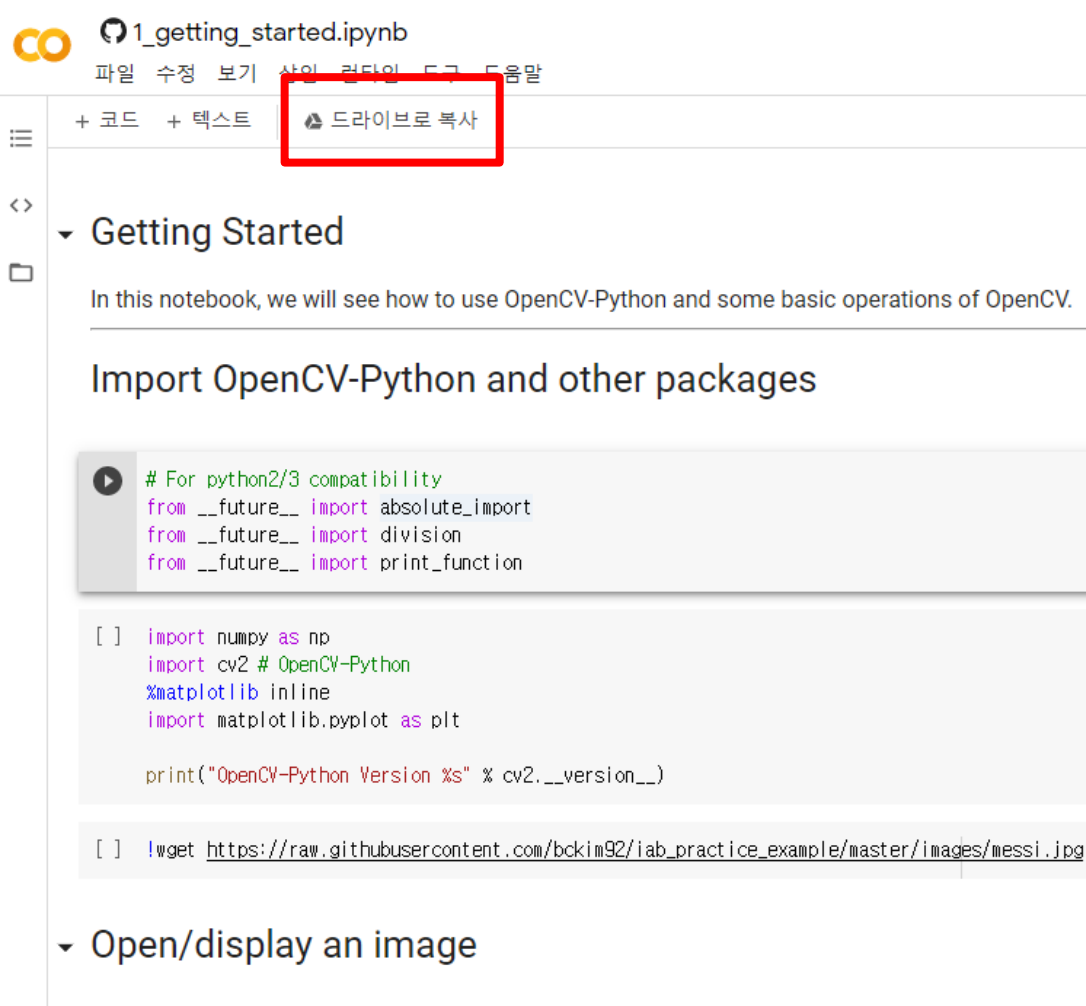
파일 -> 노트열기 -> github 탭 ->

bckim92/iab\_practice\_example 입력 -> 해당 ipynb file 클릭



# 실습환경 세팅 google colab

## 드라이브로 복사



1\_getting\_started.ipynb

파일 수정 보기 삽입 런타임 도구 도움말

+ 코드 + 텍스트 **드라이브로 복사**

### Getting Started

In this notebook, we will see how to use OpenCV-Python and some basic operations of OpenCV.

#### Import OpenCV-Python and other packages

```
# For python2/3 compatibility
from __future__ import absolute_import
from __future__ import division
from __future__ import print_function

[ ] import numpy as np
import cv2 # OpenCV-Python
%matplotlib inline
import matplotlib.pyplot as plt

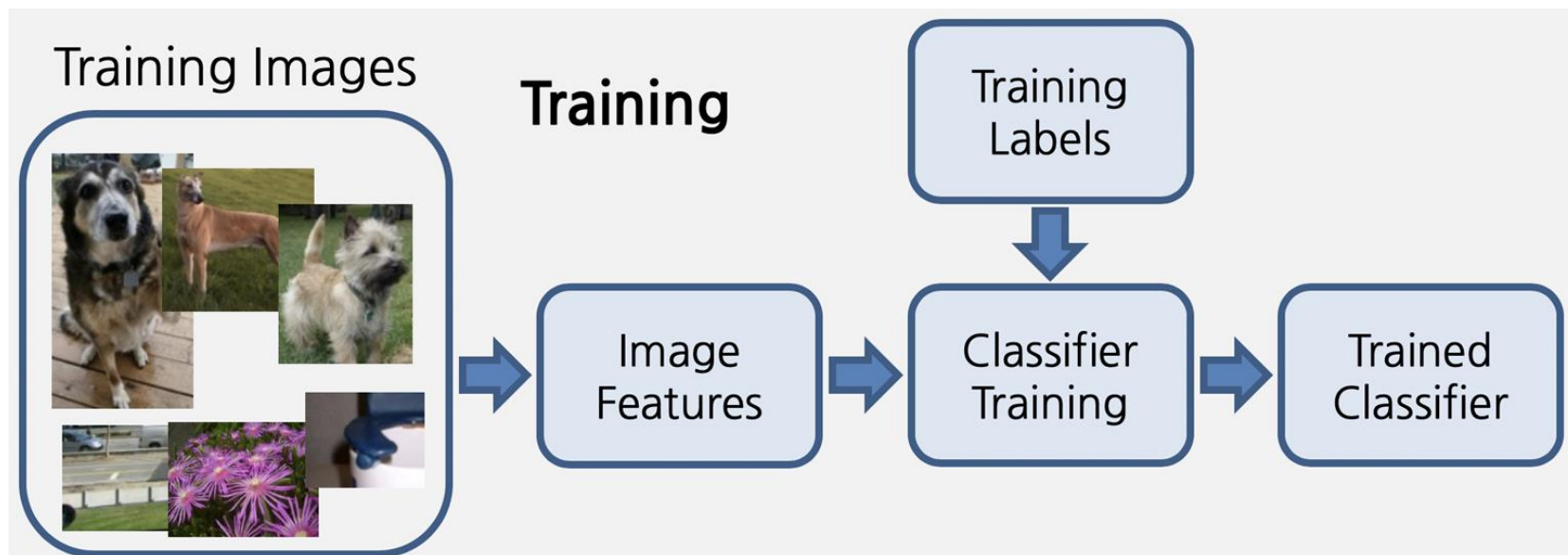
print("OpenCV-Python Version %s" % cv2.__version__)

[ ] !wget https://raw.githubusercontent.com/bckim92/iab_practice_example/master/images/messi.jpg
```

### Open/display an image

# Contents

- Backgrounds
  - OpenCV-Python
  - Image features (e.g. SIFT)
  - Classifier (e.g. SVM)
- Bag-of-words based image classification



# OpenCV-Python

Introduction

Image manipulation

Draw objects

# OpenCV-Python

- OpenCV
  - **Computer vision library** started from 1995(Intel)
  - Now supports a multitude of algorithms related to CV and ML (a little of)
- OpenCV-Python
  - OpenCV is basically written in C++
  - OpenCV-Python is a **Python wrapper** of OpenCV
- Prior knowledge of **Python** and **Numpy** is needed
  - A Quick guide to Python - [A Byte of Python](#)
  - [Numpy Quickstart Tutorial](#) / [Justin Johnson's Numpy Tutorial](#)





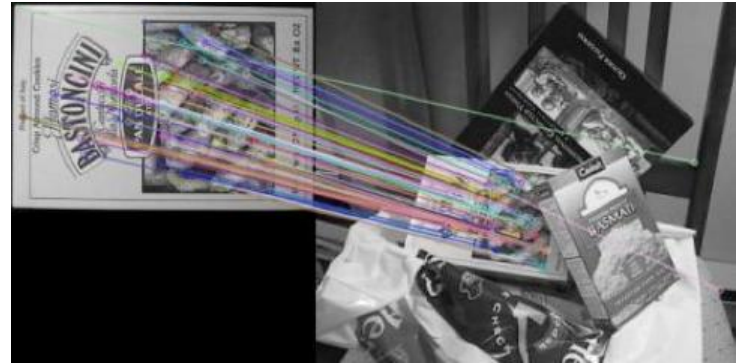
# OpenCV-Python

- Examples of algorithms with OpenCV

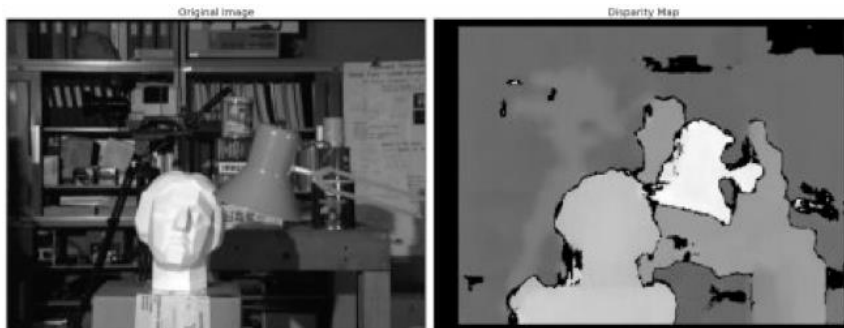
- Face Detection



- Feature extraction/matching



- Depth map for stereo images



- Image inpainting



# Using OpenCV-Python

- Import OpenCV-Python package “cv2”

```
import numpy as np
import cv2 # OpenCV-Python
%matplotlib inline
import matplotlib.pyplot as plt
```

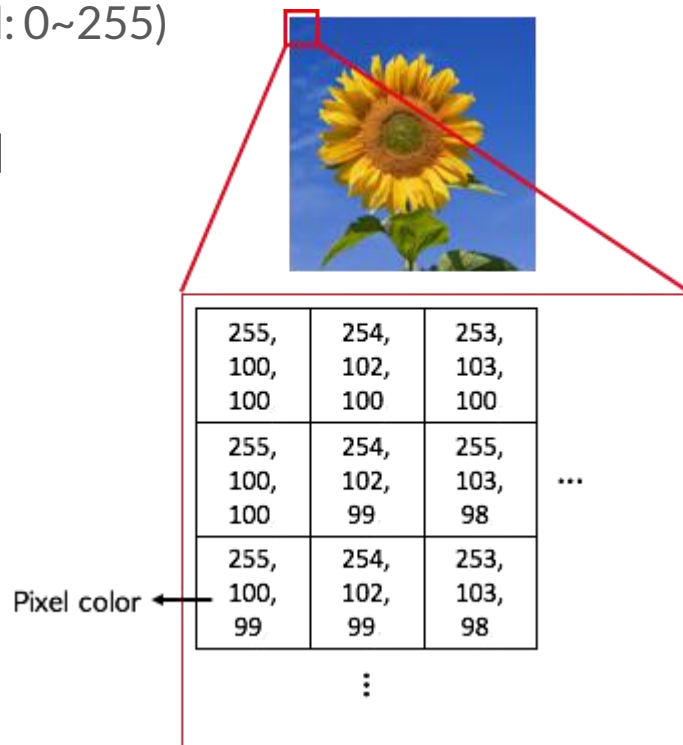
- **Numpy** arrays are data structure used in cv2
  - Converted from/to CvMat(OpenCV in C++) by OpenCV-Python
  - Also used in many python packages

# Open/Display an Image

- Open an image

```
img = cv2.imread('image.jpg', cv2.IMREAD_COLOR)
```

- The output is a Numpy array
  - 3D (H x W x C for color) / 2D (H x W for grayscale)
  - Top-left to bottom-right
  - Data type (dtype): np.uint8 (1-byte unsigned: 0~255)
- Flag specifies the way image should be read
  - cv2.IMREAD\_COLOR
  - cv2.IMREAD\_GRAYSCALE
  - cv2.IMREAD\_UNCHANGED

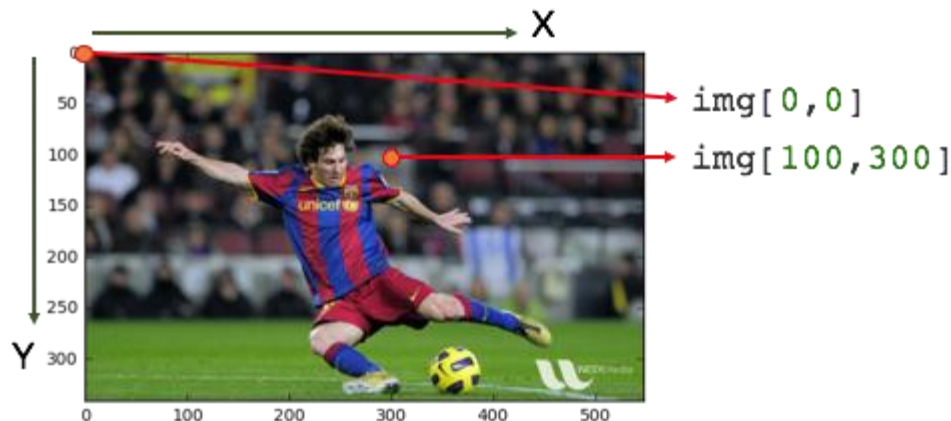


# Open/Display an Image

- Display an image using Matplotlib

```
# display an image using matplotlib  
plt.imshow(img) # => The output is wrong color!!  
plt.imshow(cv2.cvtColor(img, cv2.COLOR_BGR2RGB))
```

- `plt.imshow(img)` displays an (RGB, RGBA, grayscale) image
- OpenCV represents RGB images as Numpy arrays in **REVERSE** order (**BGR** not RGB)
- `cv2.cvtColor(img, conversion)` provides conversion among many colortypes



# Modify Pixels & ROI

- Pixel and ROI(Region of Interest) can be accessed by Numpy indexing
  - [row,column] ordering - same as matrix indexing

```
# Access a pixel value (BGR order)  
img[50, 235]
```

```
=> array([27, 25, 24], dtype=uint8)
```

```
# ROI is obtained using Numpy indexing  
ball = img[280:340, 330:390]  
img[273:333, 100:160] = ball
```



# Draw Objects

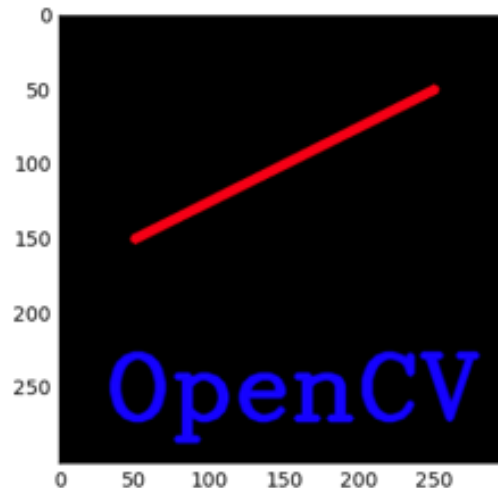
- Draw object (line, rectangle, circle, ellipse, polygon)
  - `cv2.line()`, `cv2.rectangle()`, `cv2.circle()`, `cv2.ellipse()`, `cv2.polyline()`
- Put some text
  - `cv2.putText()`
- Arguments
  - `cv2.function(image, {properties of object})`
  - **RGB** order in color (not BGR)
  - **X, Y** order in position (not row, column)

# Draw Objects

- Example

```
# cv2.line(image, startPoint, endPoint, rgb, thickness)
cv2.line(img, (50,150), (250,50), (255,0,0), 5)

# cv2.putText(image, text, bottomLeft, fontType, fontScale,
rgb, thickness, lineType)
font = cv2.FONT_HERSHEY_COMPLEX
cv2.putText(img, 'OpenCV', (30,270), font, 2, (0,0,255), 3,
cv2.LINE_AA)
```



Let's Check the Code

**1\_getting\_started.ipynb**



# Image Features

Understanding Features

Feature detection / description

SIFT

Feature Extraction in OpenCV-Python

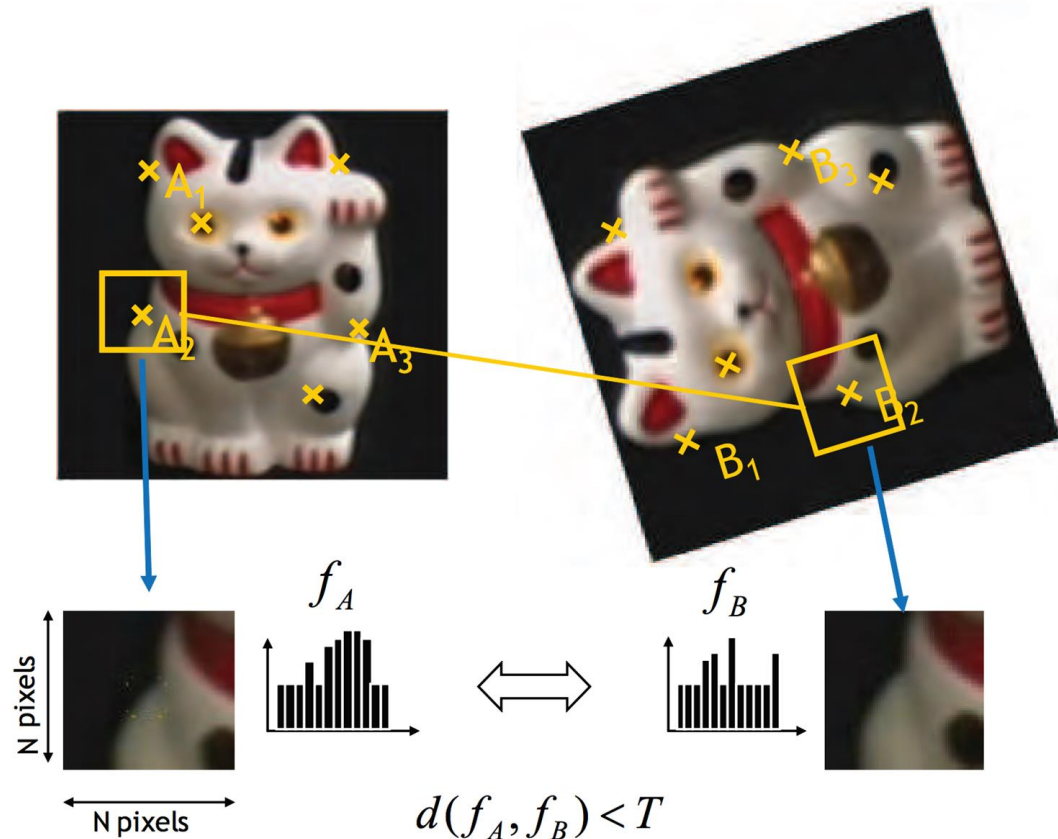
# Understanding Features

- Feature : A piece of information relevant for solving task.
  - Specific patterns which are unique, which can be easily tracked, which can be easily compared
  - Local visual feature: **Salient point** and its **representation**
- An example: Find patches A~F in the picture



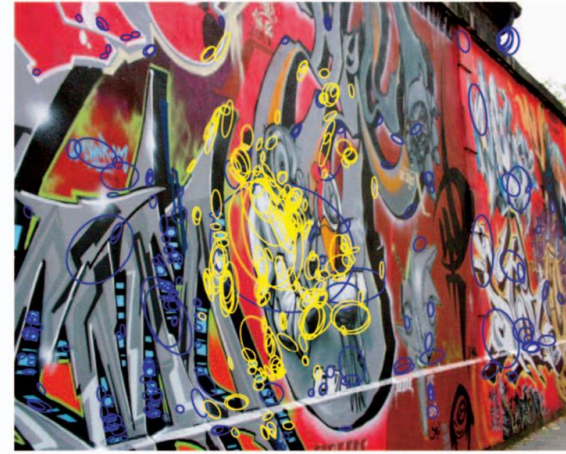
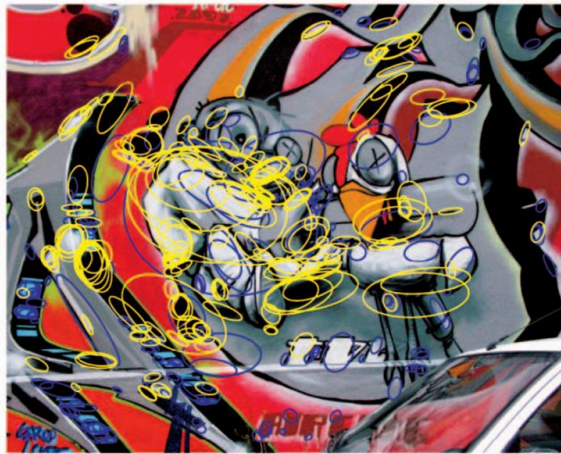
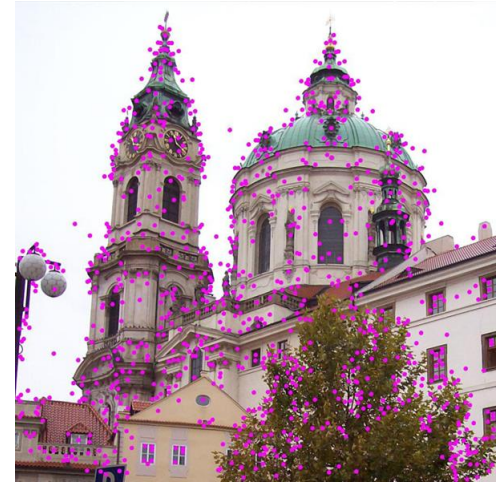
# Understanding Features

- Local Visual Features
  - Feature detection / Feature description
- Keypoint Matching
  - Find distinctive points
  - Define local regions around the points
  - Compute local descriptors from the region
  - Match local descriptors of two images



# Feature Detection / Description

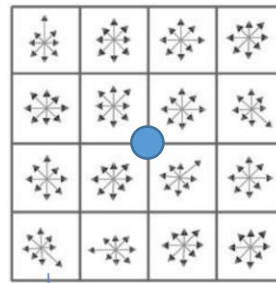
- Keypoint(Local Feature) Detection
  - Finding keypoints / interest points
  - Usually corners and blob centers
  - Harris corner detector, DoG, MSER
- Good Feature Detection
  - Repeatable: Robust to scaling / rotation / viewpoint change
  - Distinctive: Different features should look differently



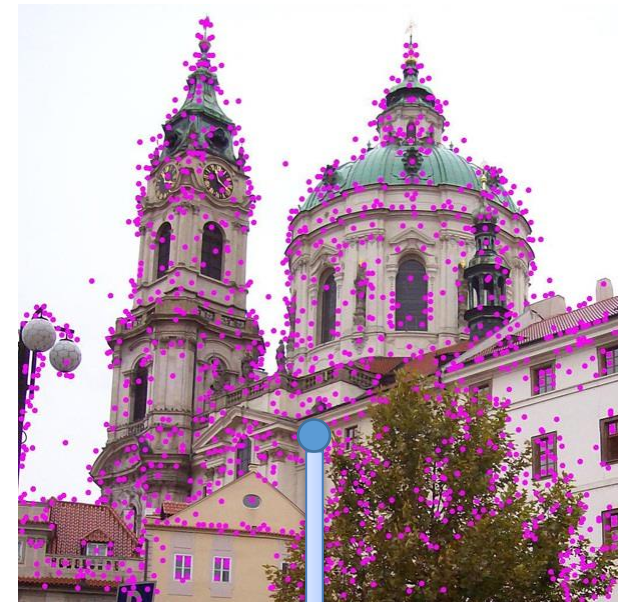


# Feature Detection / Description

- Feature Description
  - Represent keypoints or local regions around them as a vector
  - SIFT, SURF
- ex) SIFT descriptor – 128-D vector



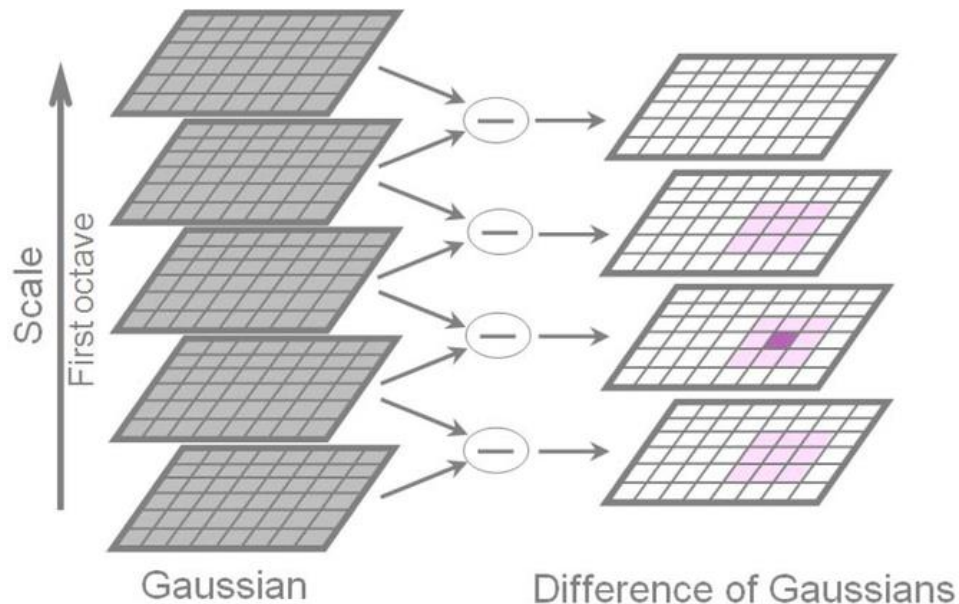
128 Dimension



(7.0, 0.0, 0.0, 0.0, 128.0,  
86.0, ..., 4.0 2.0 0.0)

# SIFT (Scale-Invariant Feature Transform)

- An algorithm to **detect** and **describe** local features in images
  - Feature detector + Feature descriptor
  - Published by David Lowe in 1999
- We will skip the details of SIFT in this class

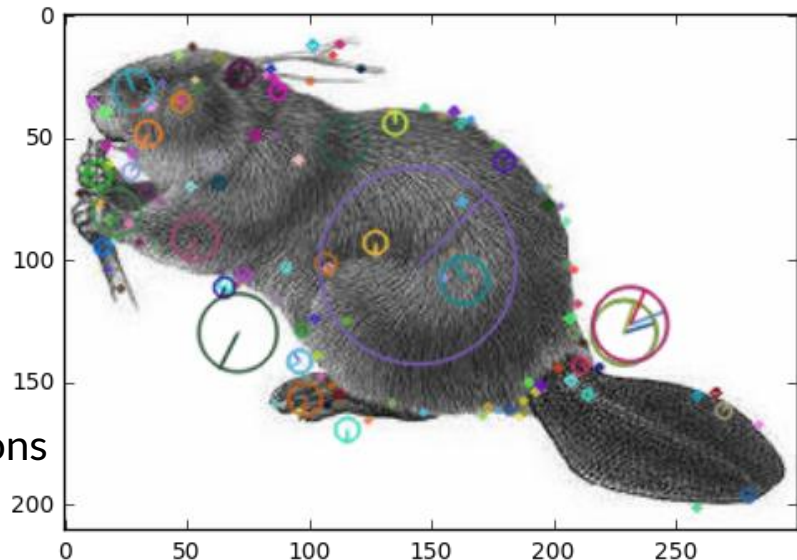


# Feature Extraction in OpenCV

- `Cv2.xfeatures2d` module
  - Sub-module for feature detectors / descriptors (opencv\_contrib package)
  - SIFT, SURF, BRIEF ...
- Module for extracting and computing SIFT

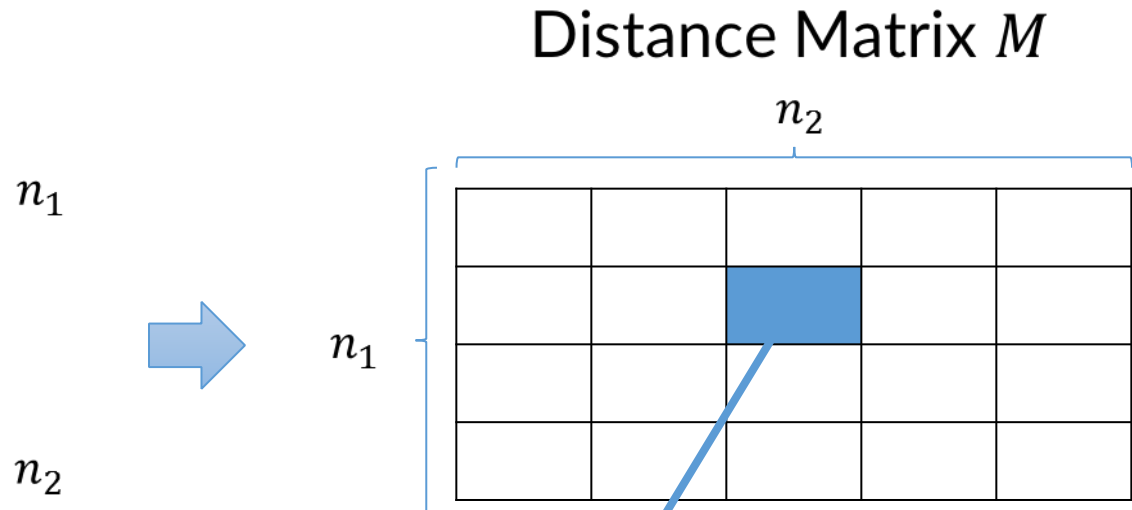
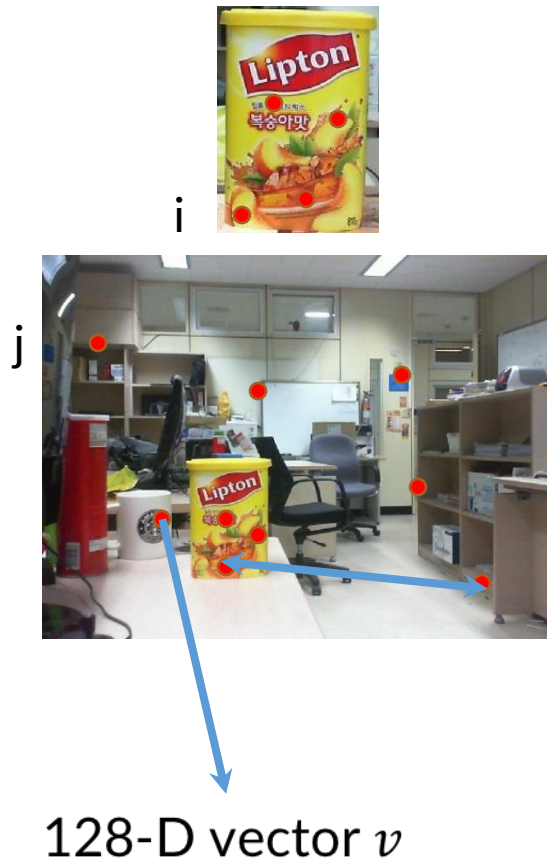
```
# SIFT feature detector/descriptor
sift = cv2.xfeatures2d.SIFT_create()
kp, des = sift.detectAndCompute(gray, kp)
```

- Output
  - **kp**: A list of  $N$  'cv2.KeyPoint'
    - position, scale, orientation
  - **des**: Descriptors ( $N \times 128$  Numpy array)
- Caveat
  - SIFT is excluded in recent OpenCV versions
  - Solution: use opencv-python==3.4.2.17



# Feature Matching

- Brute Force Matching



$$M_{i,j} = d(u_i, v_j)$$

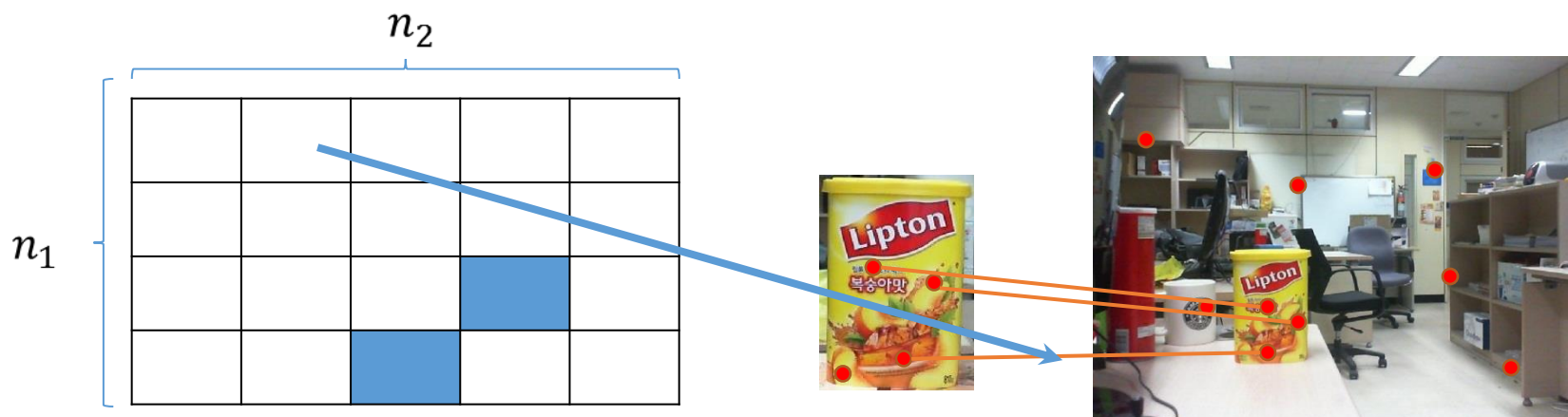
## Feature Distance

- L2-norm:  $\|u_i - v_j\|$
- Hellinger distance:  $\sqrt{1 - \sqrt{u_i v_j}}$



# Feature Matching

- Brute Force Matching



- Brute Force Matching in OpenCV-Python

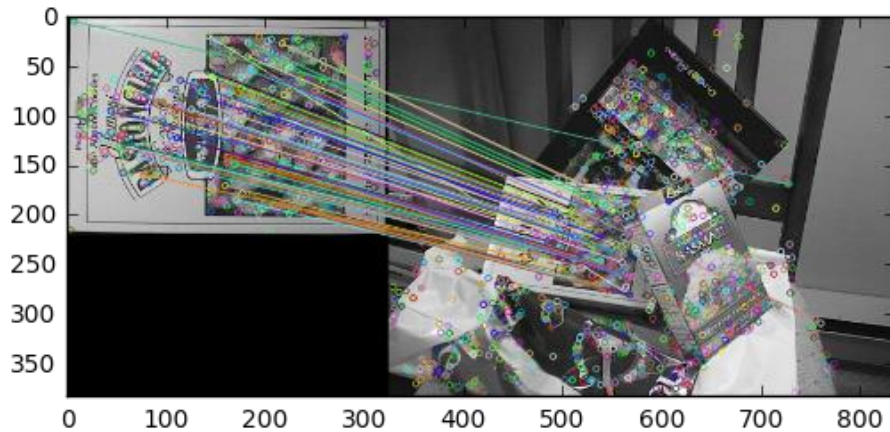
```
# BFMatcher(Brute Force Matcher) with default setting(L2 distance)
bf = cv2.BFMatcher(cv2.NORM_L2)
# Find closest 2 des2 points for each point in des1
matches = bf.knnMatch(des1, des2, k=2)
```

# Feature Matching

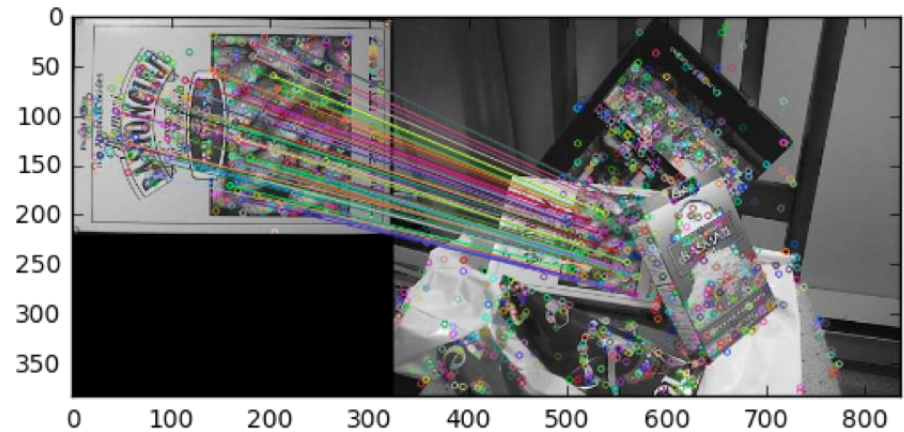
- Display Matches

```
# Display matches  
img_draw = cv2.drawMatches(img1, kp1, img2, kp2, matches, None)  
plt.imshow(cv2.cvtColor(img_draw, cv2.COLOR_BGR2RGB))
```

- Euclidean distance



- Hellinger distance



Let's Check the Code

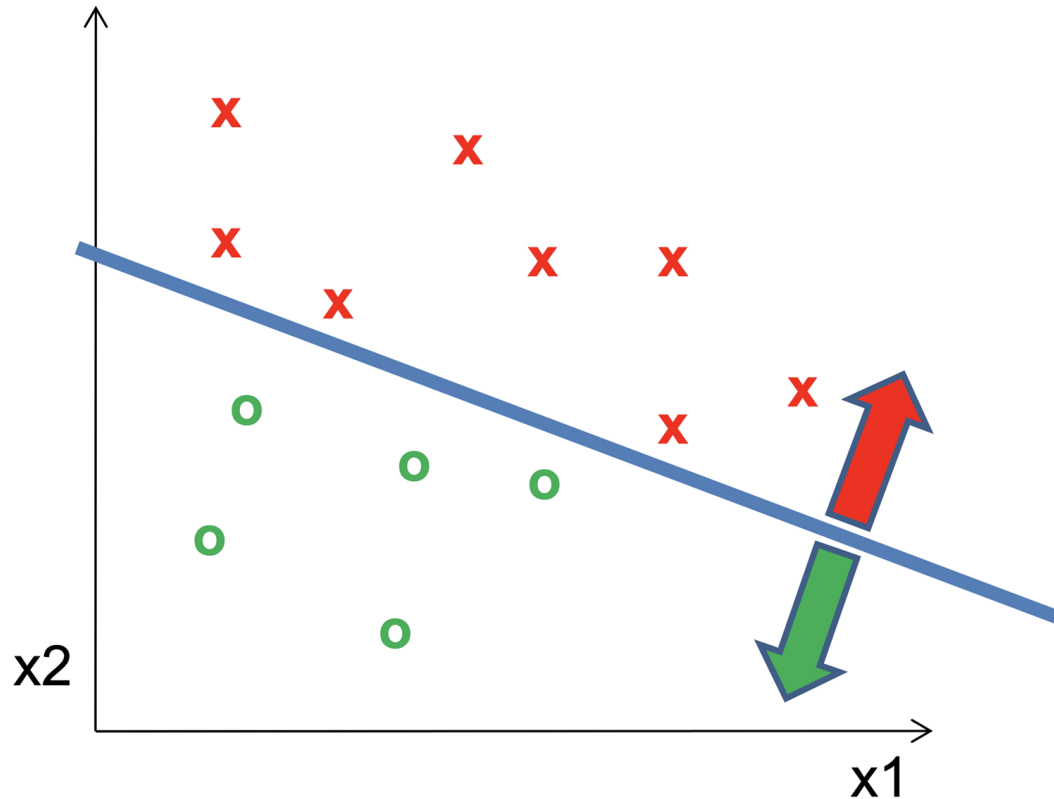
**2\_feature\_matching.ipynb**

# Classifier

Support Vector Machine (SVM)

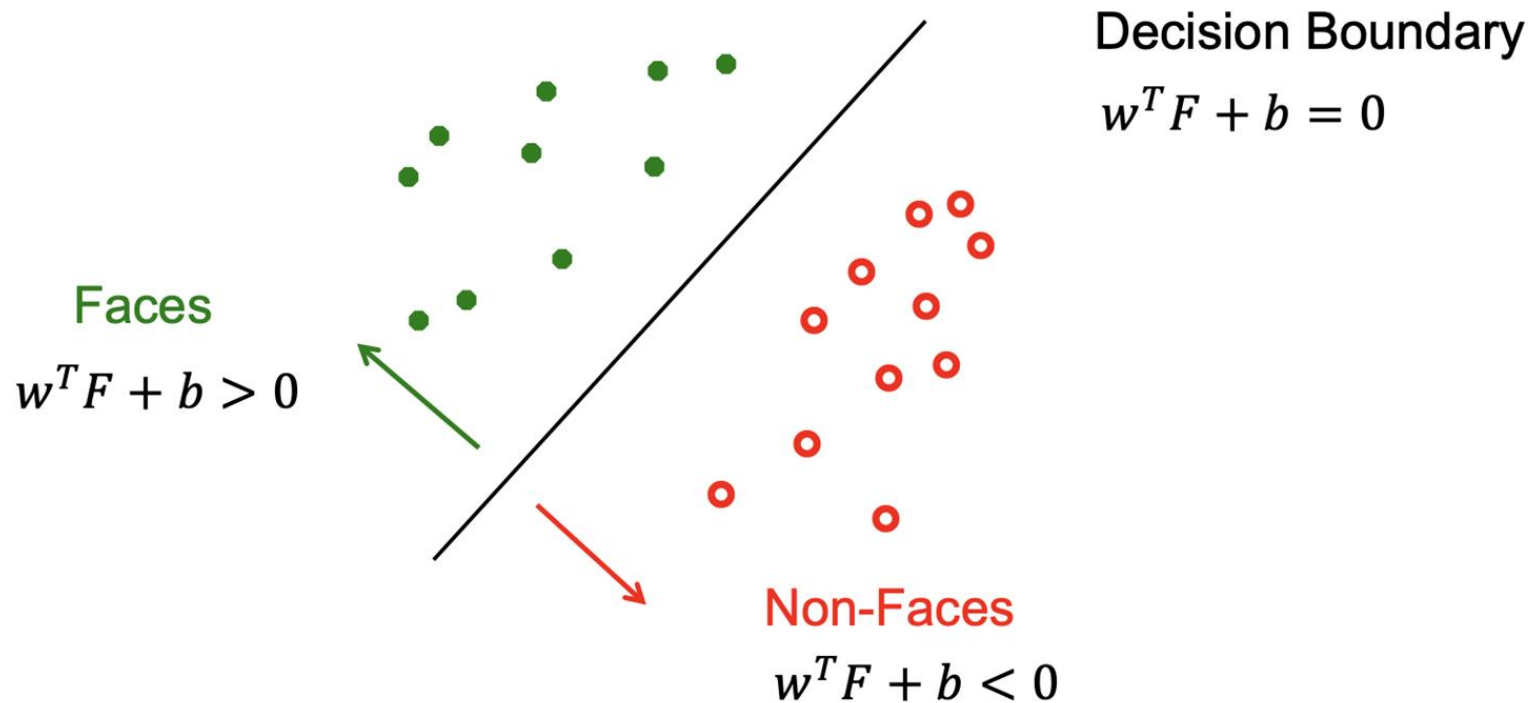
# Classifier

- A classifier maps from the feature space to label
  - Training labels dictate that two examples are the same or different, in some sense
  - We want the simplest function that is confidently correct



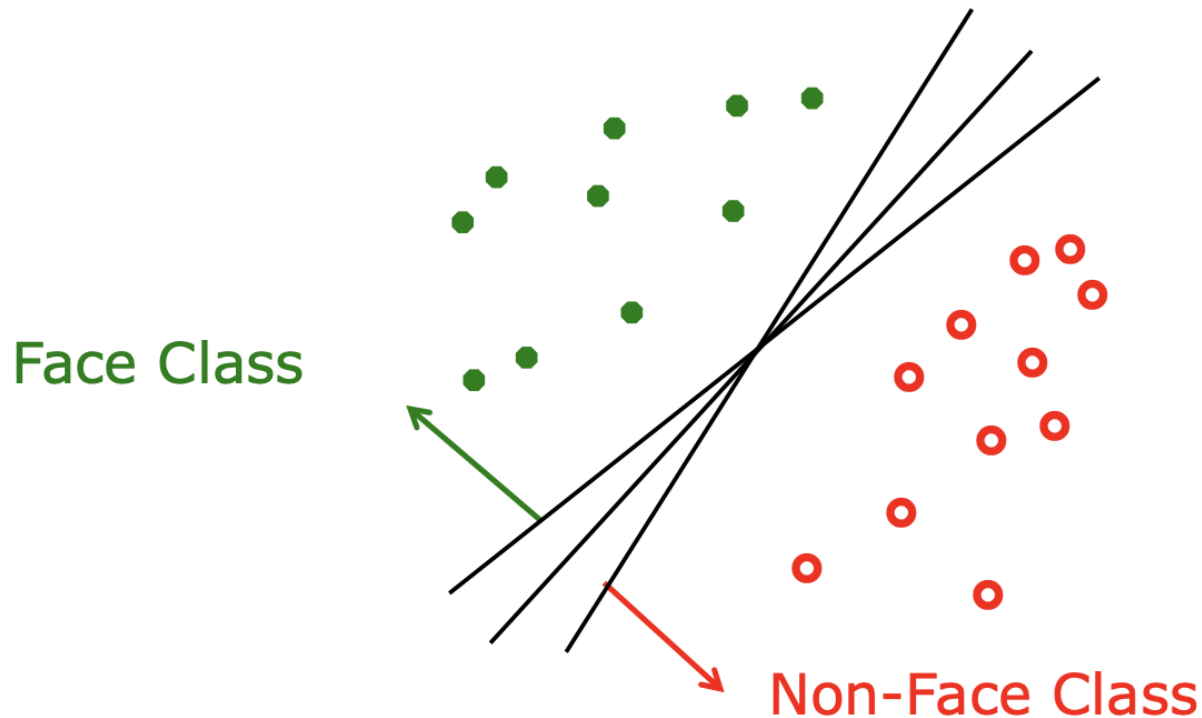
# Decision Boundary

- Find a decision boundary in a feature space



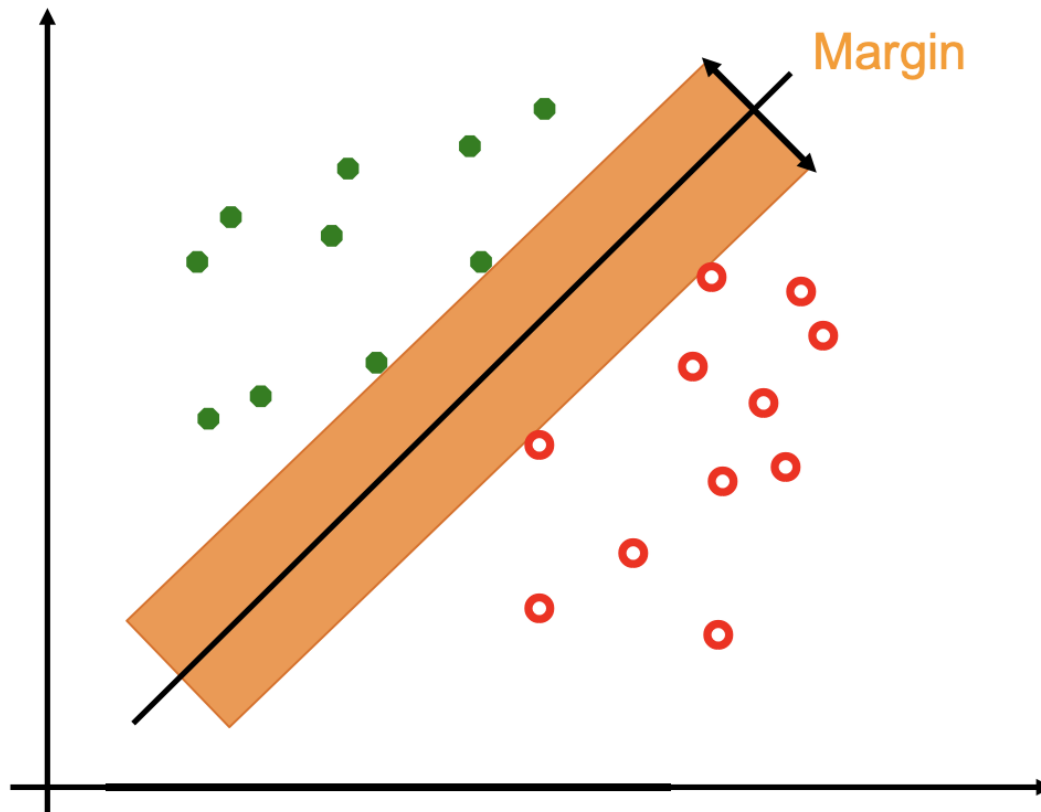
# Decision Boundary

- How to find the **optimal** decision boundary?



# Evaluating a Decision Boundary

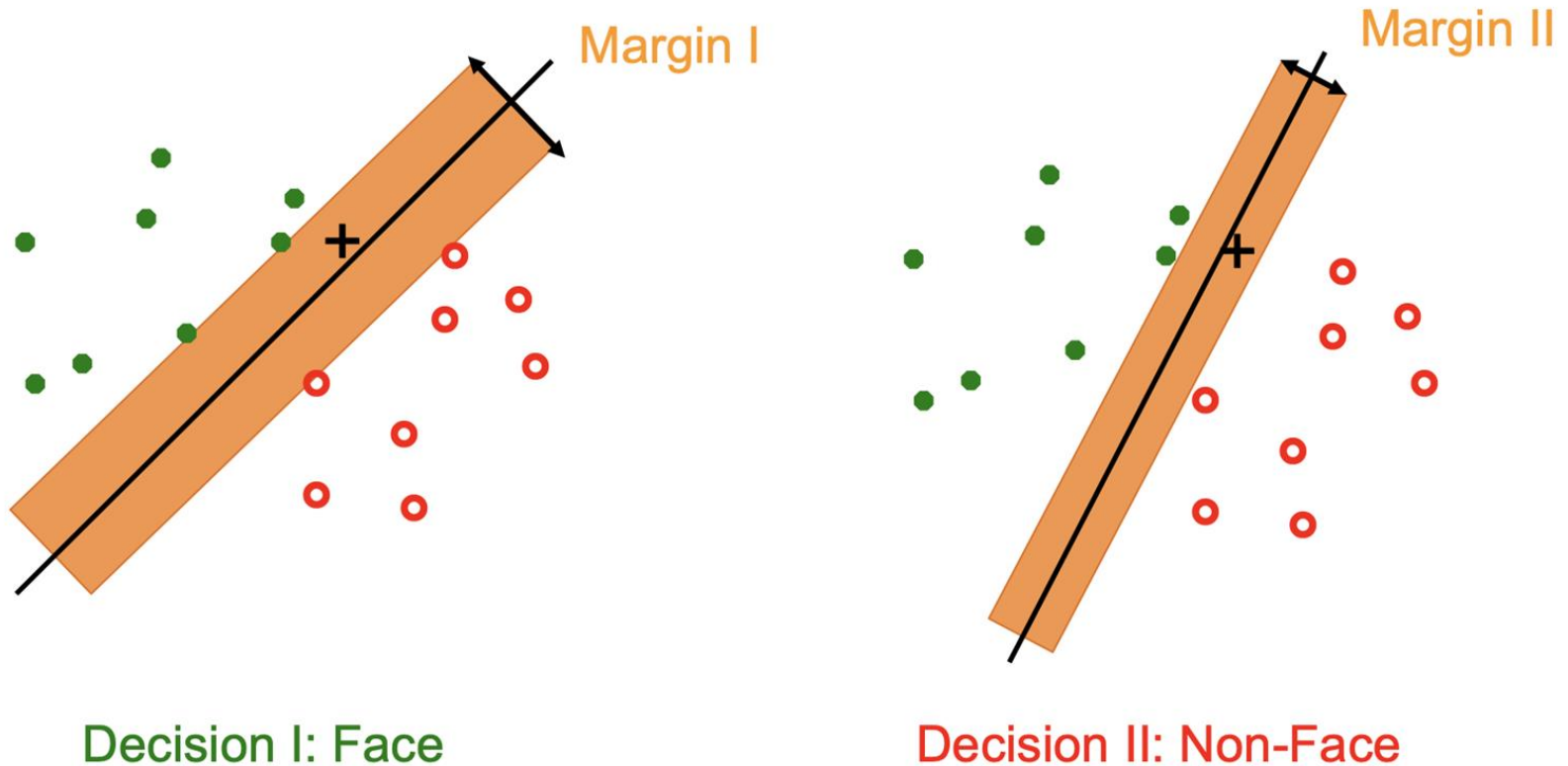
- **Margin** or **safe zone**: The width that the boundary could be increased by before hitting a data point





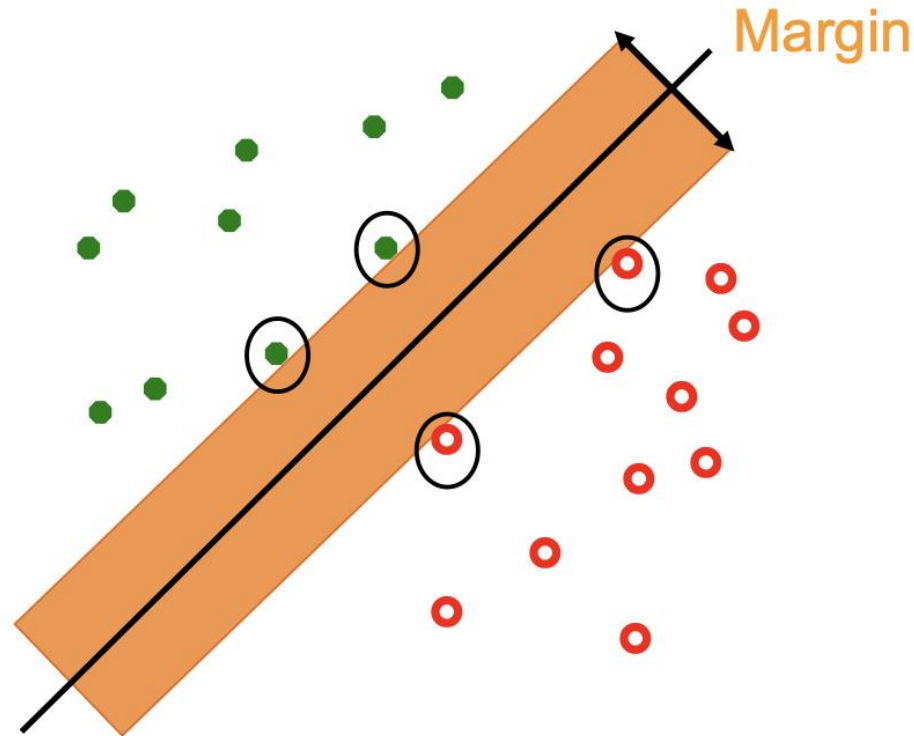
# Evaluating a Decision Boundary

- Choose decision boundary with the **maximum margin**



# Support Vector Machine (SVM)

- Support vectors: Closest data samples to the boundary
- Decision boundary and the margin depend only on the support vectors



# Bag-of-Words based Image Classification

Image classification

Image feature

Dense SIFT(PHOW)

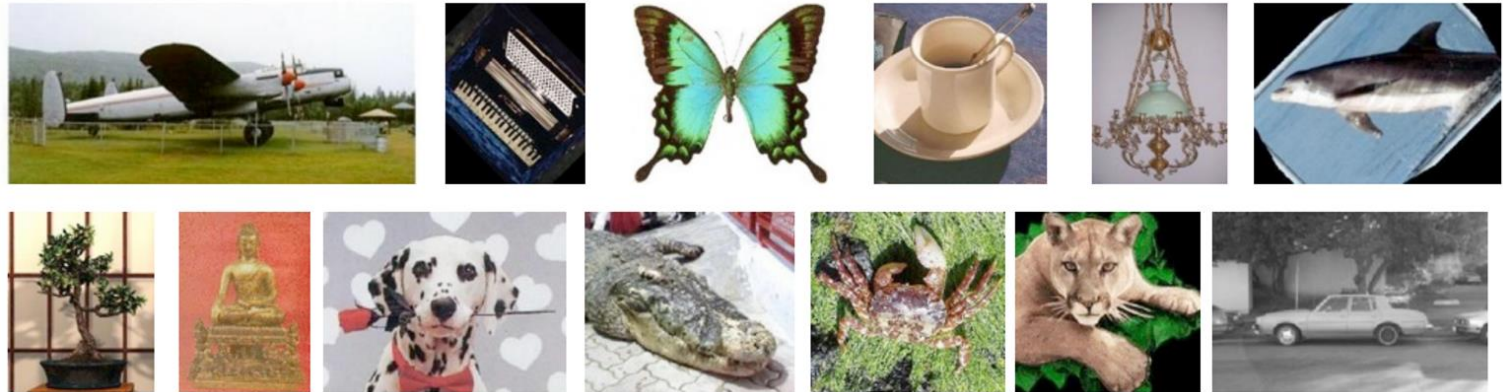
VBoW

Spatial histogram

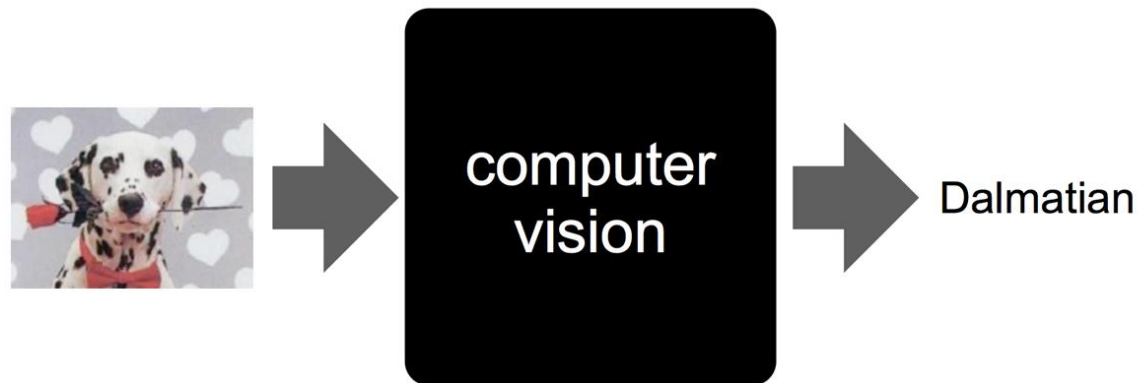
Classifier

# Image Classification

- The Dataset – Caltech 101

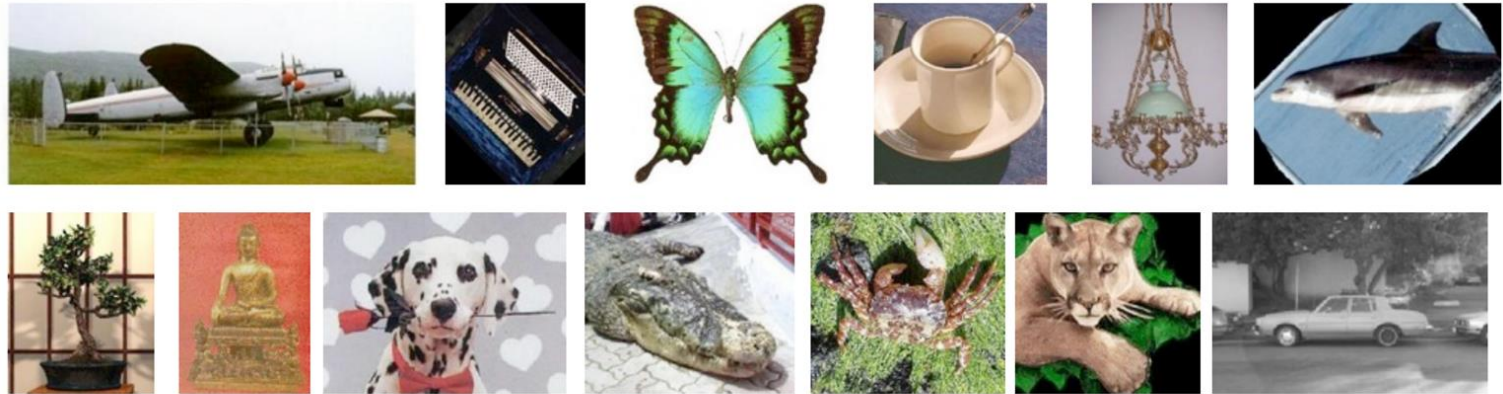


[Fei-Fei et al. 2003]

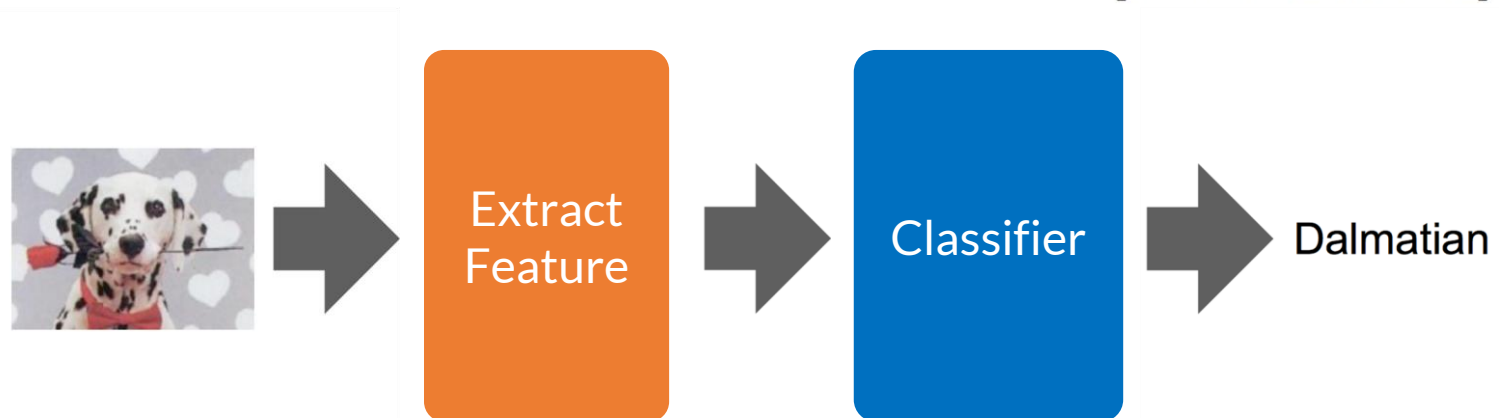


# Image Classification

- The Dataset – Caltech 101

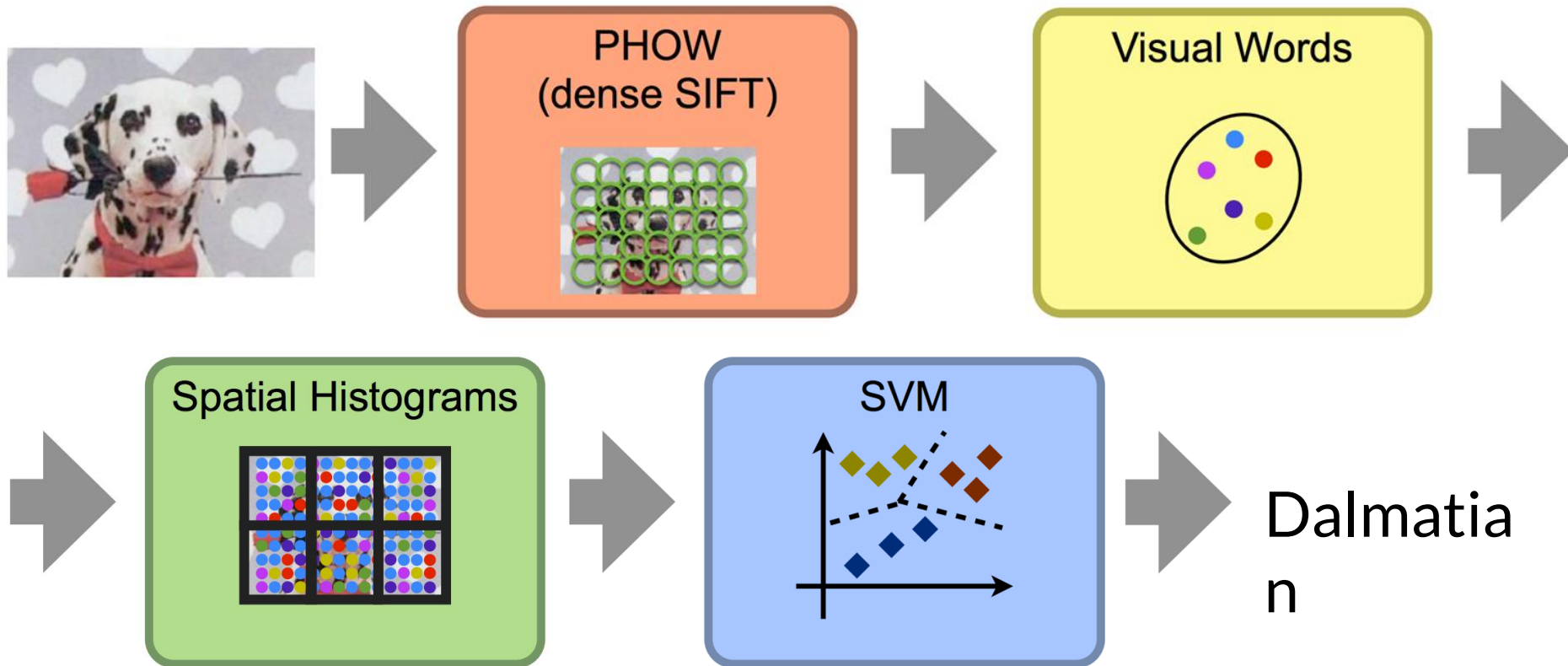


[Fei-Fei et al. 2003]



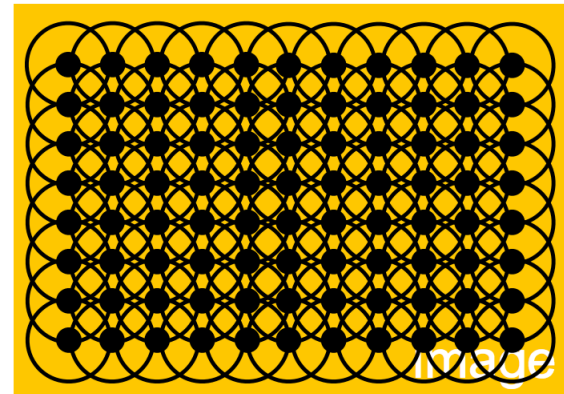
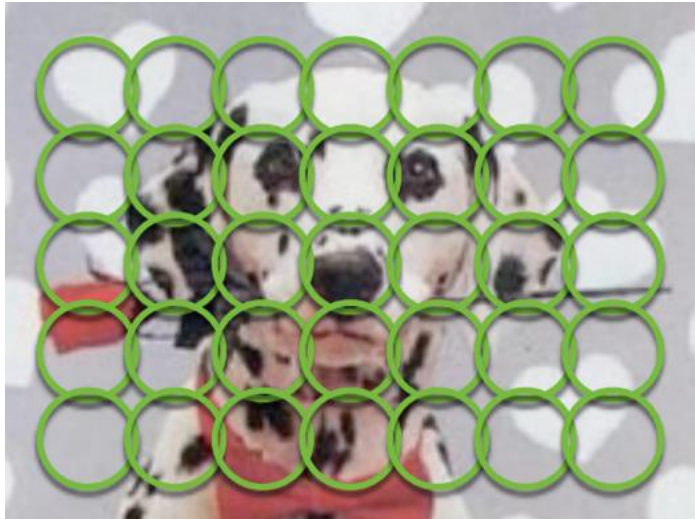
# Image Classification

- At a Glance - A Pipeline



# Feature Extraction

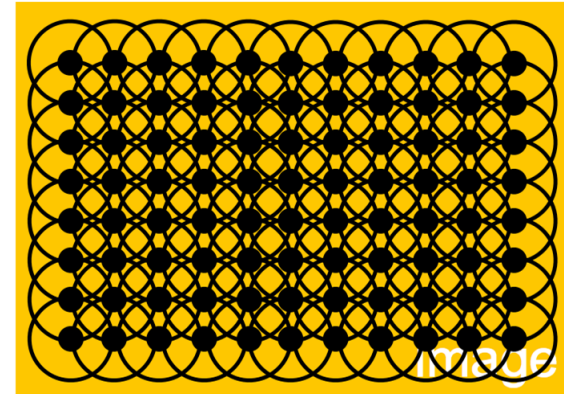
- Dense SIFT
  - Uniform keypoints – No detection
  - Dense multiscale SIFT
  - Descriptors from the uniform keypoints





# Feature Extraction

- Dense SIFT
  - Uniform keypoints – No detection
  - Dense multiscale SIFT
  - Descriptors from the uniform keypoints

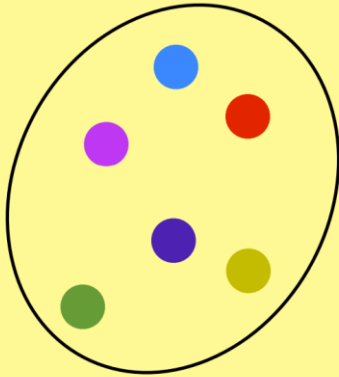


```
sift = cv2.xfeatures2d.SIFT_create()
# Dense SIFT(Extract SIFT descriptor in grid points over an image)
def denseSIFT(img, step = 5, size = 7):
    rows, cols = img.shape[:2]
    kp = []
    for x in xrange(step, cols, step):
        for y in xrange(step, rows, step):
            kp.append(cv2.KeyPoint(x, y, size))
    kp, des = sift.compute(img, kp)
    return des
```

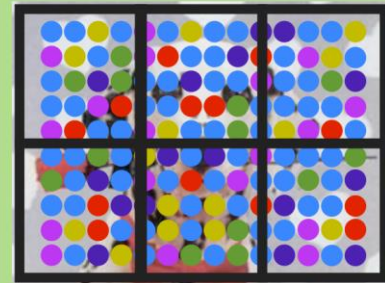


# Feature Extraction

Visual Words

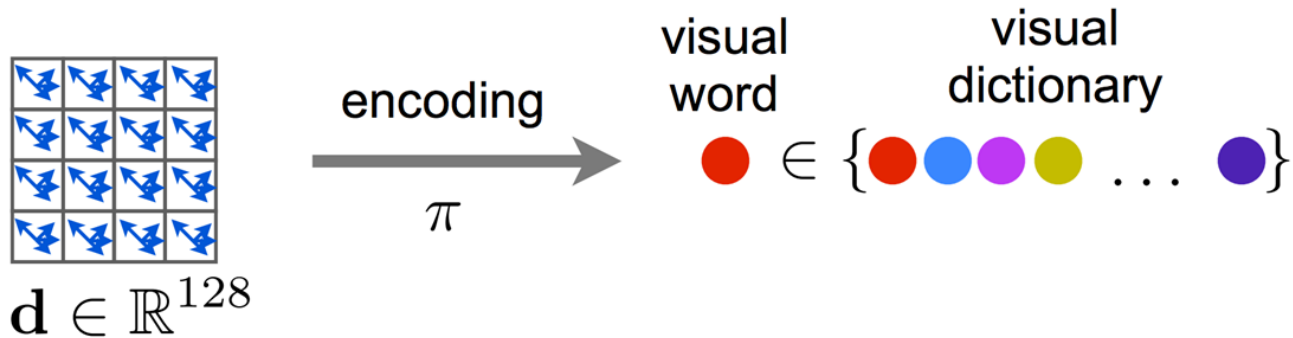


Spatial Histograms



# Feature Extraction

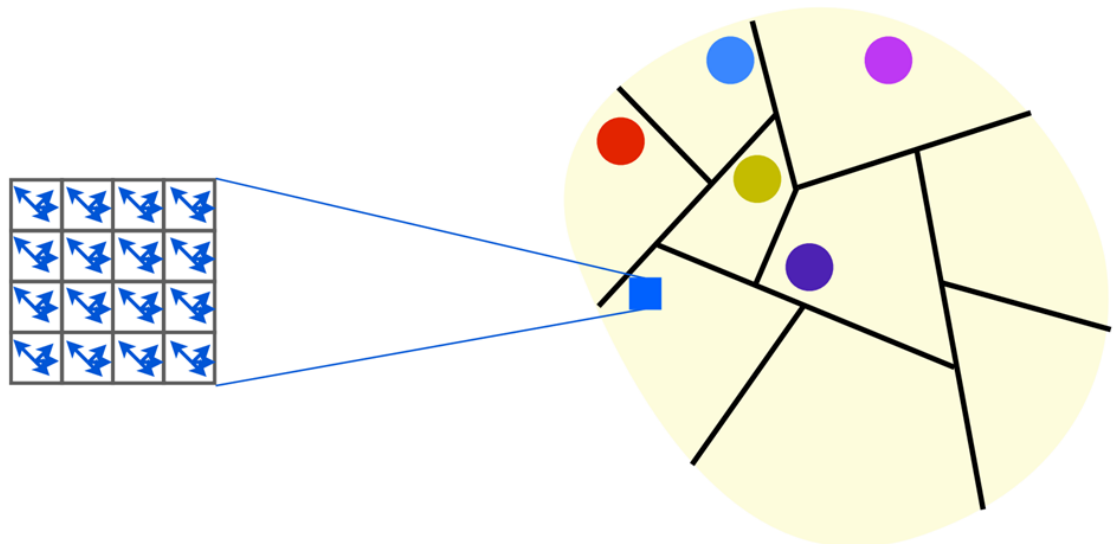
- Visual words



- Encoding = clustering

- vector quantization (k-means)  
[Lloyd 1982]
- agglomerative clustering  
[Leibe et al. 2006]
- affinity propagation  
[Frey and Dueck 2007]
- ...

[Sivic and Zisserman 2003]



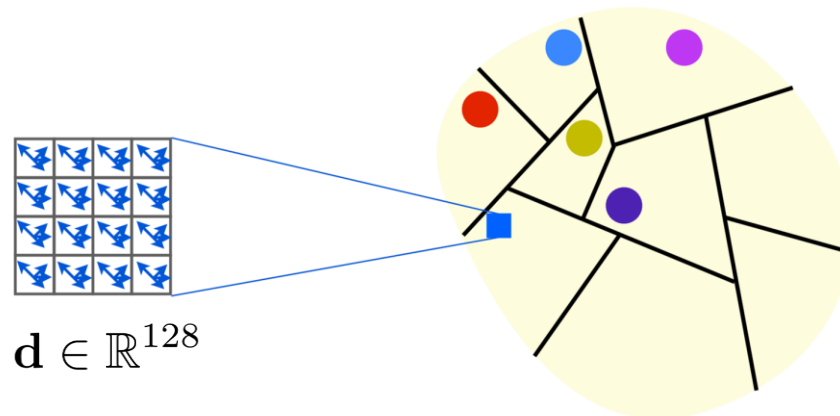
# Feature Extraction

- K-means clustering in OpenCV

```
# Quantize the descriptors to get the visual words
print "Running K-means clustering (%d -> %d)..." % (PHOW_descrs.shape[0],
                                                    numWords)

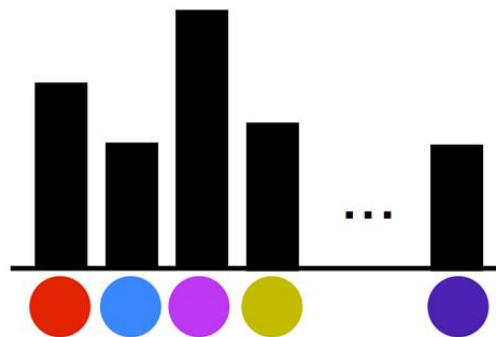
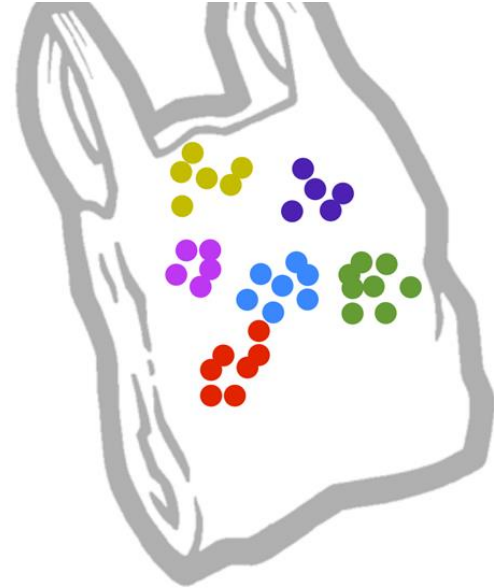
criteria = (cv2.TERM_CRITERIA_EPS + cv2.TERM_CRITERIA_MAX_ITER, 500, 1.0)
attempts = 10
flags = cv2.KMEANS_RANDOM_CENTERS
retval, bestLabels, vocab = cv2.kmeans(PHOW_descrs, numWords,
                                       None, criteria, attempts, flags)

# match SIFT features to the words from K-means
bf = cv2.BFMatcher()
matches = bf.knnMatch(des, vocab, k=1)
words = [m[0].trainIdx for m in matches]
```



# Feature Extraction

- Visual Bag of Words

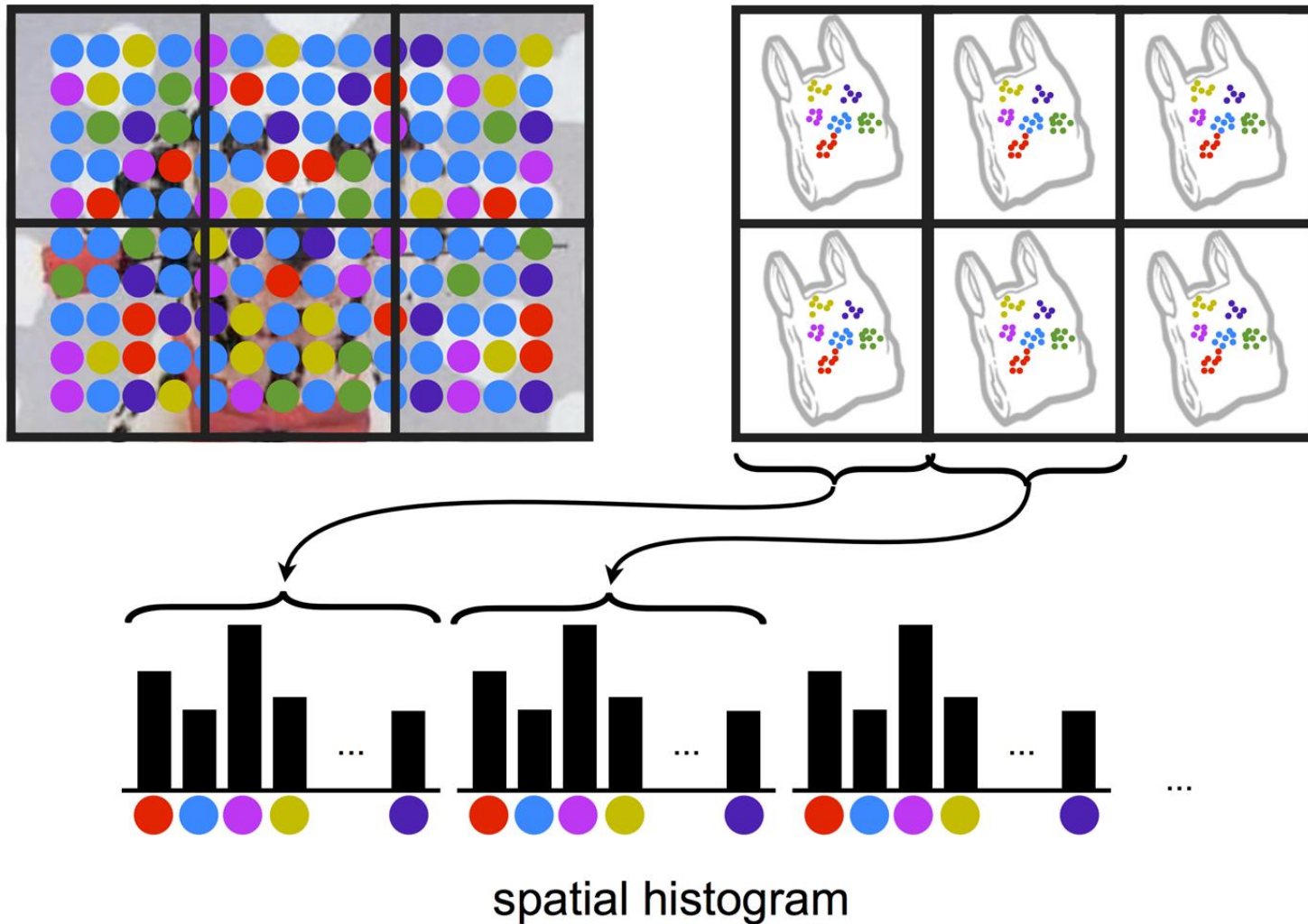


histogram (bag) of visual words

[Csurka et al. 2004]

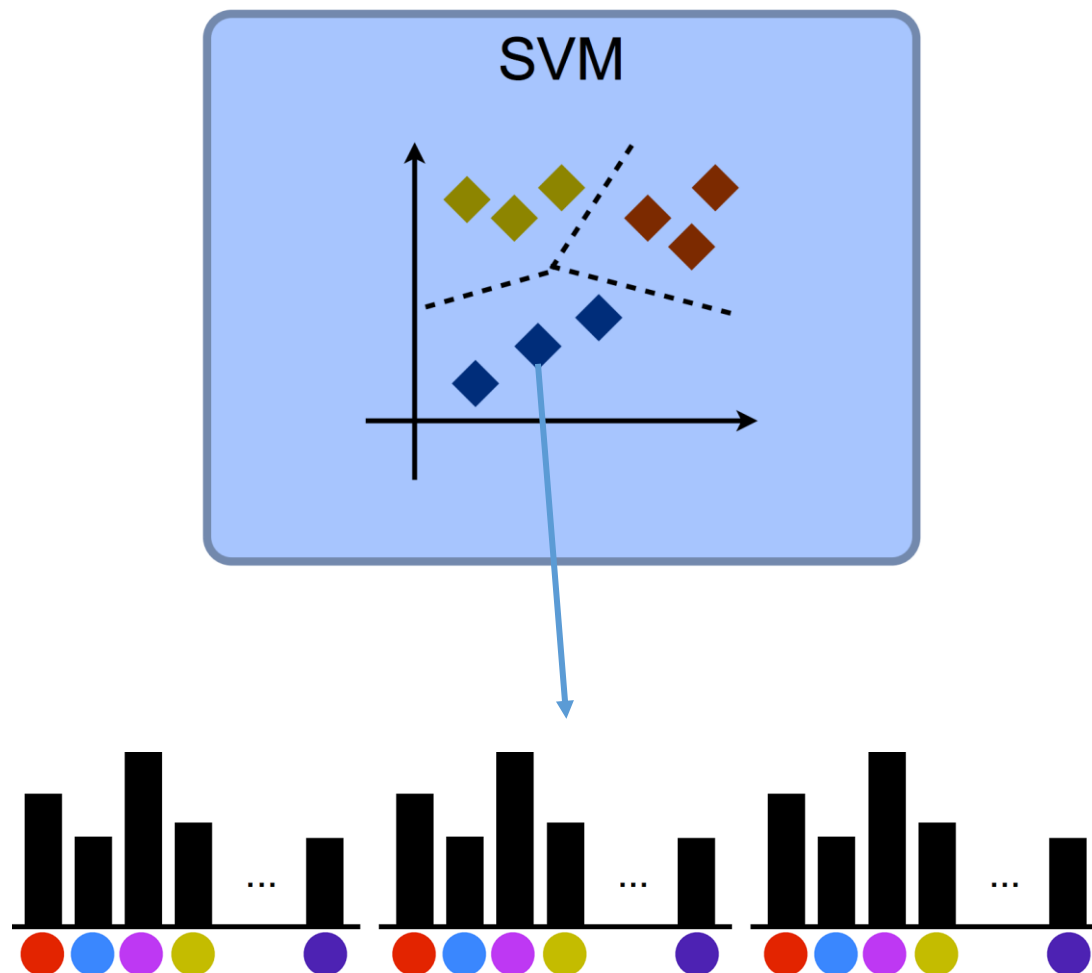
# Feature Extraction

- Spatial Histogram



# Classifier

- SVM



# Classifier

- SVM in OpenCV
  - Training SVM

```
# Train SVM
print 'Training SVM...'
svm = cv2.ml.SVM_create()
svm.setType(cv2.ml.SVM_C_SVC) # classification(n > 2)
svm.setKernel(cv2.ml.SVM_LINEAR) # linear kernel
svm.setC(0.01)
svm.setTermCriteria((cv2.TERM_CRITERIA_COUNT, 10, 1.0)) # term. criteria

svm.train(train_bow, cv2.ml.ROW_SAMPLE, train_labels)
```

- Predict with SVM

```
train_preds = svm.predict(train_bow)[1]
print('Training Accuracy: %.6f' % np.average(train_preds == train_labels))
```

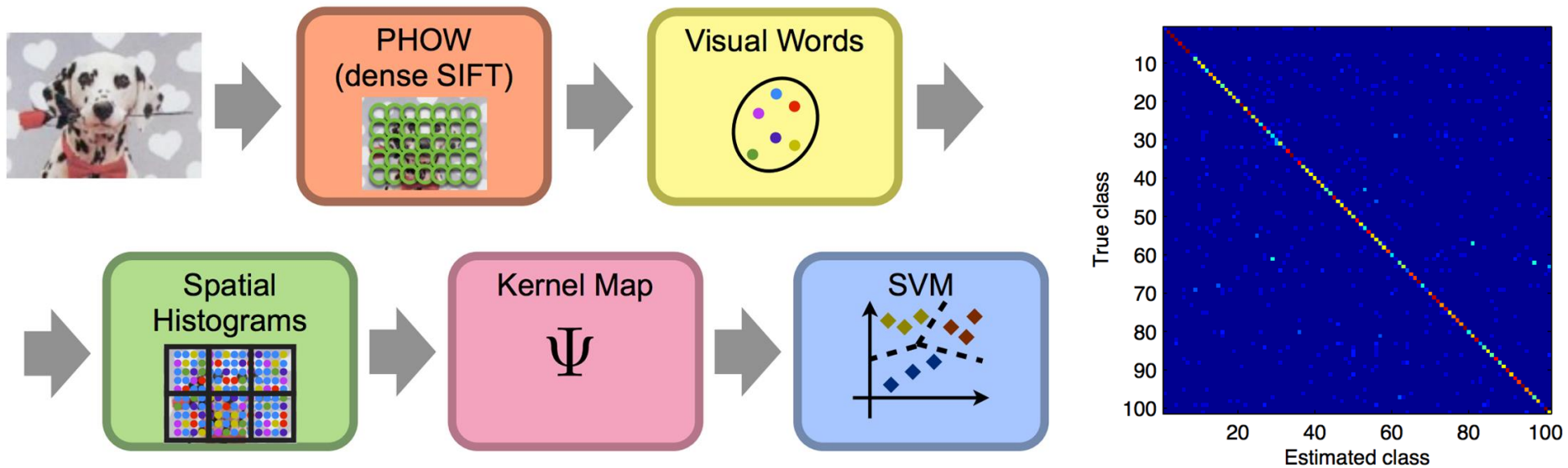
Training Accuracy: 0.990850



# Image Classification

- Results

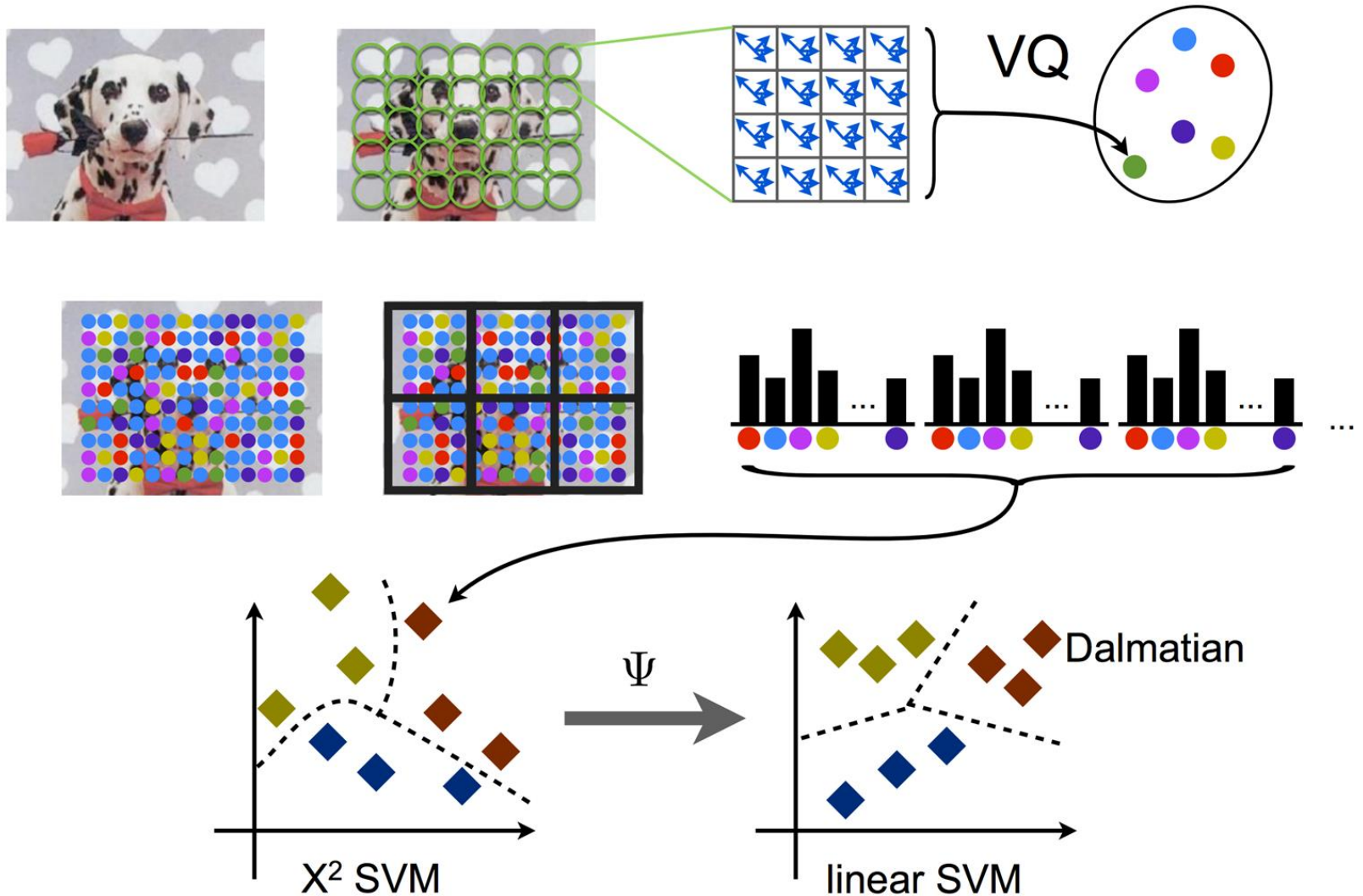
- ~60% Accuracy using Dense SIFT / VBoW / Spatial Histogram





# Image Classification

- Image Classification Summary



Let's Check the Code

**3\_classification.ipynb**

# Trouble-shooting

- Kernel crashes with ModuleNotFoundError on 'prompt\_toolkit.formatted\_text'
  - <https://github.com/jupyter/notebook/issues/4050>
- OpenCV(3.4.3)  
/io/opencv\_contrib/modules/xfeatures2d/src/sift.cpp:1207: error
  - <https://stackoverflow.com/questions/52305578/sift-cv2-xfeatures2d-sift-create-not-working-even-though-have-contrib-instal>