

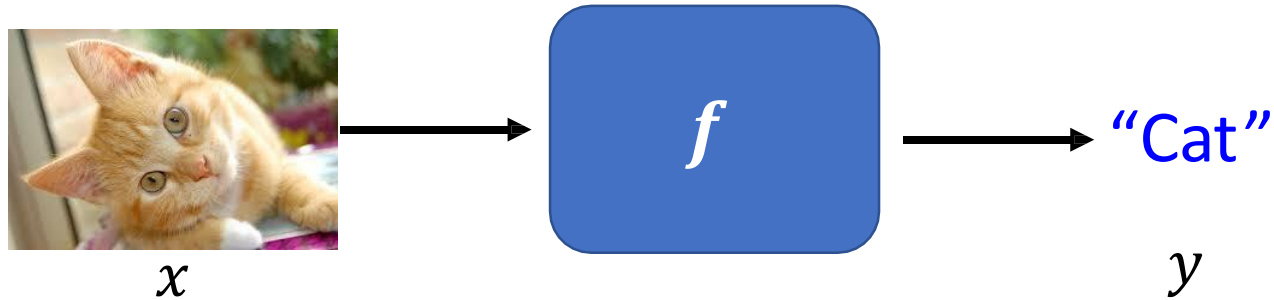
COMP SCI 1400

AI Technologies

Computer Vision Intro
Dr. Kamal Mammadov

What is CV

Supervised Learning



x_1 :



y_1 : "Cat"

x_2 :



y_2 : "Cat"

x_3 :



y_3 : "Dog"

x_4 :



y_4 : "Dog"

Labelled Data

- Digital images



08	02	22	97	38	15	00	40	00	75	04	05	07	78	52	12	50	77	91	82
49	49	99	40	17	81	18	57	60	87	17	40	98	43	69	45	09	56	62	00
81	49	31	73	55	79	14	29	93	71	40	67	53	88	30	03	49	13	36	65
92	70	95	23	04	60	11	42	69	31	68	56	01	32	56	71	37	02	36	91
22	31	16	71	51	62	83	89	41	92	36	54	22	40	40	28	66	33	13	80
24	47	32	60	99	03	45	02	44	75	33	53	78	36	84	20	35	17	12	50
32	98	81	28	64	23	67	10	26	38	40	67	59	54	70	66	18	38	64	70
67	26	20	68	02	62	12	20	95	63	94	39	63	08	40	91	66	49	94	21
24	55	58	05	66	73	99	26	97	17	78	78	96	83	14	88	34	89	63	72
21	36	23	09	75	00	76	44	20	45	35	14	00	61	33	97	34	31	33	95
78	17	53	28	22	75	31	67	15	94	03	80	04	62	16	14	09	53	56	92
16	39	05	42	96	35	31	47	55	58	88	24	00	17	54	24	36	29	85	57
86	56	00	48	35	71	89	07	05	44	44	37	44	60	21	58	51	54	17	58
19	80	81	68	05	94	47	69	28	73	92	13	86	52	17	77	04	89	55	40
04	52	08	83	97	35	99	16	07	97	57	32	16	26	26	79	33	27	98	66
85	36	68	87	57	62	20	72	03	46	33	67	46	55	12	32	63	93	53	69
04	42	16	73	38	85	39	11	24	94	72	18	08	46	29	32	40	62	76	36
20	69	36	41	72	30	23	88	34	82	99	69	82	67	59	85	74	04	36	16
20	73	35	29	78	31	90	01	74	31	49	71	48	54	81	16	23	57	05	54
01	70	84	71	83	51	54	69	16	92	33	48	61	43	52	01	89	25	62	48

What the computer sees

image classification → 82% cat
15% dog
2% hat
1% mug

Three Categories of Tasks in CV

Computer Vision: Stages

- Image formation
- Low-level
 - Single image processing
 - Multiple views
- Mid-level
 - Estimation, segmentation (main topic of Image Analysis and Foundations of Image Analysis and will only be covered briefly here)
- High-level
 - Recognition
 - Classification

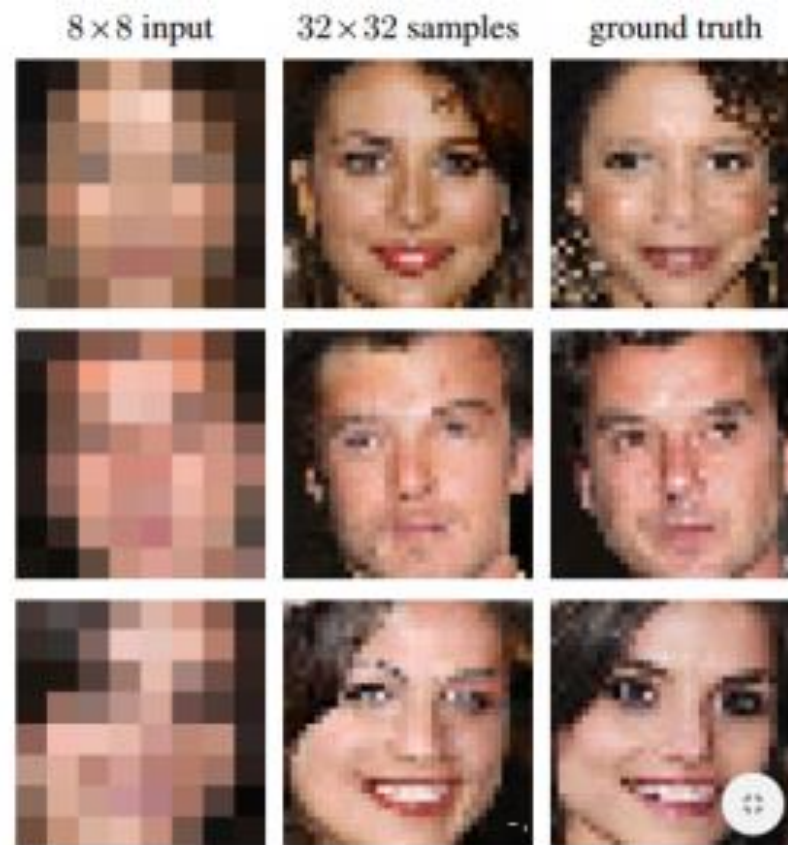


The IT University
of Copenhagen

Low level CV --- Denoising



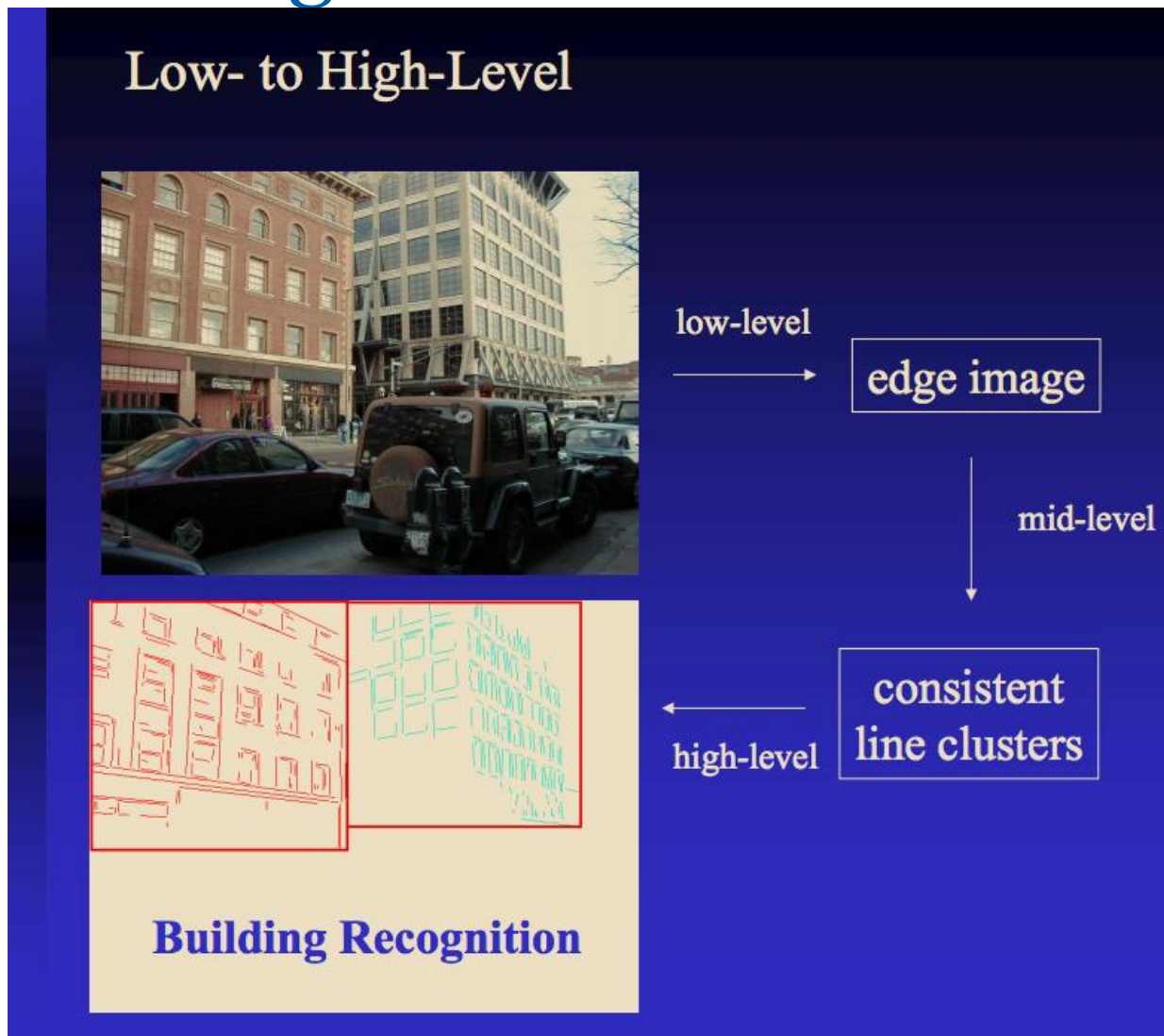
Low level CV --- Super-Resolution



Low level CV --- Dehaze

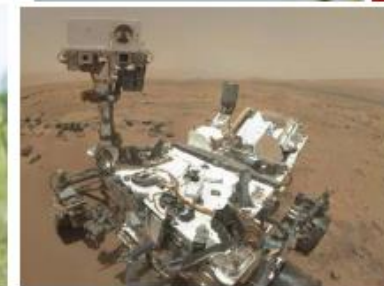


Three Categories of Tasks in CV



History of CV

- **Data --- images**

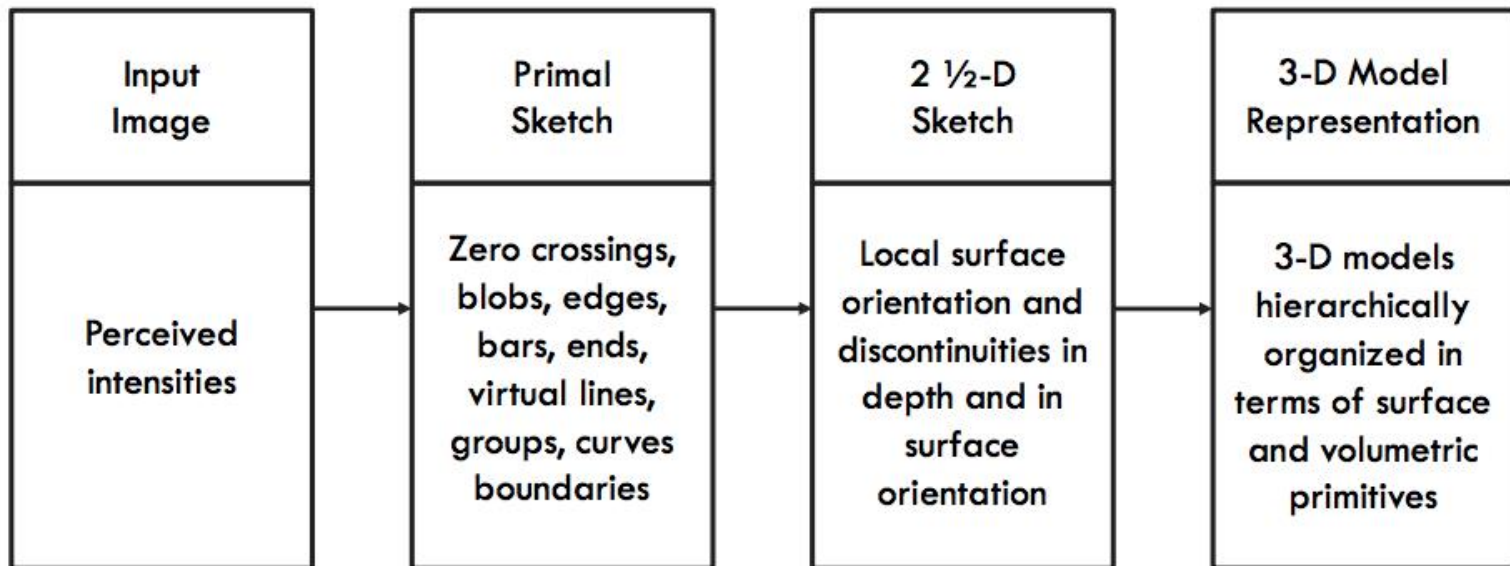
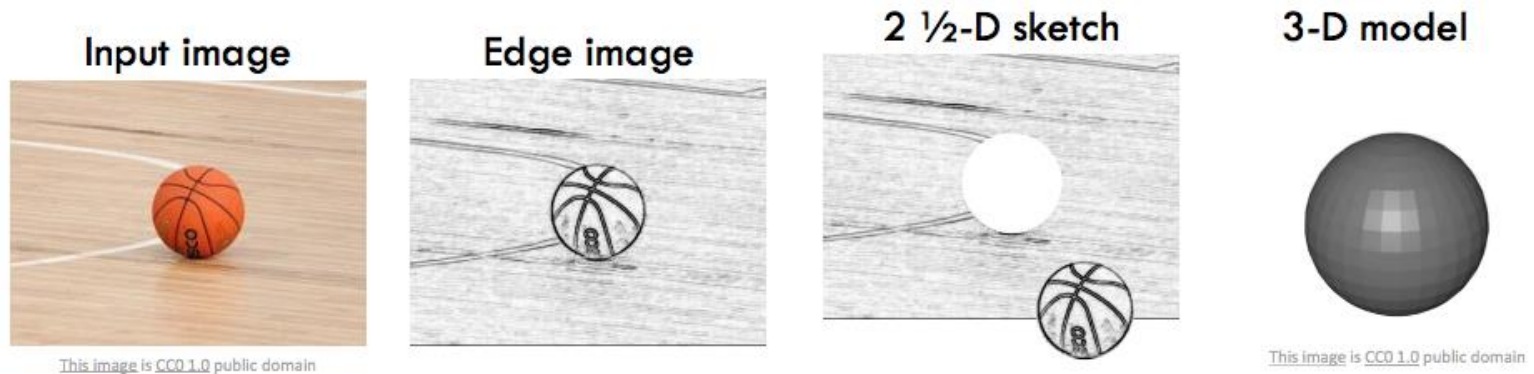


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Emerging of Computer Vision



Stages of Visual Representation, David Marr, 1970s

- **Canny Edge Detector**

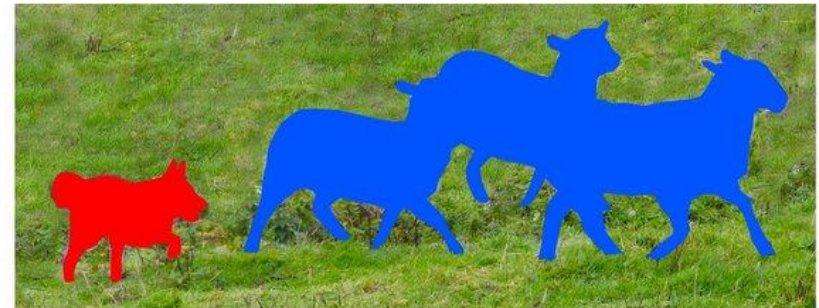


By John F. Canny in 1986

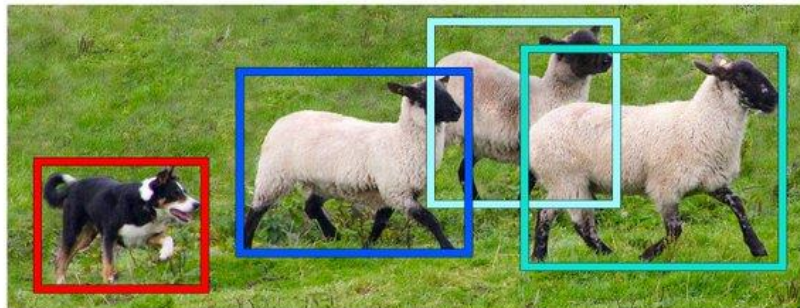
- **Computer Vision tasks**



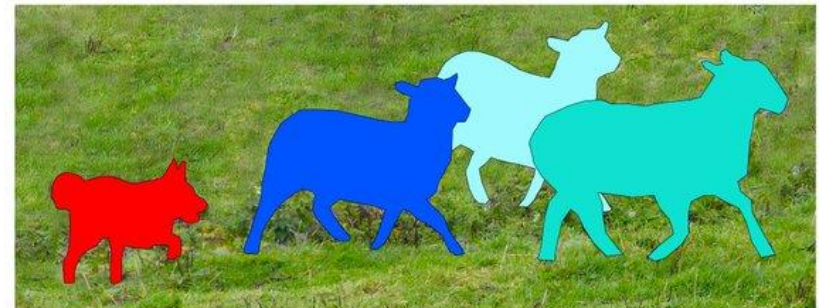
Image Recognition



Semantic Segmentation



Object Detection



Instance Segmentation

- # History of Computer Vision



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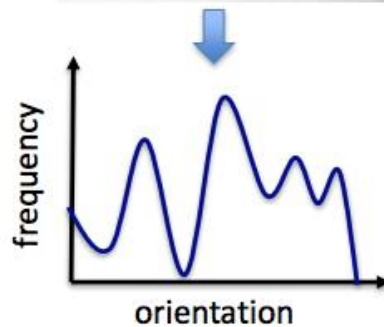


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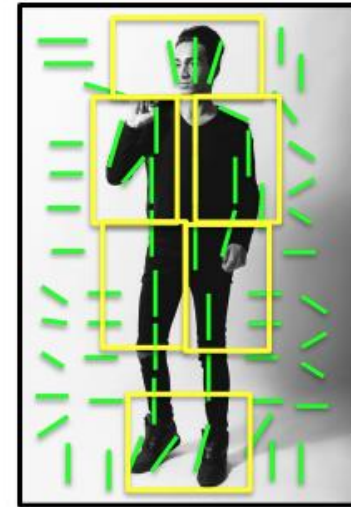
“SIFT” & Object Recognition, David Lowe, 1999

- History of Computer Vision

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Histogram of Gradients (HoG)
Dalal & Triggs, 2005



Deformable Part Model
Felzenswalb, McAllester, Ramanan, 2009

- History of Computer Vision

PASCAL Visual Object Challenge (20 object categories)

[Everingham et al. 2006-2012]

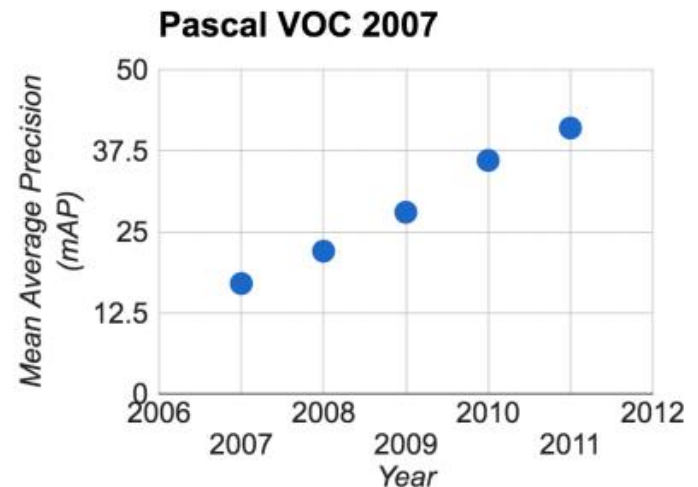
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- # History of Computer Vision



IMGENET

www.image-net.org

22K categories and **15M** images


- Animals
 - Bird
 - Fish
 - Mammal
 - Invertebrate
- Plants
 - Tree
 - Flower
 - Food
 - Materials
- Structures
 - Artifact
 - Tools
 - Appliances
 - Structures
- Person
 - Scenes
 - Indoor
 - Geological Formations
 - Sport Activities



- History of Computer Vision

IMAGENET Large Scale Visual Recognition Challenge

The Image Classification Challenge:
1,000 object classes
1,431,167 images



Output:
Scale
T-shirt
Steel drum
Drumstick
Mud turtle

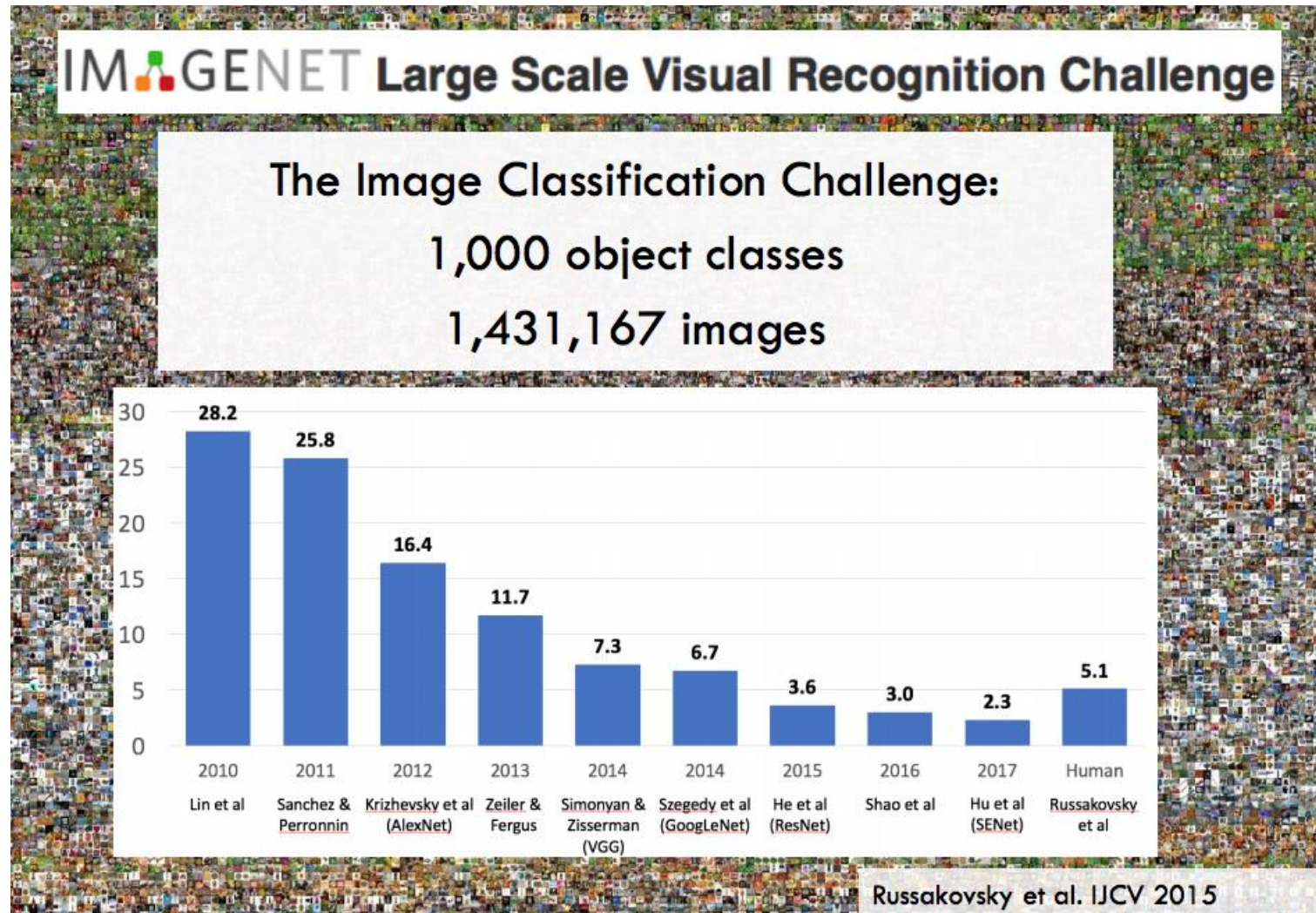
✓

Output:
Scale
T-shirt
Giant panda
Drumstick
Mud turtle

✗

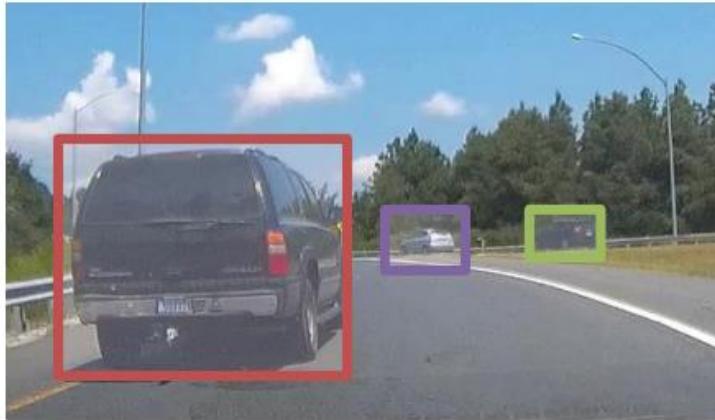
Russakovsky et al. IJCV 2015

- History of Computer Vision



CV Tasks

- # Computer Vision tasks



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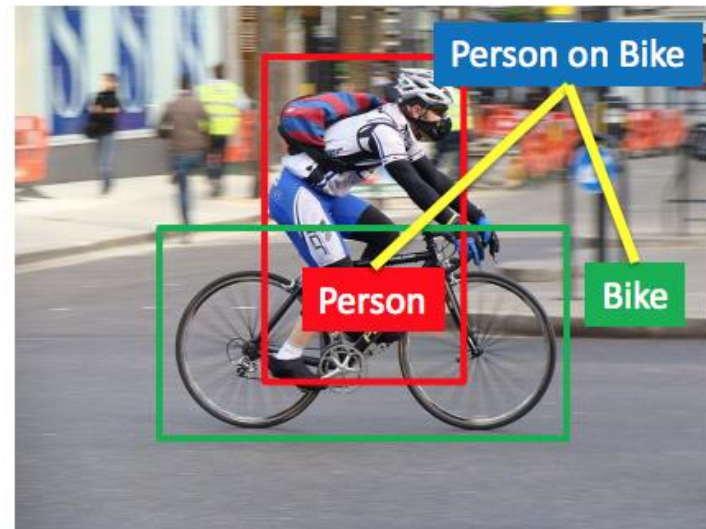
- Object detection
- Action classification
- Image captioning
- ...



Person

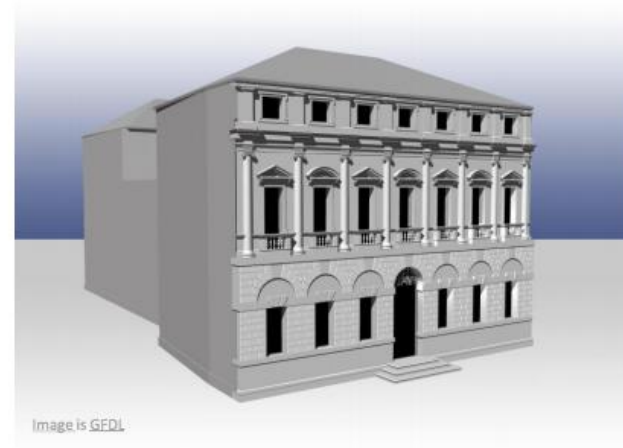
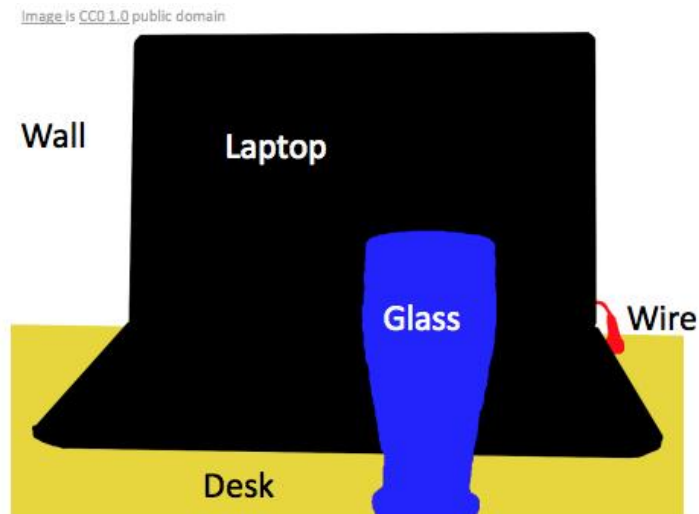
Hammer

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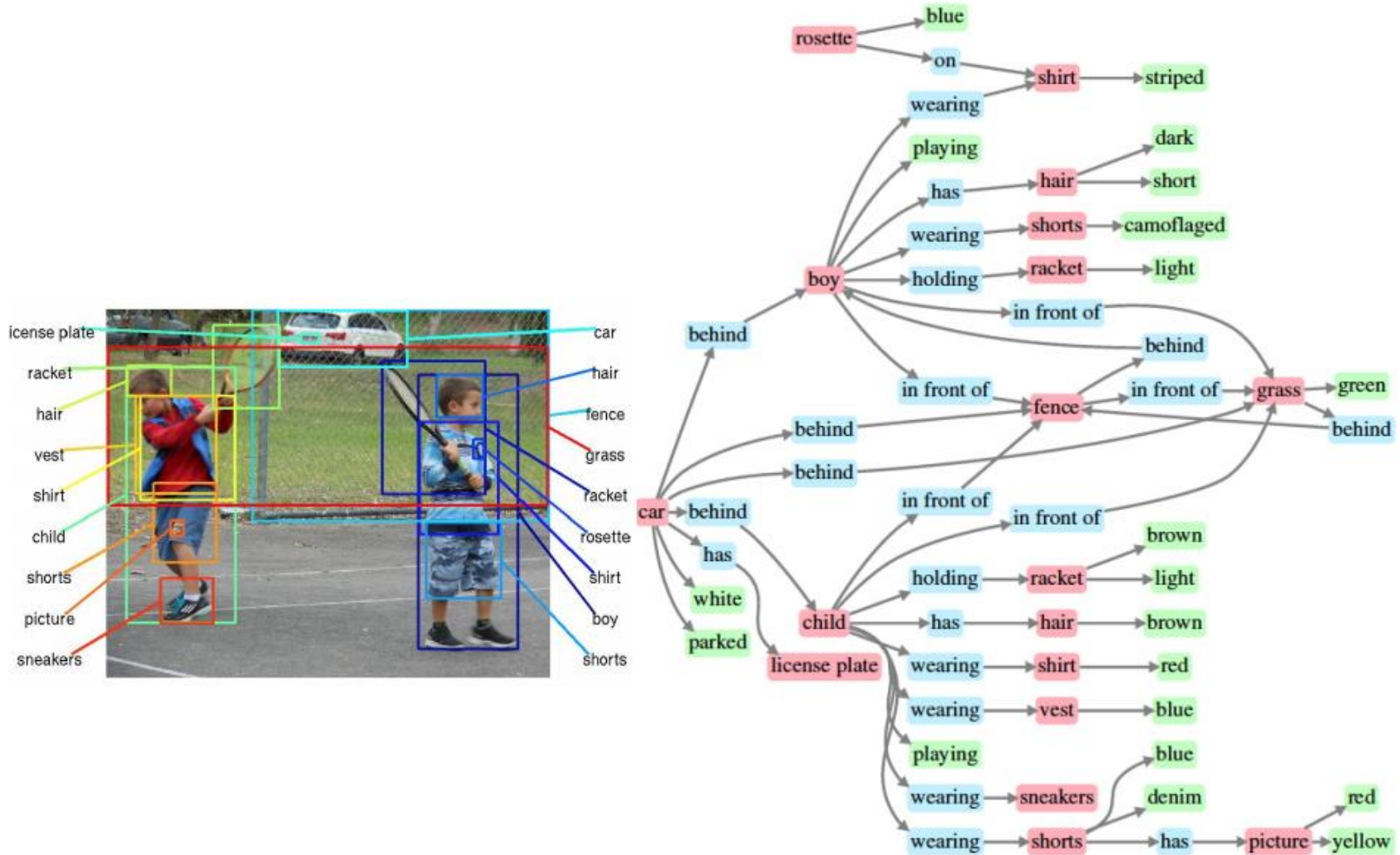


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- **Computer Vision tasks**



- Computer Vision tasks



Johnson *et al.*, "Image Retrieval using Scene Graphs", CVPR 2015

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Deep Dream



Style Transfer

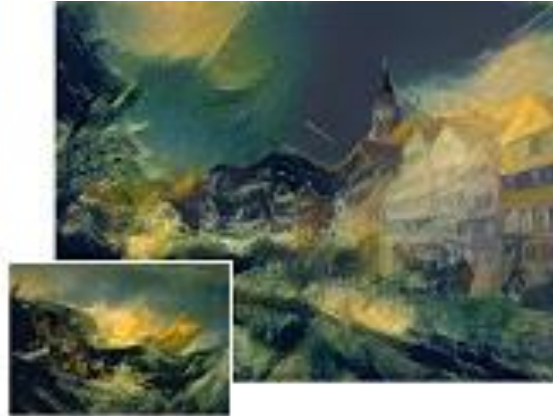


Image Classification with KNN

- **Problems**

Image Classification: A core task in Computer Vision



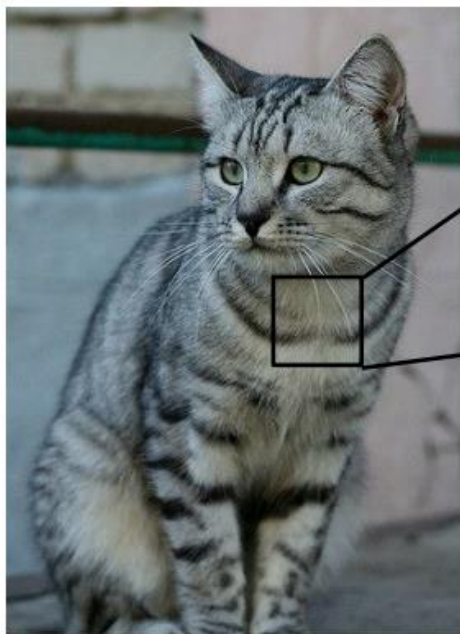
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(assume given set of discrete labels)
{dog, cat, truck, plane, ...}

—————→ cat

- Problems

The Problem: Semantic Gap



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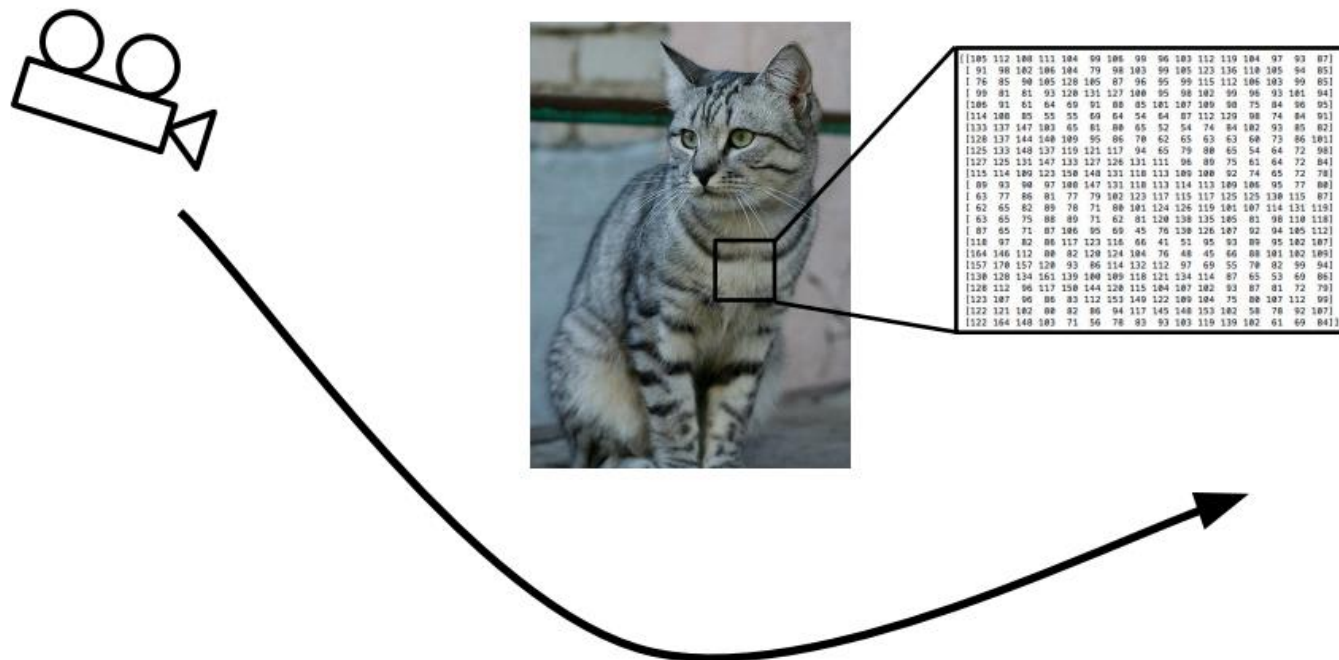
```
[[105 112 108 111 104 99 106 99 96 103 112 119 104 97 93 87]
 [ 91 98 102 106 104 79 98 103 99 105 123 136 110 105 94 85]
 [ 76 85 90 105 128 105 87 96 95 99 115 112 106 103 99 85]
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 [106 91 61 64 69 91 88 85 101 107 109 98 75 84 96 95]
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 [133 137 147 103 65 81 80 65 52 54 74 84 102 93 85 82]
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 [123 107 96 86 83 112 153 149 122 109 104 75 80 107 112 99]
 [122 121 102 80 82 86 94 117 145 148 153 102 58 78 92 107]
 [122 164 148 103 71 56 78 83 93 103 119 139 102 61 69 84]]
```

What the computer sees

An image is just a big grid of
numbers between [0, 255]:

e.g. 800 x 600 x 3
(3 channels RGB)

Challenges: Viewpoint variation



All pixels change when the camera moves!

Challenges: Background Clutter



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Challenges: Illumination



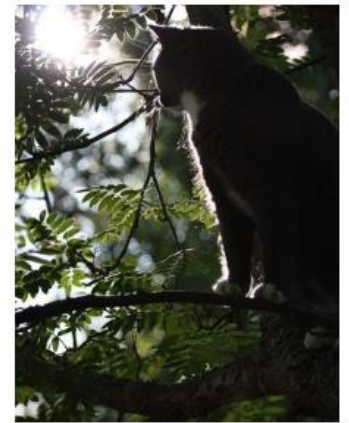
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Challenges: Deformation



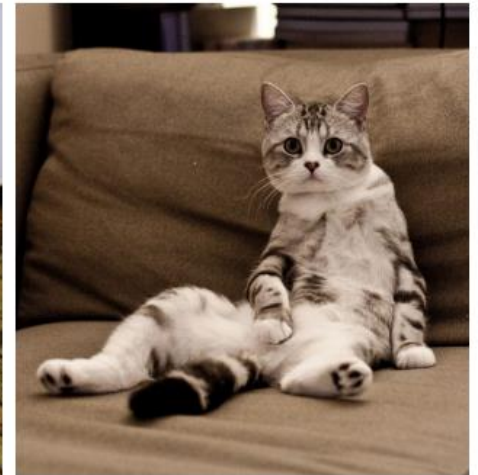
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Challenges: Occlusion



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Challenges: Intraclass variation



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Machine Learning: Data-Driven Approach

1. Collect a dataset of images and labels
2. Use Machine Learning to train a classifier
3. Evaluate the classifier on new images

Example training set

```
def train(images, labels):  
    # Machine learning!  
    return model
```

```
def predict(model, test_images):  
    # Use model to predict labels  
    return test_labels
```

airplane



automobile



bird



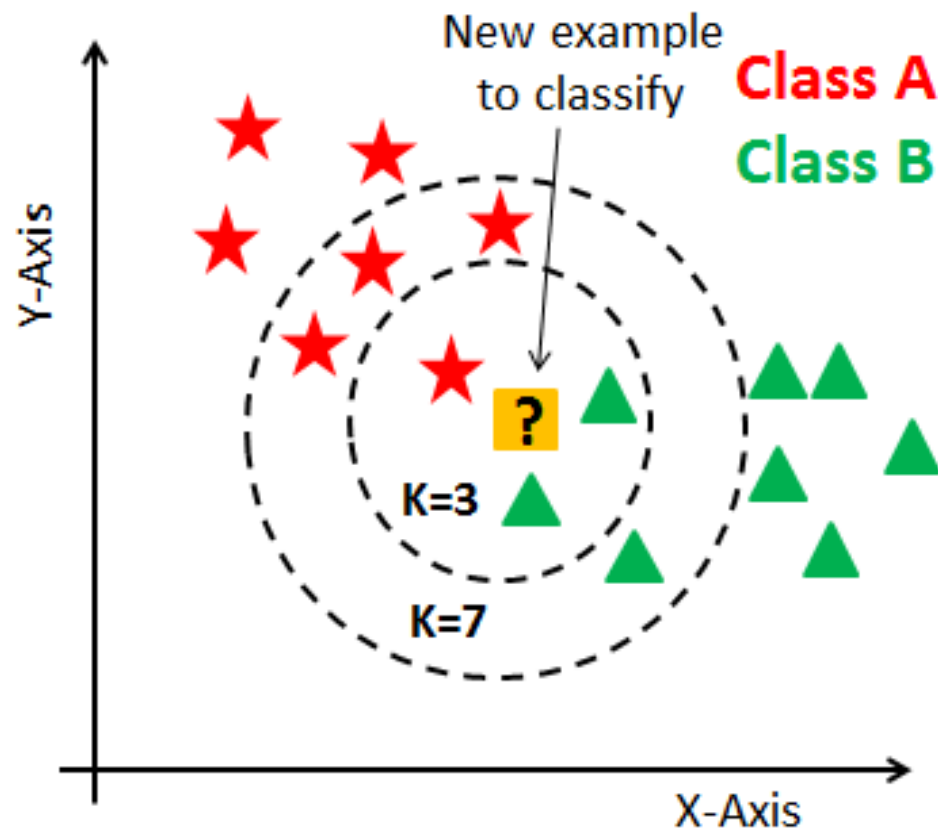
cat



deer



KNN



Example Dataset: **CIFAR10**

10 classes

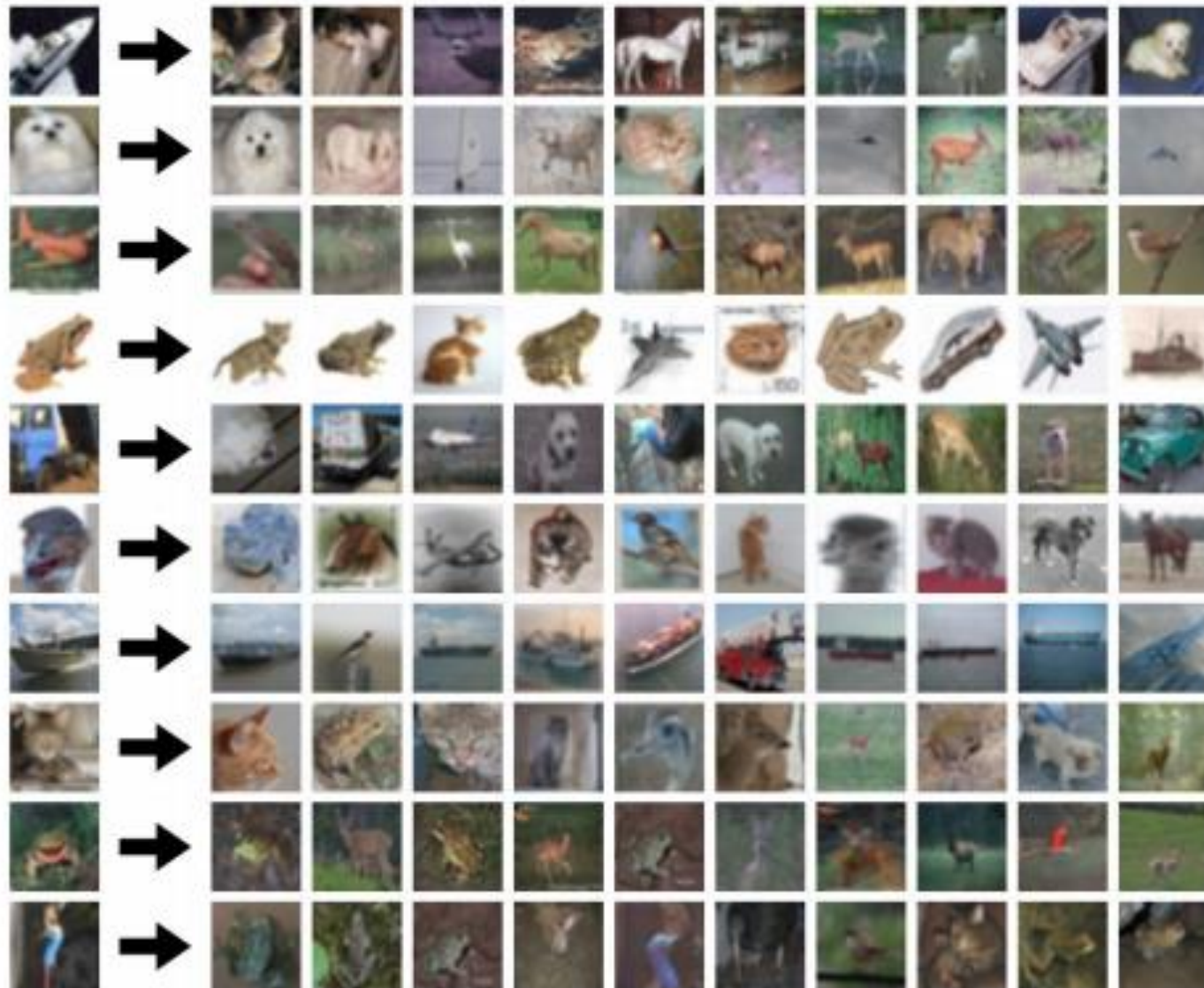
50,000 training images

10,000 testing images



Alex Krizhevsky, "Learning Multiple Layers of Features from Tiny Images", Technical Report, 2009.

Test images and nearest neighbors



KNN classifier implementation

Nearest Neighbor classifier

```
import numpy as np

class NearestNeighbor:
    def __init__(self):
        pass

    def train(self, X, y):
        """ X is N x D where each row is an example. Y is 1-dimension of size N """
        # the nearest neighbor classifier simply remembers all the training data
        self.Xtr = X
        self.ytr = y

    def predict(self, X):
        """ X is N x D where each row is an example we wish to predict label for """
        num_test = X.shape[0]
        # lets make sure that the output type matches the input type
        Ypred = np.zeros(num_test, dtype = self.ytr.dtype)

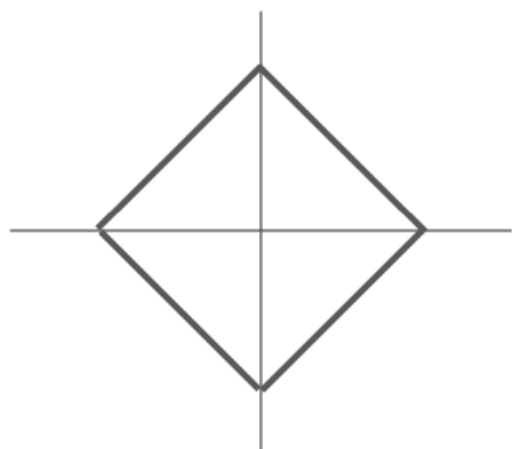
        # loop over all test rows
        for i in xrange(num_test):
            # find the nearest training image to the i'th test image
            # using the L1 distance (sum of absolute value differences)
            distances = np.sum(np.abs(self.Xtr - X[i,:]), axis = 1)
            min_index = np.argmin(distances) # get the index with smallest distance
            Ypred[i] = self.ytr[min_index] # predict the label of the nearest example

        return Ypred
```


K-Nearest Neighbors: Distance Metric

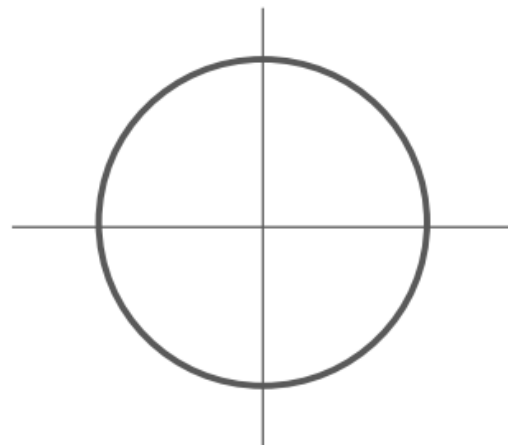
L1 (Manhattan) distance

$$d_1(I_1, I_2) = \sum_p |I_1^p - I_2^p|$$



L2 (Euclidean) distance

$$d_2(I_1, I_2) = \sqrt{\sum_p (I_1^p - I_2^p)^2}$$



k-Nearest Neighbor on images **never used**.

- Very slow at test time
- Distance metrics on pixels are not informative

Original



Boxed



Shifted



Tinted



(all 3 images have same L2 distance to the one on the left)

[Original image is](#)
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Reference

- Cs231n Stanford University Tutorial

<http://cs231n.stanford.edu/>

<https://www.youtube.com/watch?v=vT1JzLTH4G4>

- Canny edge detector

https://en.wikipedia.org/wiki/Canny_edge_detector