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Image Classification pipelineImage Classification

↳ A Core task in Computer Vision

⇒ Assume given set of discrete labels

Semantic Gap

⇒ 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 image.

① Nearest Neighbor Classifier

def train(images, labels):

⑥ # Machine learning,  
return model

→ Memorize all  
data & labels

def predict(model, test\_images):

⑥ # Use model to predict labels  
return test\_labels

→ Predict the label  
of the most  
similar training  
image

⇒ Distance Metric to Compare Images

① L1 distance:  $d_1(I_1, I_2) = \sum_p |I_1^p - I_2^p|$

{also called Manhattan distance}



## ② K-Nearest Neighbors

⇒ Instead of copying label from nearest neighbor, take majority vote from  $K$  closest points.

⇒ L2 (Euclidean) distance

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

Hyperparameters: Choices about the algorithm that we set rather than learn.

- Example:  $K$ , distance function.

⇒ K-Nearest Neighbor on images never used

↳ Very slow at test time

↳ Distance metrics on pixels are not informative.

## \* Linear Classification

### Parametric Approach

Image  
 $32 \times 32 \times 3$

→  $f(x, W)$

10 numbers  
giving class  
scores

Parameters  
on  
weights

$$f(x, W) = Wx + b$$

$$f(x, W) = \overset{b+}{W} x \rightarrow 3072 \times 1$$

$\downarrow$   
 $10 \times 3072$