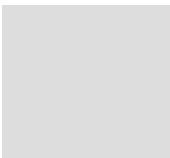


TRƯỜNG ĐẠI HỌC CÔNG NGHỆ THÔNG TIN
KHOA KHOA HỌC MÁY TÍNH

CS231. Nhập môn Thị giác máy tính

Machine Learning Image
Classification



Mai Tiến Dũng

Content

1. Introduction to image classification
2. Image classification with KNN
3. Image feature



1. Introduction to Image Classification

- What is image classification
- Challenges of image classification



What is image classification

- Process of taking an image or picture



Cat



98%



Use of image classification

Organize Photo albums
on smart devices



Augment Medical
professionals



Identify images around
Self-driving cars



Targets:

$y = 0$

$y = "cat"$



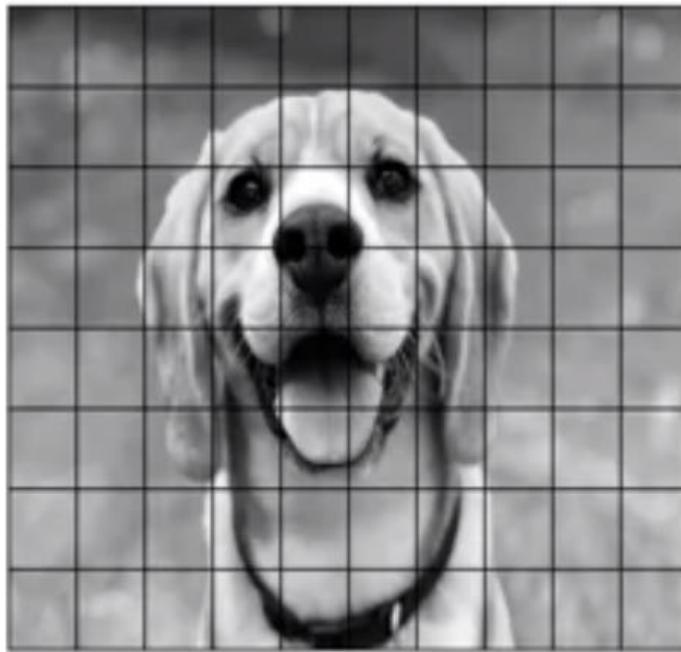
$y = 1$

$y = "dog"$





$$X = \begin{bmatrix} 2 & 3 & .. & 40 \\ 3 & 10 & .. & 50 \\ 2 & 3 & .. & 40 \\ 5 & 34 & .. & 24 \\ 3 & 10 & .. & 50 \\ 2 & 3 & .. & 40 \\ 5 & 34 & .. & 24 \\ 3 & 10 & .. & 50 \\ 5 & 34 & .. & 24 \end{bmatrix}$$



$$X = \begin{bmatrix} 2 & 3 & .. & 40 \\ 3 & 10 & .. & 50 \\ 5 & 34 & .. & 24 \end{bmatrix}$$

$x_1, y_1 = 1$



$x_3, y_3 = 0$



$x_5, y_5 = 0$



$x_2, y_2 = 1$



$x_4, y_4 = 1$



$x_6, y_6 = 0$

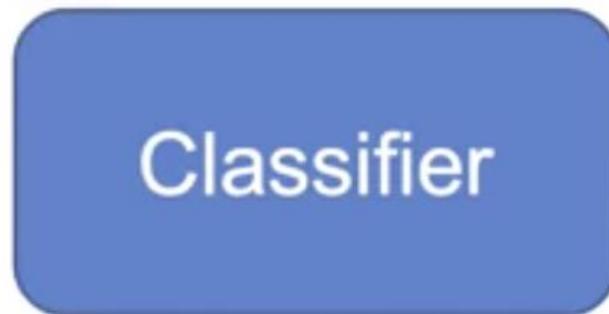


$x_7, y_7 = 0$



$y = 0$ 0
 $y = 1$ 1
 $y = 2$ 2
 $y = 3$ 3
 $y = 4$ 4
 $y = 5$ 5
 $y = 6$ 6
 $y = 7$ 7
 $y = 8$ 8
 $y = 9$ 9


```
def classify(X):
```



Cat

Challenges of image classification



Challenges: change in viewpoint



Challenges: Illumination



Challenges: deformation



Challenges: occlusion



Challenges: background clutter



Image Classification with KNN



Image Classification with KNN

What is k-NN?

- k-NN means k-Nearest Neighbor
- Simplest classification algorithm
- Uses the most common and nearest classes to find closes match

Image Classification with KNN

What is k-NN?

- k-NN means k-Nearest Neighbor
- Simplest classification algorithm
- Uses the most common and nearest classes to find closes match



Image Classification with KNN

Is it a Cat or a Dog?

$y = 0$



$y = 1$

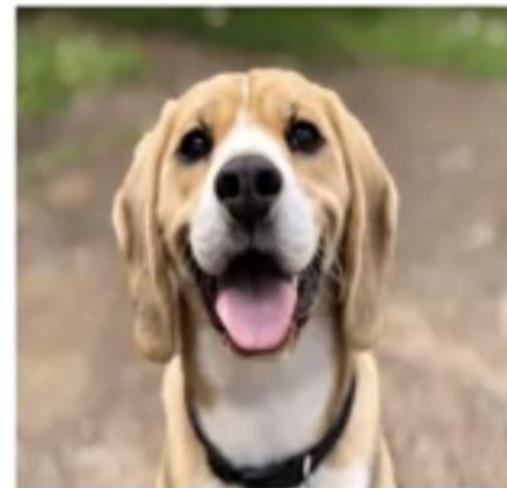


Image Classification with KNN

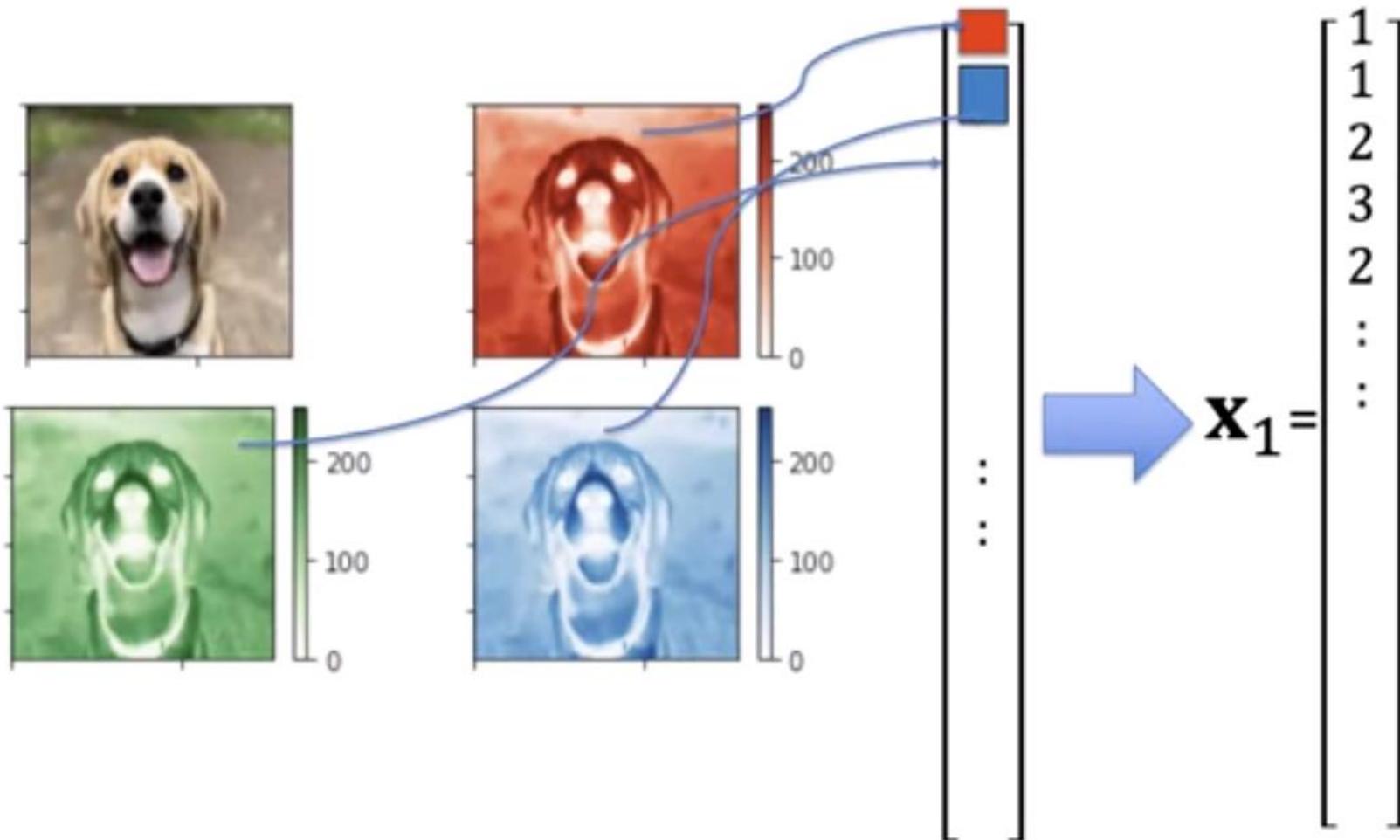


Image Classification with KNN

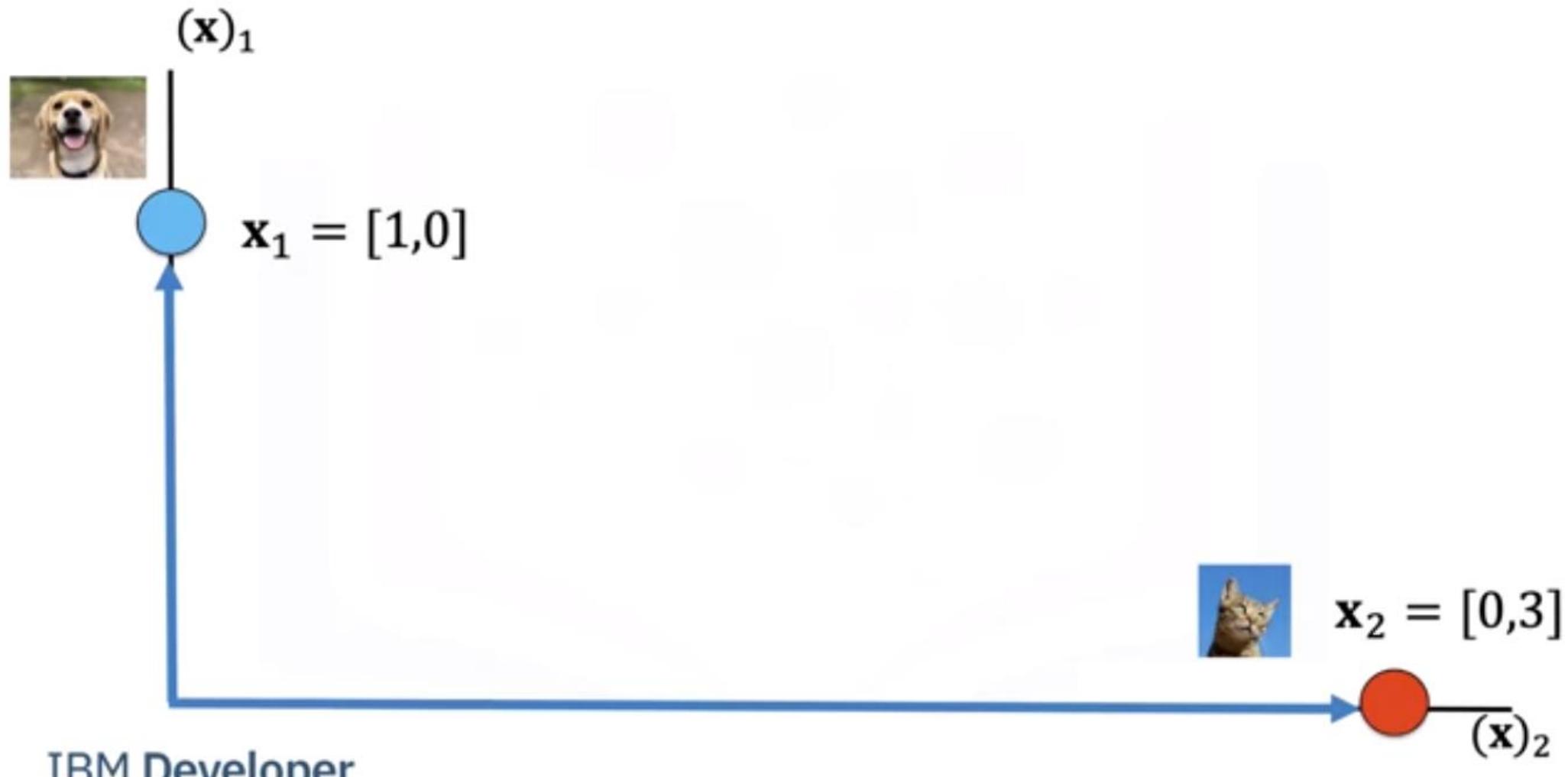


Image Classification with KNN

$(\mathbf{x})_1$



$\mathbf{x}_1 = [1, 0]$

$$d'(\mathbf{x}_1, \mathbf{x}_2) = \|\mathbf{x}_1 - \mathbf{x}_2\|$$

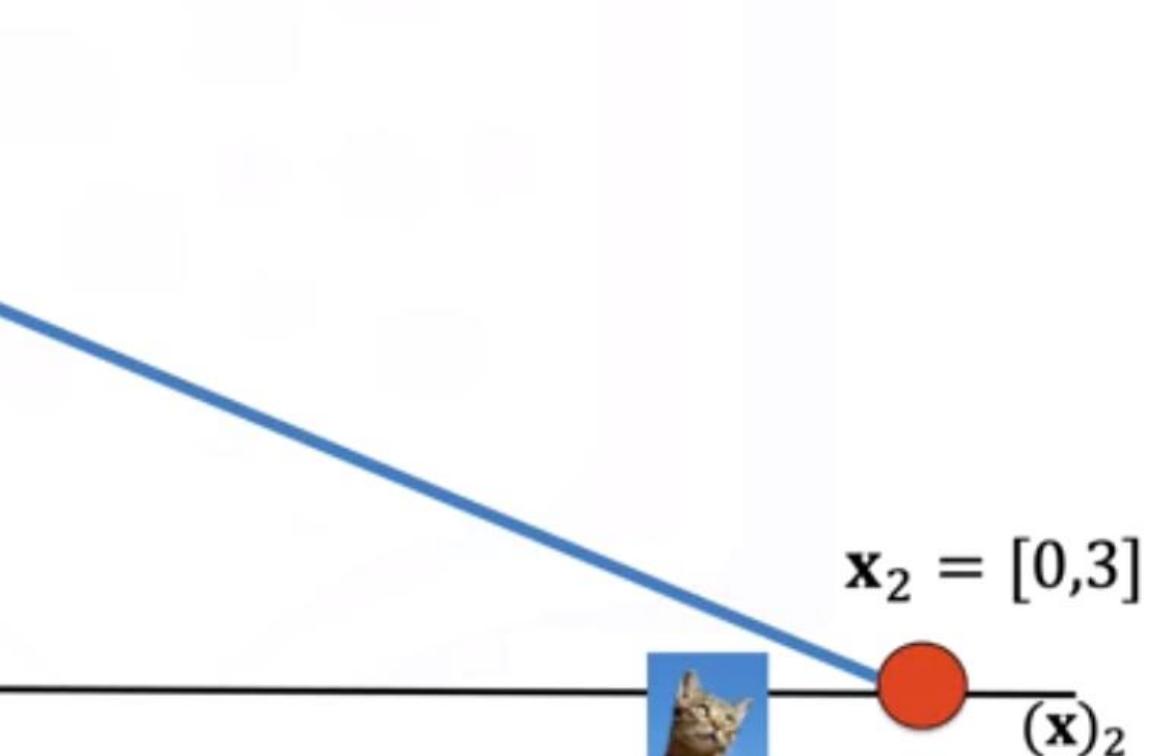


Image Classification with KNN

$(\mathbf{x})_1$



$$\mathbf{x}_1 = [1, 0]$$



the unknown sample \mathbf{x}_j



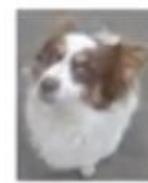
$$\mathbf{x}_2 = [0, 3]$$



$$(\mathbf{x})_2$$

the unknown sample \mathbf{x}_j

$(\mathbf{x})_1$



$\mathbf{x}_1 = [1, 0]$



$\mathbf{x}_2 = [0, 3]$

$(\mathbf{x})_2$

Image Classification with KNN

n	y	$d(\mathbf{x}_n, \mathbf{x}_j)$
1		
2		
3		
4		
5		
6		



the unknown sample

Image Classification with KNN

n		y	$d(\mathbf{x}_n, \mathbf{x}_j)$
1			
2			
3			
4			
5			
6			

$\hat{y} =$



the unknown sample

Image Classification with KNN

n	y	$d(\mathbf{x}_n, \mathbf{x}_j)$
1		 1.0
2		 1.5
3		 1.7
4		 2.0
5		 2.5
6		 3.0



the unknown sample

Image Classification with KNN

n		y	$d(\mathbf{x}_n, \mathbf{x}_j)$
1			1.0
2			1.5
3			1.7
4			2.0
5			2.5
6			3.0

$$\hat{y} = \textcolor{red}{\bullet}$$



the unknown sample

Image Classification with KNN

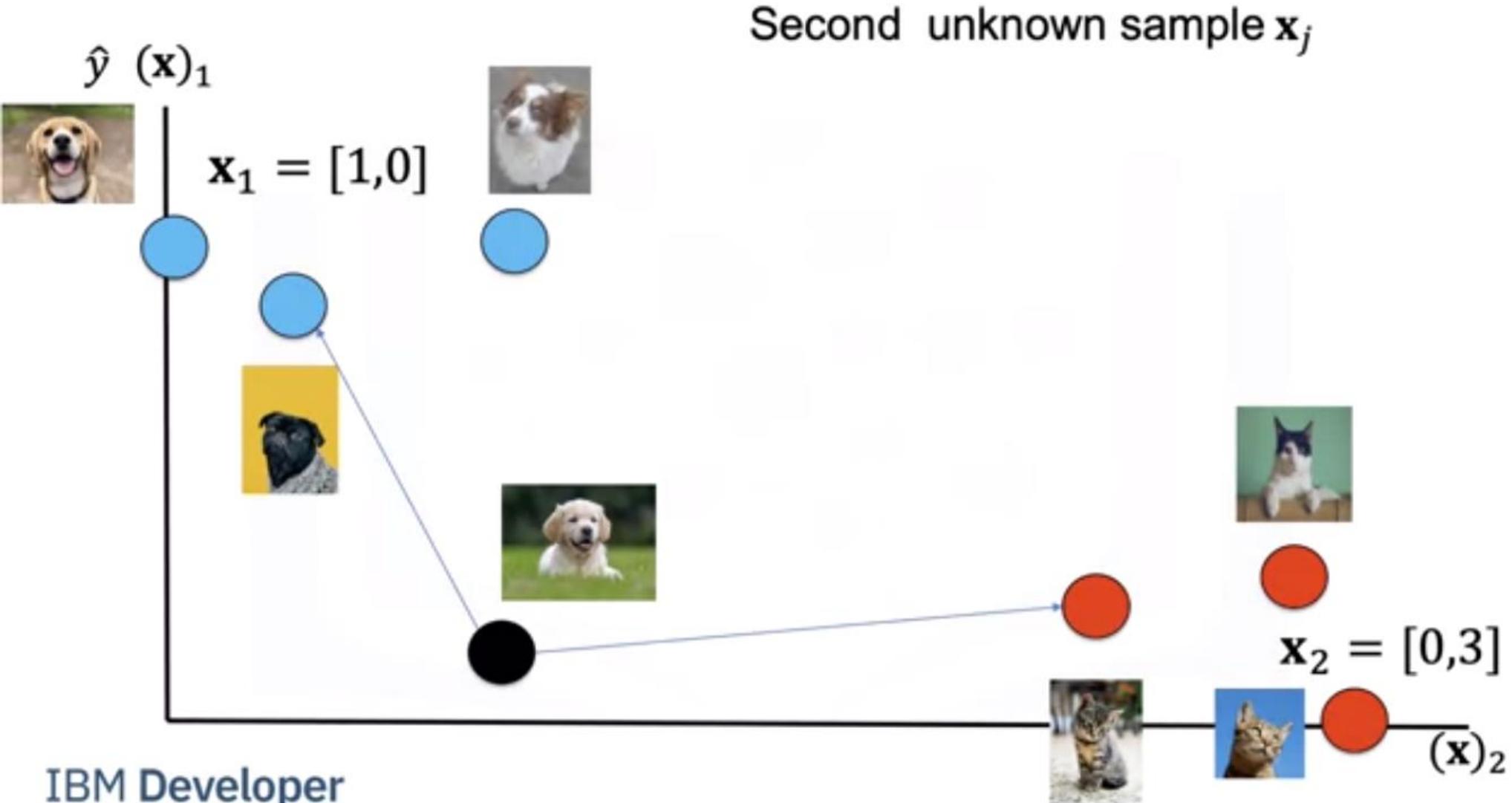


Image Classification with KNN

Training/Testing Sets

Data:



- Split dataset into:

Training/Testing Sets

Data:

- Split dataset into:
 - training set (70%), 
 - testing set (30%) 
- Build the model with a training set.

Image Classification with KNN

Training/Testing Sets

Data:

- Split dataset into:
 - training set (70%), 
 - testing set (30%) 
- Build the model with a training set.
- Use testing set to assess the performance.
- When we have completed testing our model we should use all the data

Image Classification with KNN

Accuracy

- What is the accuracy of a classifier?
 - the number of samples that have been predicted correctly divided by the total number of samples

Sample: n	1	2	3	4
y	1	0	1	0
\hat{y}	1	0	0	1
<i>Correct</i>				

Image Classification with KNN

Accuracy

- What is the accuracy of a classifier?
 - the number of samples that have been predicted correctly divided by the total number of samples

Samples: n	1	2	3	4
y	1	0	1	0
\hat{y}	1	0	0	1
<i>Correct</i>	1	1	0	0

$$= \frac{1}{4}(1+1+0+0) = 0.5$$

Image Classification with KNN

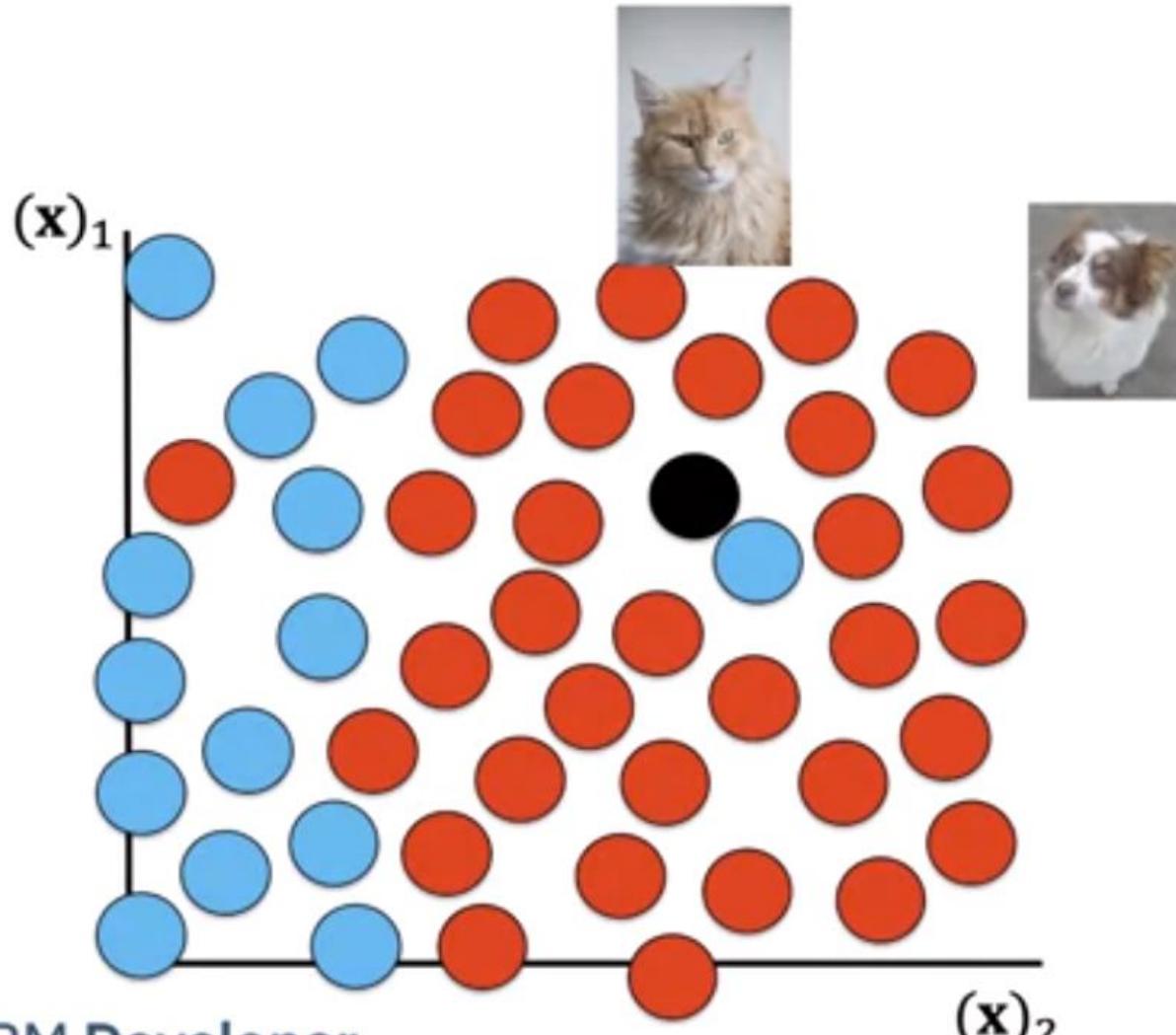


Image Classification with KNN

n	y	$d(\mathbf{x}_n, \mathbf{x}_j)$
1		 1.0
2		 1.5
3		 1.7
4		 2.0
5		 2.5
6		 3.0

$$\hat{y} =$$



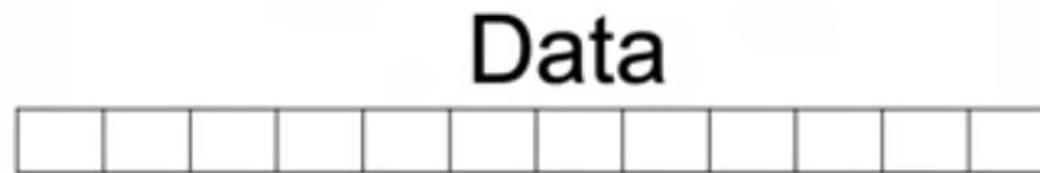
Image Classification with KNN

n	y	$d(\mathbf{x}_n, \mathbf{x}_l)$
1		 1.0
2		 1.5
3		 1.7
4		 2.0
5		 2.5
6		 3.0

$$\hat{y} = \textcolor{red}{\bullet}$$



Image Classification with KNN



Training

Validation

Test

Image Classification with KNN



Image Classification with KNN

Data

Training

Validation

Test

Hyperparameters: **K**

Image Classification with KNN

Data



Training

Validation

Test

Hyperparameters: **K**

Image Classification with KNN

Data

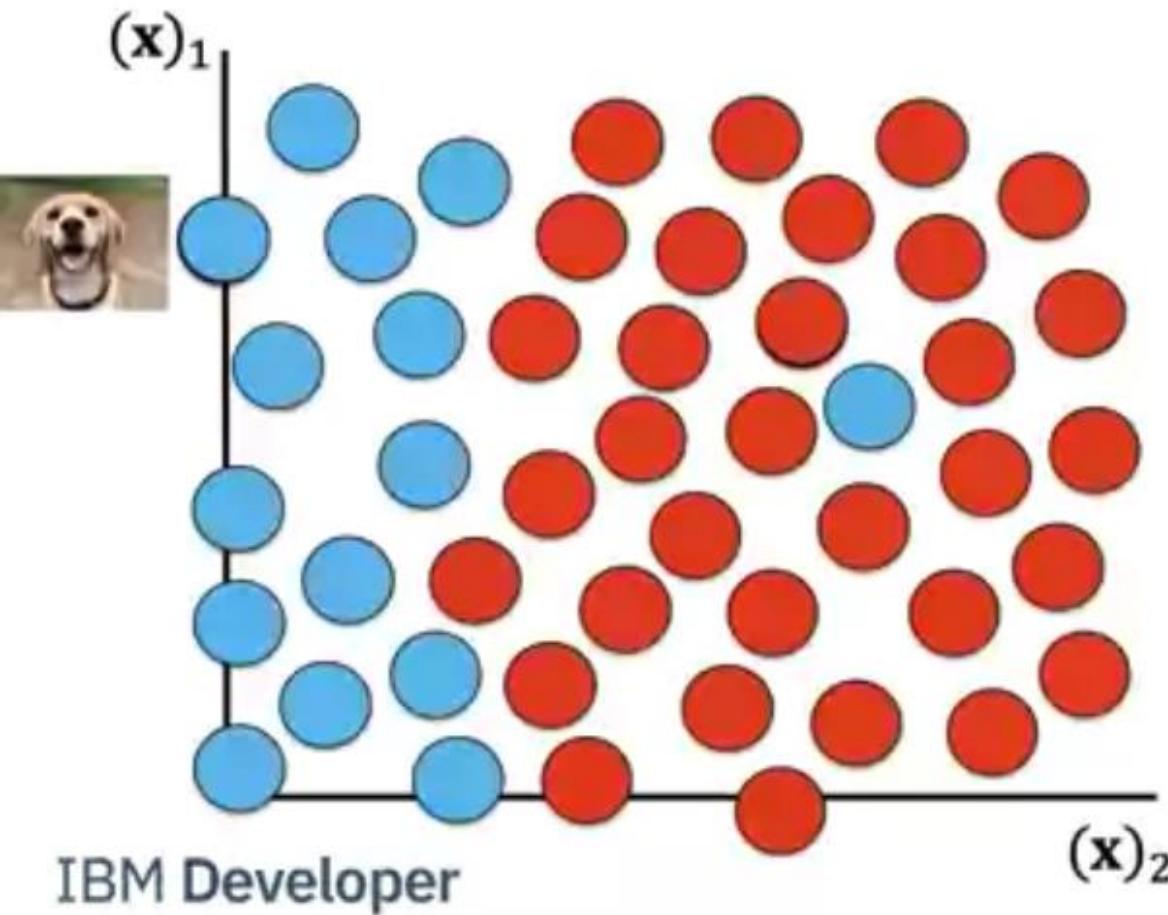
Training

Validation

Test

Hyperparameters: K

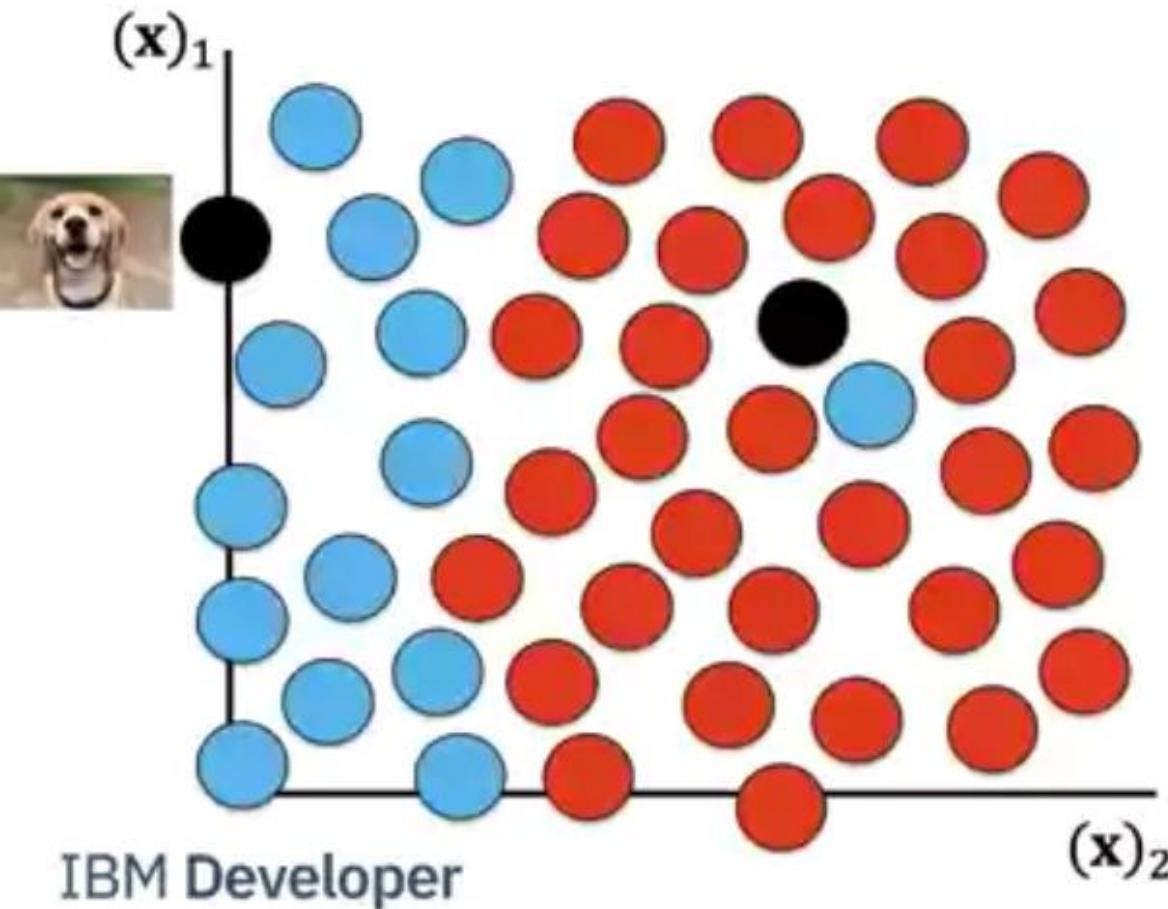
Image Classification with KNN



Validation Data

y	\hat{y}	Accuracy		
K=1				
K=3				

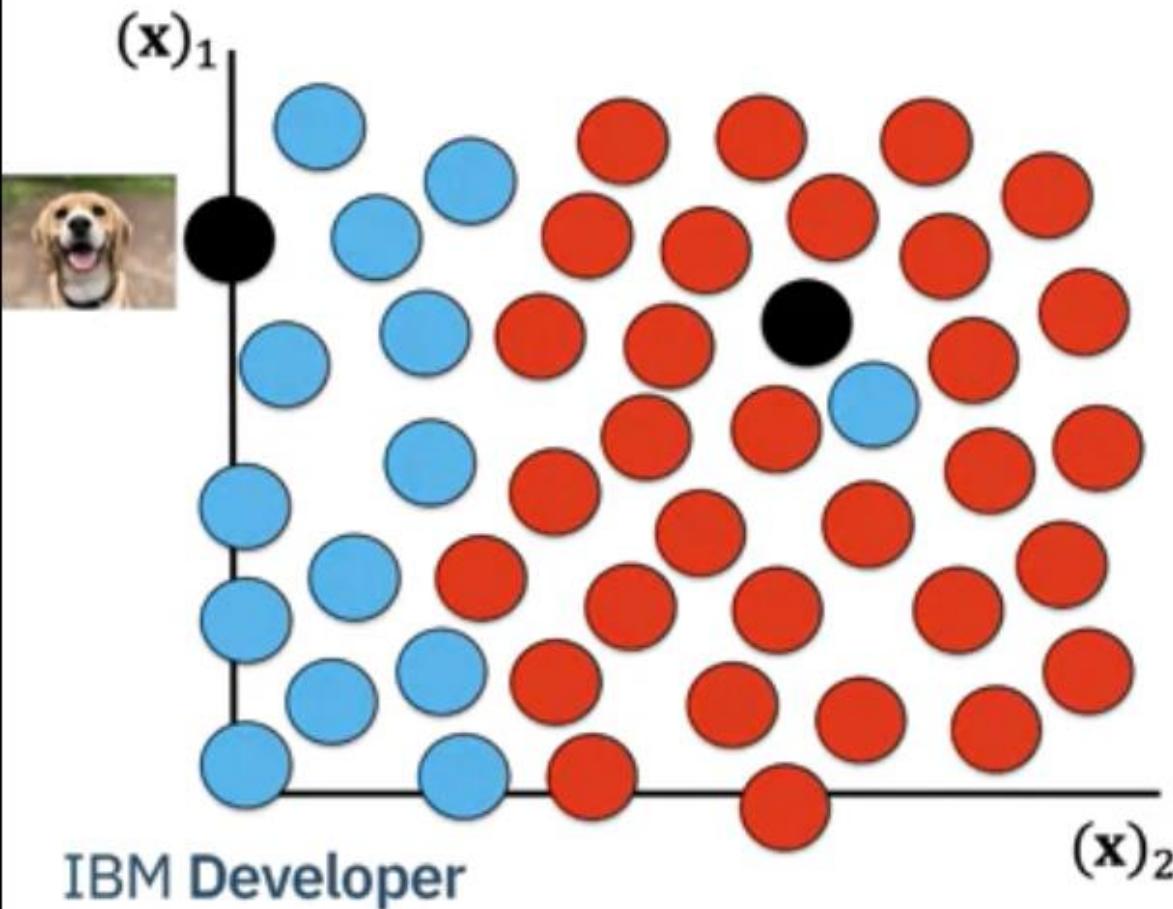
Image Classification with KNN



Validation Data

\hat{y}	y		Accuracy
K=1			
K=3			

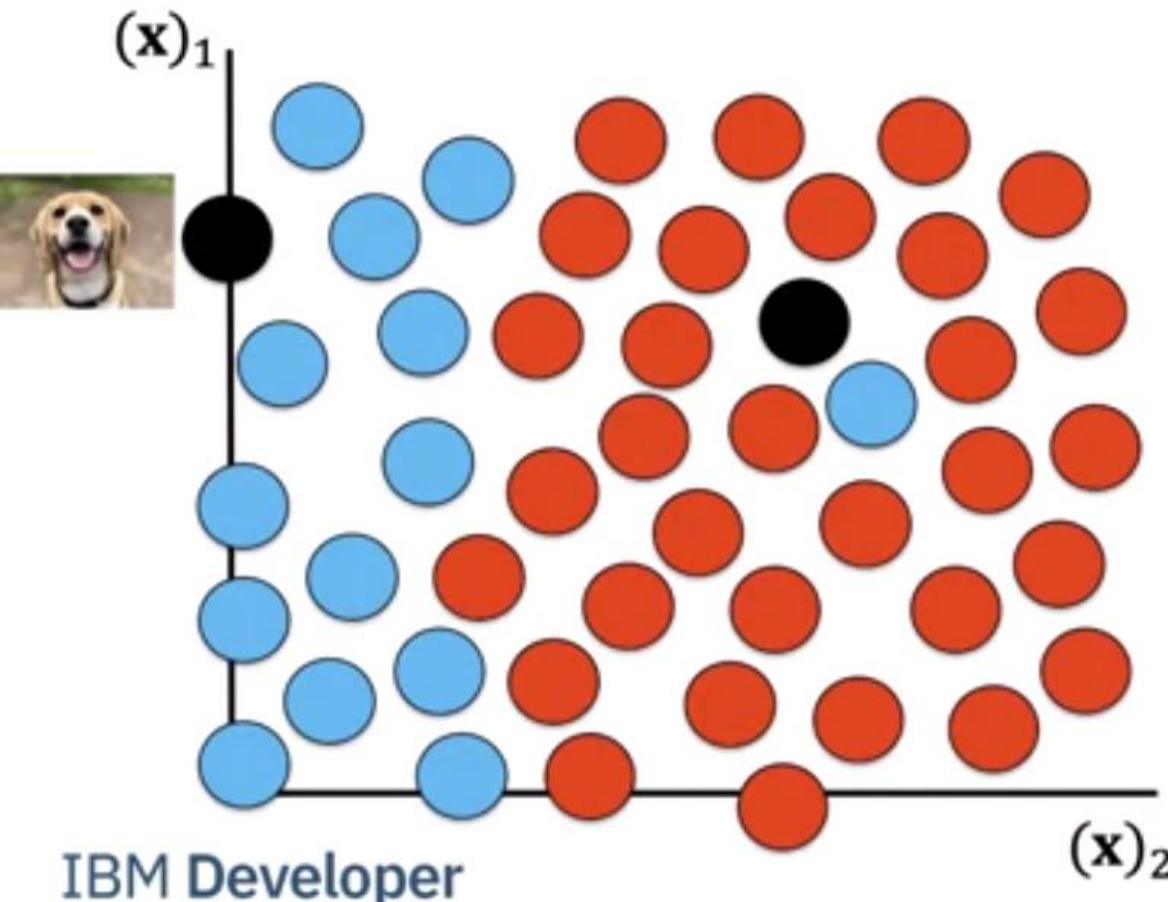
Image Classification with KNN



Validation Data

\hat{y}	y		Accuracy
K=1	Blue	Red	50%
K=3			

Image Classification with KNN



Validation Data

\hat{y}	y		Accuracy
K=1	Blue	Red	50%
K=3	Blue	Red	100%

Image Classification with KNN



n		y	$d(\mathbf{x}_n, \mathbf{x}_i)$
1			1.0
2			1.5
3			1.7
4			2.0
5			2.5
6			3.0

Image Classification with KNN

Multiclass :

$y = 0$



$y = 1$



$y = 2$



Image Classification with KNN

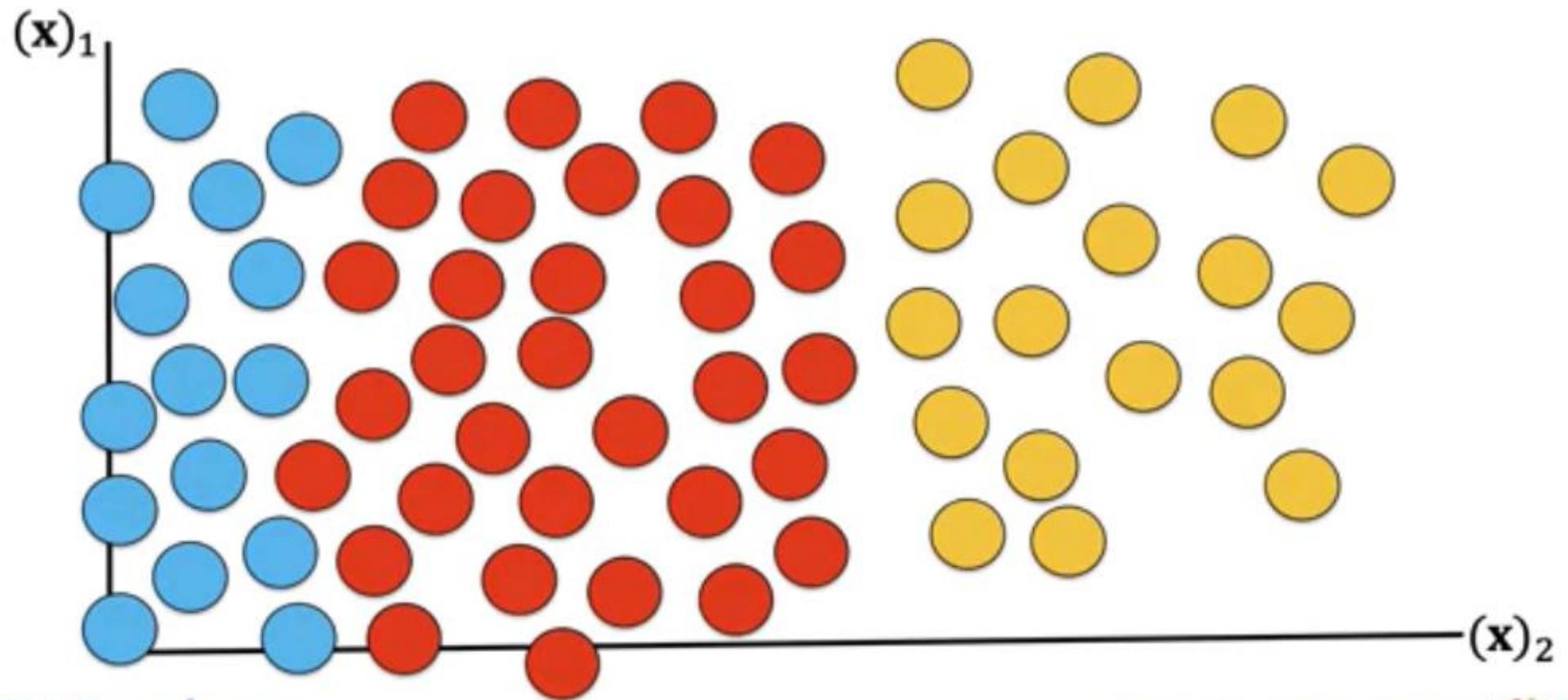
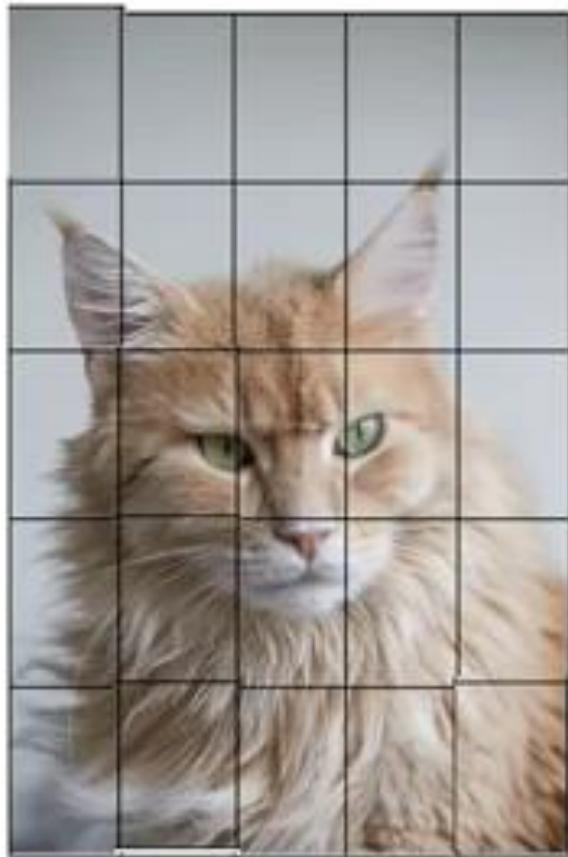


Image Features

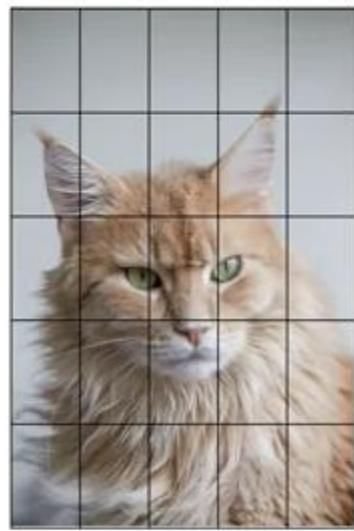


Image Features



$$\mathbf{x} = \begin{bmatrix} & \\ \vdots & \\ & \end{bmatrix}$$

Image Features

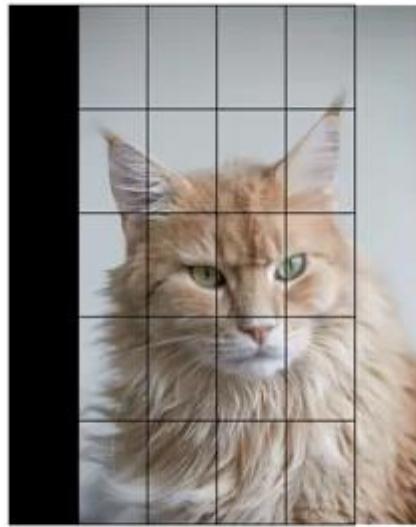


IBM Developer

$$\mathbf{x} = \begin{bmatrix} \cdot & \cdot & \cdot \\ \vdots & \vdots & \vdots \\ \cdot & \cdot & \cdot \end{bmatrix}$$



Image Features

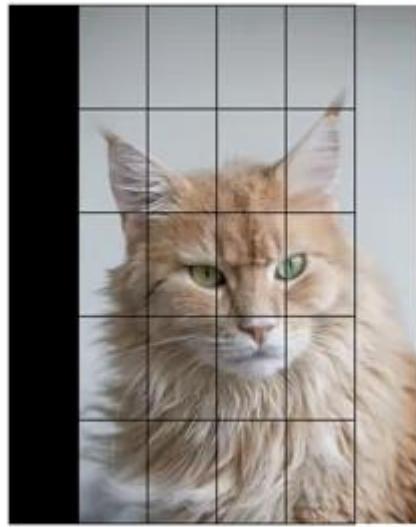


IBM Developer

$$\mathbf{x} = \begin{bmatrix} & \\ \vdots & \\ & \end{bmatrix}$$



Image Features



IBM Developer

$$\mathbf{x} = \begin{bmatrix} \vdots \\ \text{[Feature Vector]} \\ \vdots \\ \vdots \end{bmatrix}$$



Image Features

$$\mathbf{x} = \begin{bmatrix} \text{Small Image} \\ \vdots \\ \text{Large Image} \end{bmatrix}$$

$$\mathbf{x} = \begin{bmatrix} \text{Small Image} \\ \vdots \\ \text{Large Image} \end{bmatrix}$$



Image Features

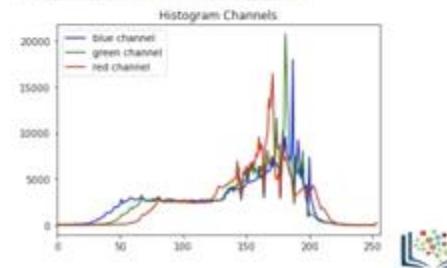
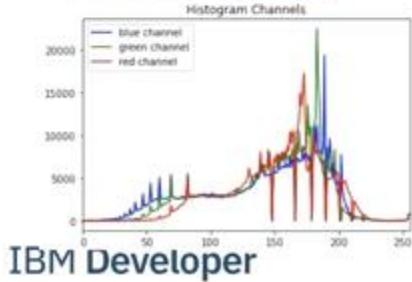


Image Features

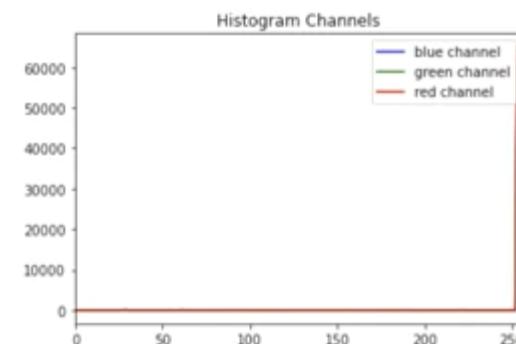
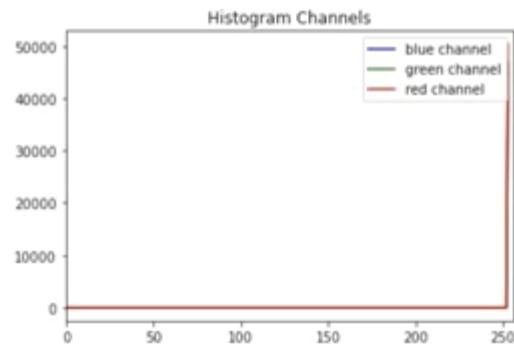


Image Features

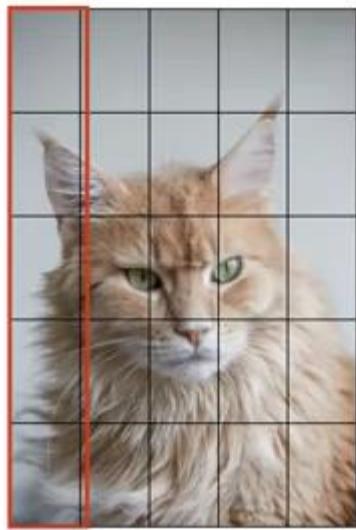


IBM Developer

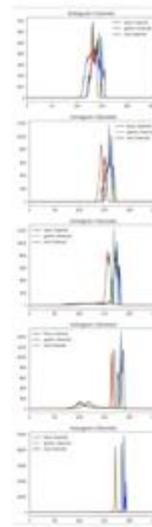
SKILLS NETWORK 



Image Features



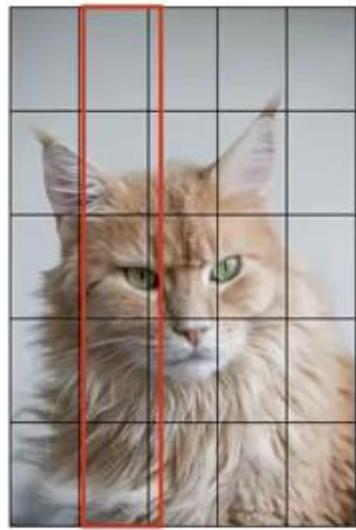
IBM Developer



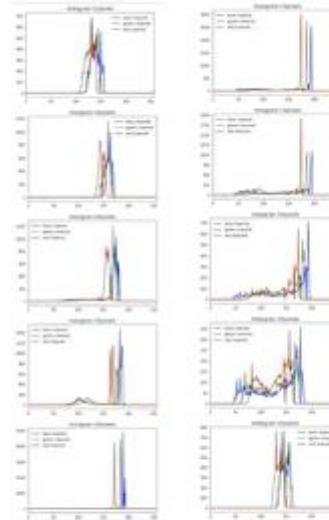
SKILLS NETWORK 



Image Features



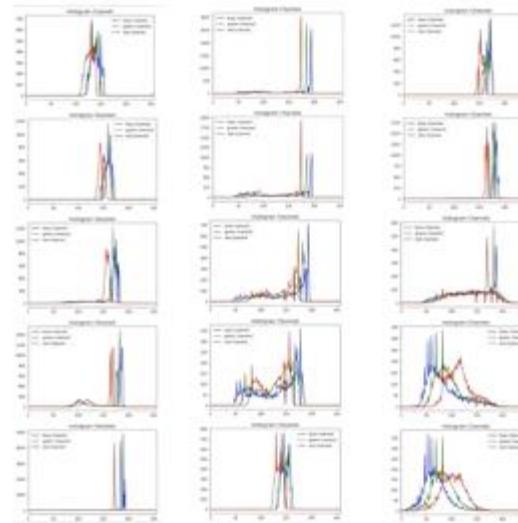
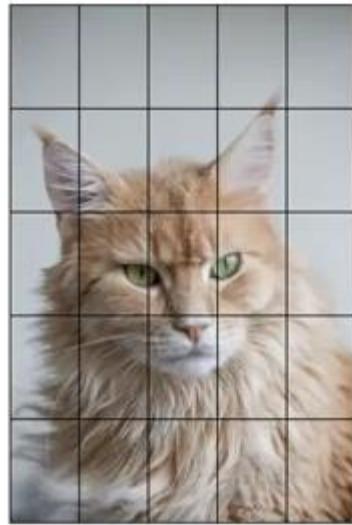
IBM Developer



SKILLS NETWORK



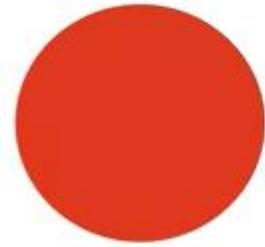
Image Features



IBM Developer

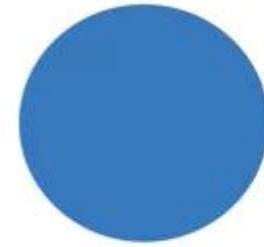
SKILLS NETWORK

Image Features



$$\mathbf{X} = \begin{bmatrix} 255 \\ 255 \\ \vdots \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

IBM Developer

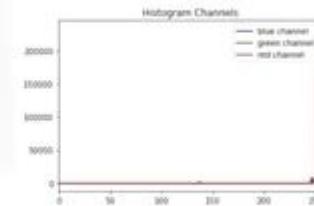
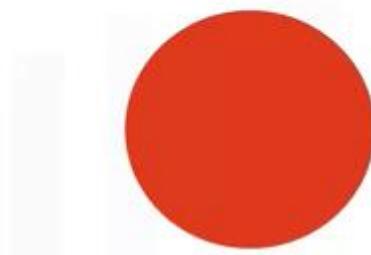


$$\mathbf{X} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ 255 \\ 255 \\ 0 \end{bmatrix}$$

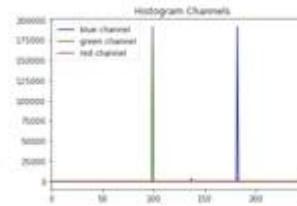
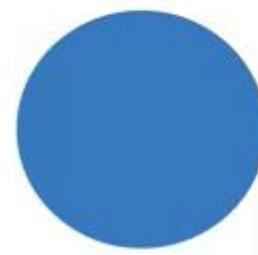
SKILLS NETWORK 



Image Features



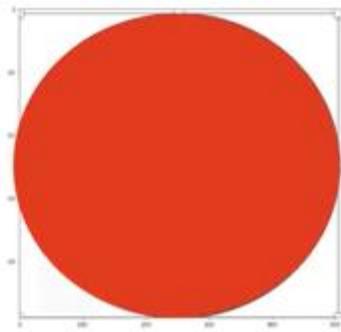
IBM Developer



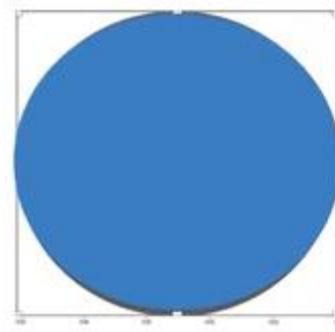
SKILLS NETWORK 



Image Features



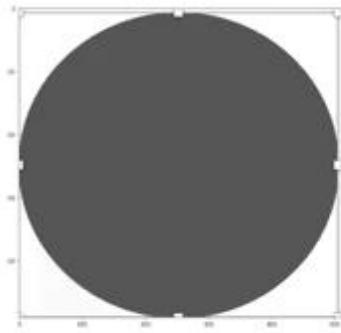
IBM Developer



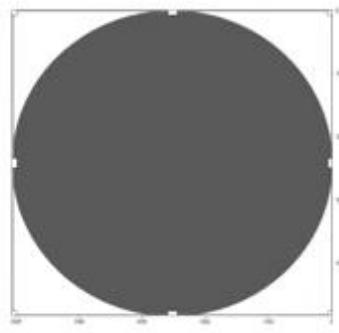
SKILLS NETWORK 



Image Features



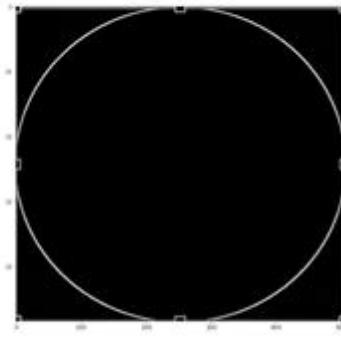
IBM Developer



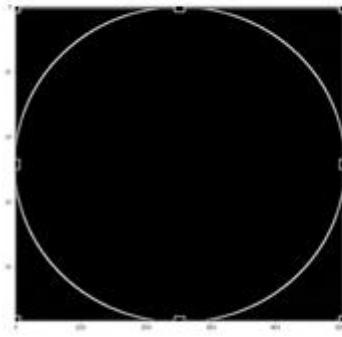
SKILLS NETWORK 



Image Features



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SKILLS NETWORK 



Image Features

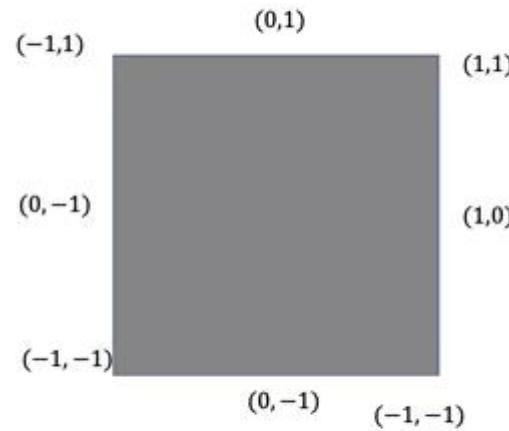
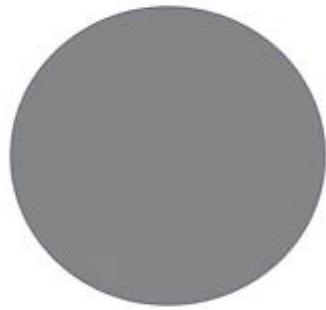
What is H.O.G.

- H.O.G. stands for Histogram of Oriented Gradients
- H.O.G uses the gradient orientation of the localized regions of an image
- H.O.G. generates a histogram for each localized region

IBM Developer

SKILLS NETWORK 

Image Features

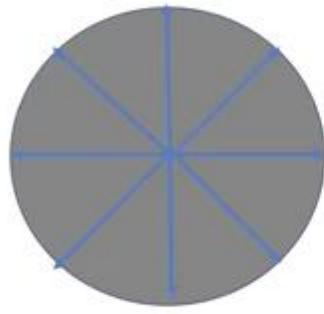


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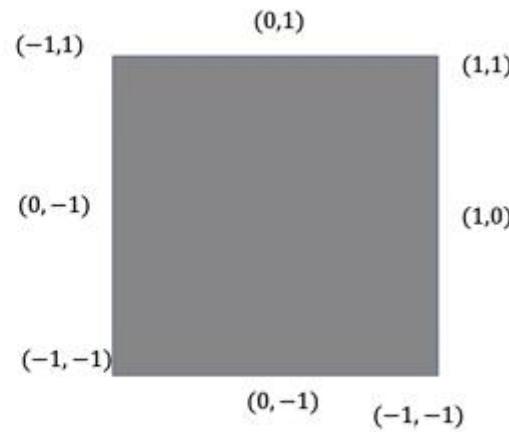
SKILLS NETWORK 



Image Features



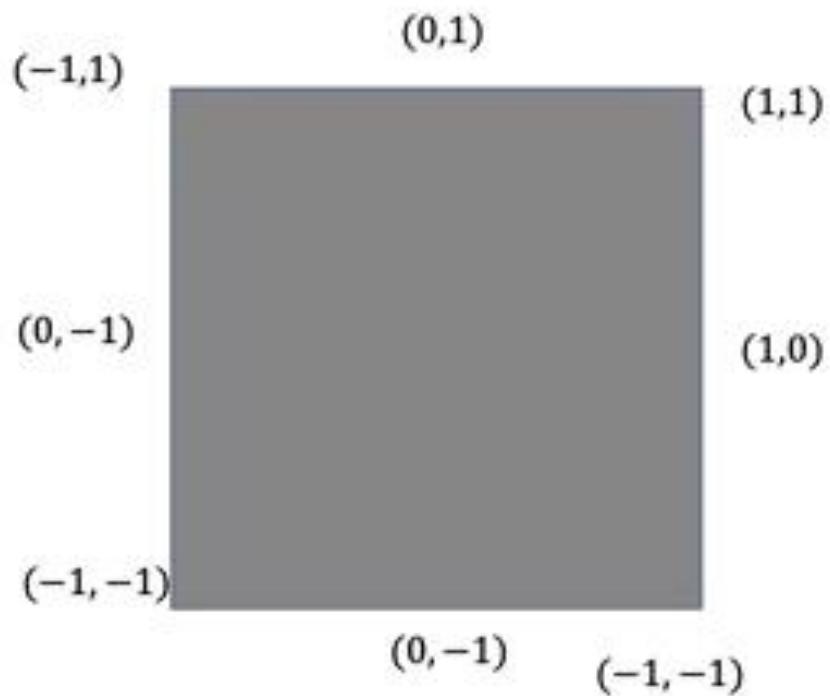
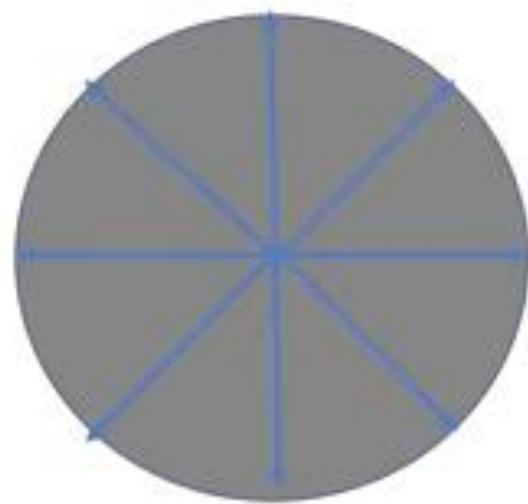
IBM Developer



SKILLS NETWORK 

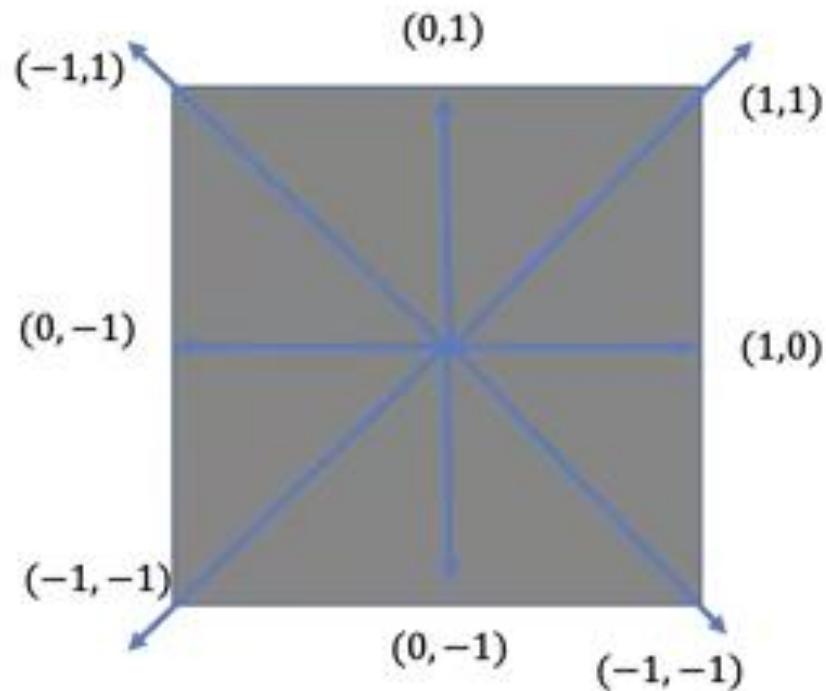
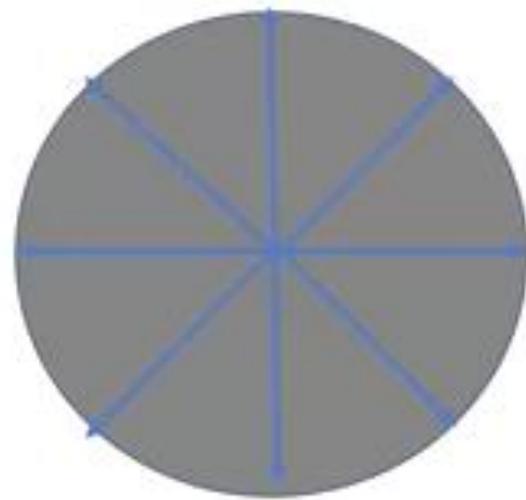


Image Features



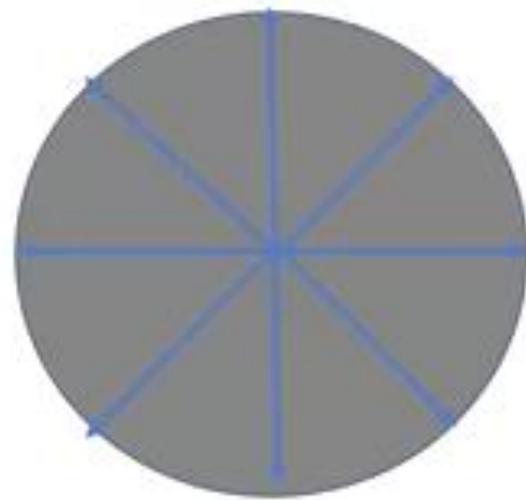
	1						
θ	0	45	90	135	180	225	270

Image Features

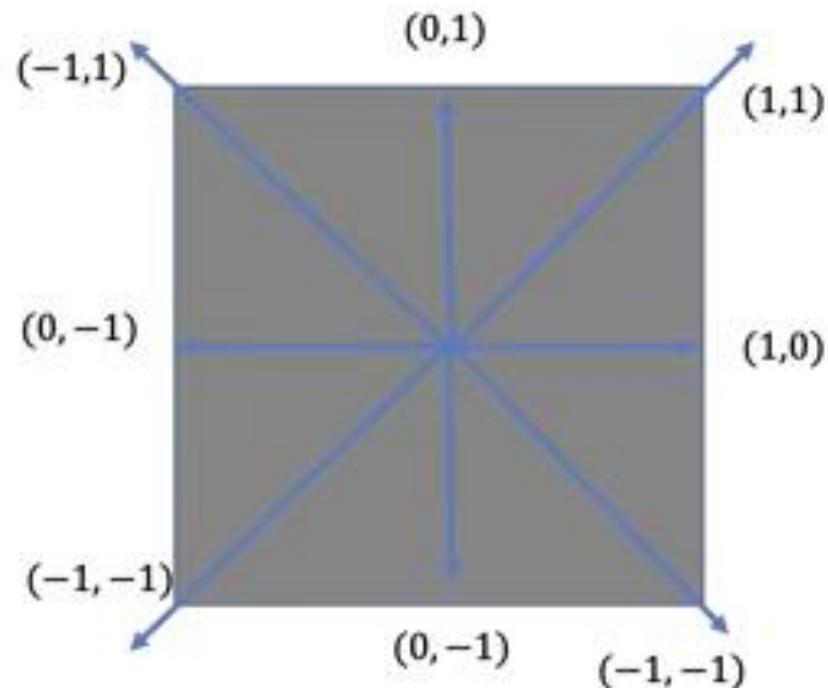


	1	1	1	1	1	1	1
θ	0	45	90	135	180	225	270

Image Features



	1						
θ	0	45	90	135	180	225	270



	1	$\sqrt{2}$	1	$\sqrt{2}$	1	$\sqrt{2}$	1
θ	0	45	90	135	180	225	270

Image Features

H.O.G.



Image Features

H.O.G.



$$\|G\| = \sqrt{G_x^2 + G_y^2}$$

→

$$\theta = \arctan\left(\frac{G_y}{G_x}\right)$$



Image Features

H.O.G.



$$\|G\| = \sqrt{G_x^2 + G_y^2}$$

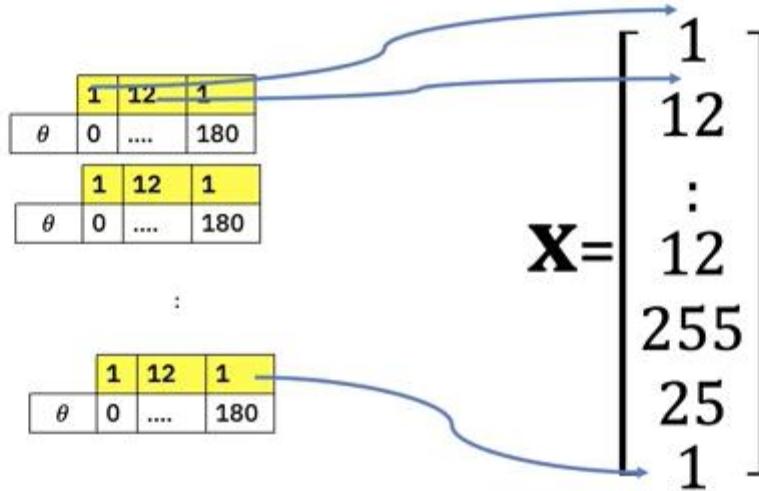
→

$$\theta = \arctan\left(\frac{G_y}{G_x}\right)$$



1	12	1	
θ	0	180
1	12	1	
⋮			
θ	0	180

Image Features



IBM Developer

SKILLS NETWORK 

Tài liệu tham khảo

- <https://www.coursera.org/learn/introduction-computer-vision-watson-opencv>



Thực hành



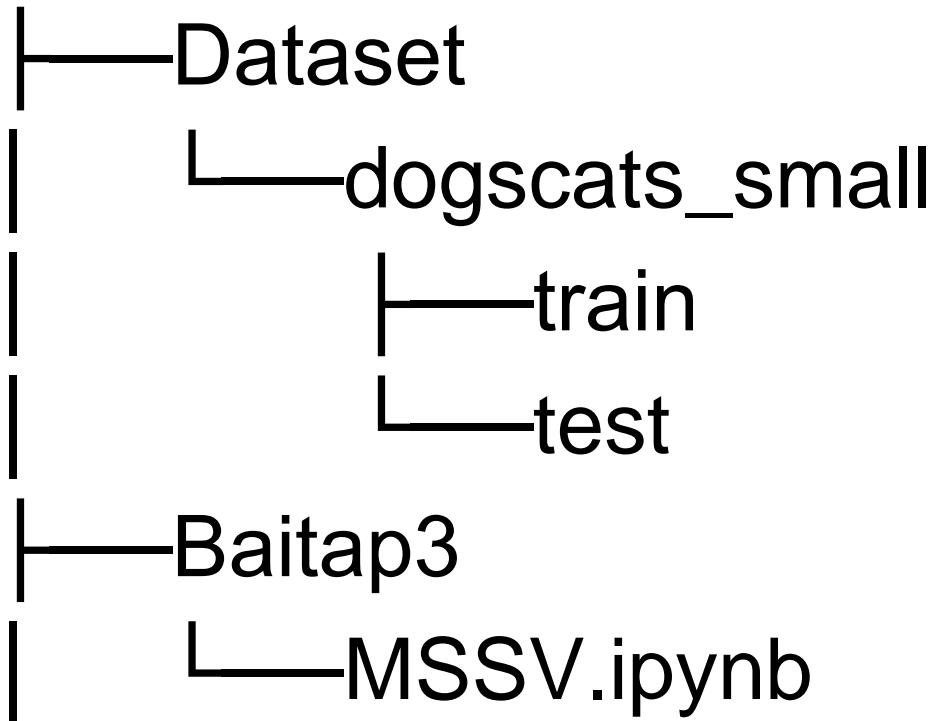
Yêu cầu

- Phân lớp ảnh: dogs vs cats
- Phương pháp: KNN
- Features: histogram
- Color: grayscale



Project structure

CS231.



Tính histogram cho ảnh

#Hàm tính histogram

```
def TinhHist(pathfilename):  
    img = cv.imread(pathfilename, 0)  
    hist = cv.calcHist([img], [0], None,  
                      [256], [0, 256])  
    size = img.shape[0]*img.shape[1]  
    hist = hist / size  
    return hist
```

Code: gợi ý

```
def ReadData(path_to_files):  
    for file_name in os.listdir(path_to_files):  
        # Lấy label của ảnh  
        # -> Thêm vào danh sách labels  
        # Tính đặc trưng của ảnh  
        # -> Thêm vào danh sách features  
    return features, labels
```



Code: gợi ý

- `os.listdir(path_to_files)` → đọc tất cả file trong thư mục `path_to_files`
- `Sorted (list_files)` → sắp xếp danh sách `list_files` theo thứ tự tăng dần



Code: gợi ý

```
from sklearn.neighbors import KNeighborsClassifier  
  
---  
label = imagePath.split(os.path.sep)[-1].split(".") [0]  
  
----  
features = np.array(features)  
labels = np.array(labels)  
  
---  
model = KNeighborsClassifier(n_neighbors=1)  
model.fit(features, labels)  
acc = model.score(test_features, test_labels)
```

