# Machine Learning Practice 3

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1 Logistic Regression

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Practice

## **Implement**

## Syntax (import)

from sklearn.linear\_model import LogisticRegression

## **Hyperparameters**

- ullet max\_iter o the maximum iterations for training model
- random\_state

#### **Examples**

- ▶ from sklearn.linear\_model import LogisticRegression
- ▷ clf = LogisticRegression(max\_iter=20000, random\_state=1)

# Implement (next)

#### **Methods**

fit

## Examples

▷ clf.fit(X\_train, y\_train)

#### **Methods**

predict

#### Examples

 $\triangleright$  y\_pred = clf.predict(X\_test)

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## Introduction



Figure: lena.jpg

#### (Grayscale Image)





[[255, 255, 255, ..., 0, 128, 226], [124, 21, 243, ..., 12, 123, 253], : : : [100, 20, 108, ..., 15, 232, 123]] 2D Matrix (256, 256)

256 x 256

#### (Color Image)

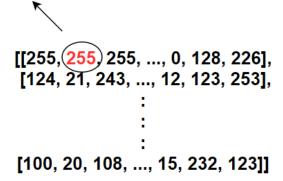




3D Matrix (256, 256, 3)

256 x 256

#### Each element is an integer in range [0, 255]



# How to make the computer learn?

- Learning on all pixels.
- Feature extraction.
- . . .

## Learning on all pixels

**Idea**: Convert 2D matrix to 1D matrix by concatenating.

Example:

## Learning on all pixels

#### Feature extraction

There are many feature extraction methods:

- Color-based (e.g., RGB histogram)
- Texture-based (e.g., Gabor filter)
- Shape-based (e.g., Edge histogram, H.O.G)
- ...

# Histogram of Oriented Gradients (H.O.G)

Input image



Histogram of Oriented Gradients



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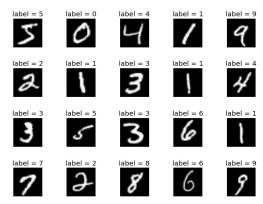
Logistic Regression

2 Image Classification

3 Practice

#### MNIST dataset

- MNIST is a large dataset of handwritten digits (0 to 9)
- Each image has the width 28 and the height 28 (28x28 pixels)
- 60.000 training images; 10.000 testing images.



# Loading MNIST dataset

#### **Loading data**

- ▶ from keras.datasets import mnist
- ▷ (X\_train, y\_train), (X\_test, y\_test) = mnist.load\_data()

# Preprocessing (Learning on all pixels)

- b def convert\_to\_1D (image):
   return np.reshape(image, image.shape[0] \* image.shape[1])
- ▶ X\_train\_new = []
- ▷ X\_test\_new = [ ]

## Training model

## **Training**

▶ from sklearn.linear\_model import LogisticRegression clf = LogisticRegression(max\_iter=100, random\_state=1, n\_jobs=2)

▷ clf.fit(X\_train\_new, y\_train)

# Testing and evaluating model

## **Testing**

## **Evaluating**

from sklearn.metrics import accuracy\_score, classification\_report

```
print('Accuracy:', accuracy_score(y_test, y_pred))
```

- ... Accuracy: 0.9255
- print(classification\_report(y\_test, y\_pred))

# Preprocessing (H.O.G)

- ▶ from skimage.feature import hog
- b def calculate\_hog (image):
   return hog(image, orientations=9, pixels\_per\_cell=(4, 4),
   cells\_per\_block=(2, 2), block\_norm='L2')
- ▷ X\_train\_new\_ = [ ]

# Training, testing and evaluating model

## **Training**

- $\triangleright$  clf\_ = LogisticRegression(max\_iter=100, random\_state=1, n\_jobs=2)
- ▷ clf\_.fit(X\_train\_new\_, y\_train)

## **Testing**

 $\triangleright$  y\_pred\_ = clf\_.predict(X\_test\_new\_)

## **Evaluating**

- print('Accuracy:', accuracy\_score(y\_test, y\_pred\_))
  - $\dots$  Accuracy: 0.9838 # improve 5.83% accuraccy
- print(classification\_report(y\_test, y\_pred\_))