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# COMP3055

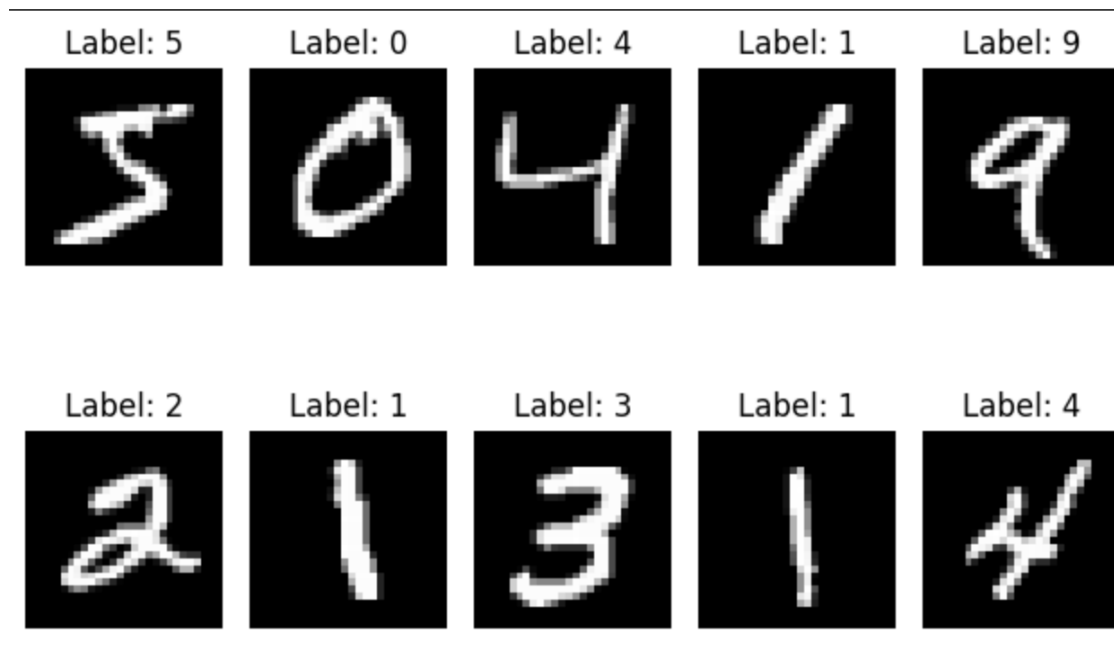
# Machine Learning

## General Idea for Lab4

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# MNIST dataset

- A handwritten digit dataset (number 0 to 9)
- Total 70000 images
- Each image is 28x28 grayscale image (pixel value 0 represents black, 255 represents white) , flatten into a 784-dimensional vector



# MNIST dataset

- An example of image 5

Label: 5



```
[0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0]
[0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0]
[0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0]
[0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0]
[0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0]
[ 0 0 0 0 0 0 0 0 0 0 0 0 3 18 18 18 126 136 175 26 166
255 247 127 0 0 0 0] [ 0 0 0 0 0 0 0 0 0 30 36 94 154 170
253 253 253 253 253 225 172 253 242 195 64 0 0 0] [ 0 0
0 0 0 0 49 238 253 253 253 253 253 253 253 253 251 93
82 82 56 39 0 0 0 0] [ 0 0 0 0 0 0 0 18 219 253 253 253
253 253 198 182 247 241 0 0 0 0 0 0 0 0] [ 0 0 0 0 0
0 0 0 80 156 107 253 253 205 11 0 43 154 0 0 0 0 0 0
0 0] [ 0 0 0 0 0 0 0 0 14 1 154 253 90 0 0 0 0 0 0
0 0 0 0 0] [ 0 0 0 0 0 0 0 0 0 0 139 253 190 2 0 0 0
0 0 0 0 0 0 0 0] [ 0 0 0 0 0 0 0 0 0 0 11 190 253
70 0 0 0 0 0 0 0 0 0 0 0 0] [ 0 0 0 0 0 0 0 0 0 0 0
35 241 225 160 108 1 0 0 0 0 0 0 0 0] [ 0 0 0 0 0 0 0
0 0 0 0 0 81 240 253 253 119 25 0 0 0 0 0 0 0] [ 0
0 0 0 0 0 0 0 0 0 0 45 186 253 253 150 27 0 0 0 0
0 0 0] [ 0 0 0 0 0 0 0 0 0 0 0 0 16 93 252 253 187
0 0 0 0 0 0 0] [ 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 249
253 249 64 0 0 0 0 0 0] [ 0 0 0 0 0 0 0 0 0 0 0 0 0
46 130 183 253 253 207 2 0 0 0 0 0 0] [ 0 0 0 0 0 0 0
0 0 0 0 39 148 229 253 253 253 250 182 0 0 0 0 0 0] [
0 0 0 0 0 0 0 0 0 24 114 221 253 253 253 253 201 78 0 0
0 0 0 0 0 0] [ 0 0 0 0 0 0 0 0 23 66 213 253 253 253
253 198 81 2 0 0 0 0 0 0 0] [ 0 0 0 0 0 0 18 171
219 253 253 253 253 195 80 9 0 0 0 0 0 0 0 0 0 0] [ 0
0 0 0 55 172 226 253 253 253 253 244 133 11 0 0 0 0 0
0 0 0 0 0 0] [ 0 0 0 0 136 253 253 253 212 135 132 16 0
0 0 0 0 0 0 0 0 0 0 0 0] [0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0] [0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0] [0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0]
```

# Naïve Bayesian Classifier

- What about using Naïve Bayesian Classifier for our handwritten digit recognition problem?
  - Each pixel is an  $x_i$ . There will be 784  $x$ 's.
  - Digit label is  $d_k$ . Note there will be 10 possible  $d$ 's.
  - $P(d_k)$  can be calculated by counting number of training images for the digit, divided to total number of training images.
  - $P(x_i/d_k)$  can be calculated by counting number of images for a given digit, given pixel position, and given an intensity value, divided by number of training images with that digit.

# Naïve Bayesian Classifier

$P(x_i/d_k)$  can be calculated by counting number of images for a given digit, given pixel position, and given an intensity value, divided by number of training images with that digit.

For example,

$$P(x_1 = 255|d = 0) = \frac{\text{Number of images whose first pixel value is 255 and contain digit 0}}{\text{Total number of images contain digit 0}}$$

# Naïve Bayesian Classifier

- For a given input image  $X$  and given digit label  $d_k$ , calculate  $P(d_k)$  and all  $P(x_i/d_k)$
- For each digit label  $d_k$ , calculate  

$$P(d_k|X) = P(d_k)P(x_1 = 0|d_k)P(x_2 = 255|d_k) \dots P(x_{784} = 0|d_k)$$
- Choose the digit label  $k$  that give the max value according to above calculation.

0	255	0	0	0	0	0	0	0	0
0	0	0	0	255	255	0	0	0	0
0	0	0	255	0	0	255	0	0	0
0	0	0	0	0	0	255	0	0	0
0	0	0	0	0	255	0	0	0	0
0	0	0	0	255	0	0	0	0	0
0	0	0	255	0	0	0	0	0	0
0	0	0	255	0	0	0	0	0	0
0	0	0	255	0	0	0	0	0	0
0	0	0	255	255	255	255	255	255	0

Input image X