

# deep learning hw #1

<b>Problem1</b>	<b>1</b>
Network structure	1
Hyperparameters	1
Initialize	2
Latent space visualization	3
Confusion matrix	4
<b>Problem2</b>	<b>5</b>
Image preprocess	5
Network structure & Hyper parameter	5
setting 1	5
setting 2	6
setting 3	6
setting 4	7
Visualized example	8
Accuracy	11

## Problem1

### Network structure

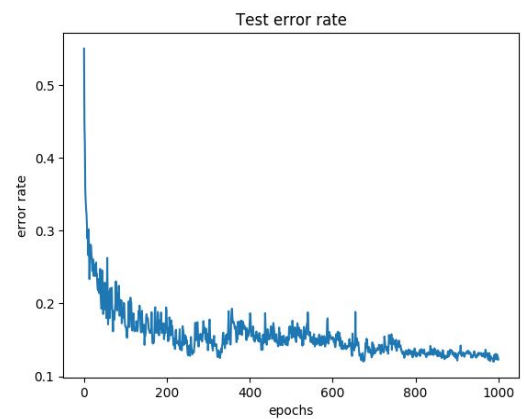
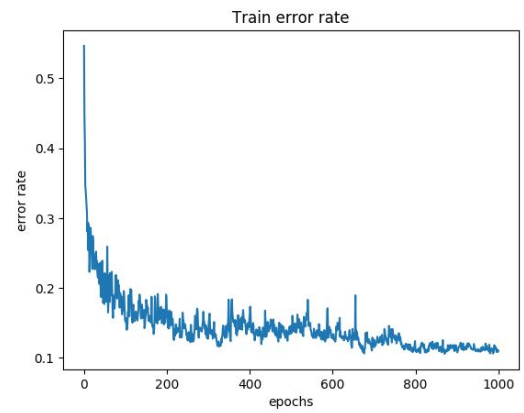
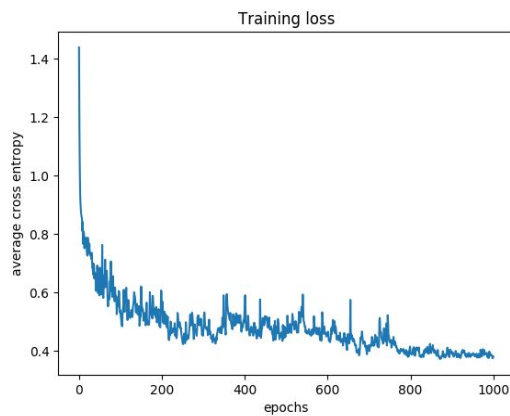
- input (28x28)
- fully connect (60)
- sigmoid (60)
- fully connect (2)
- fully connect (10)
- softmax output (10)

### Hyperparameters

- learning rate (0.4)
- batch size (16)
- epochs (1000)

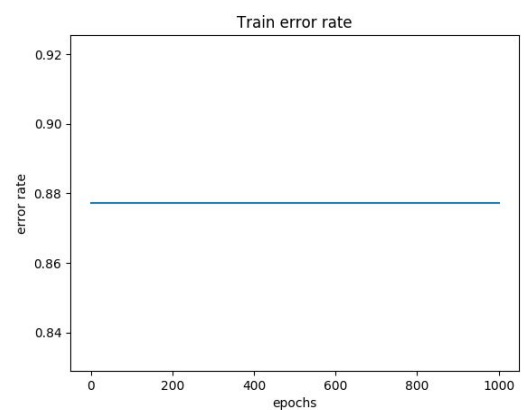
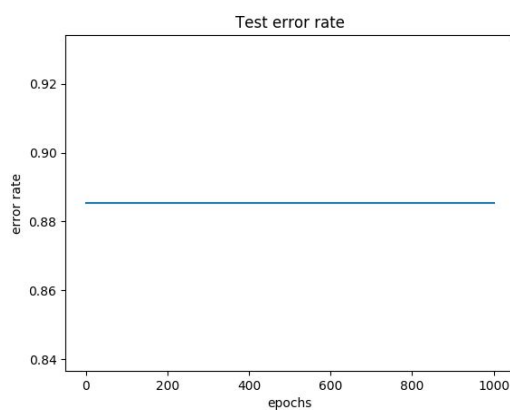
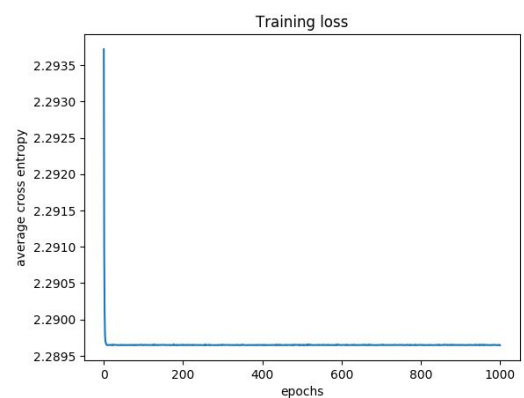
## Initialize

- random init weights



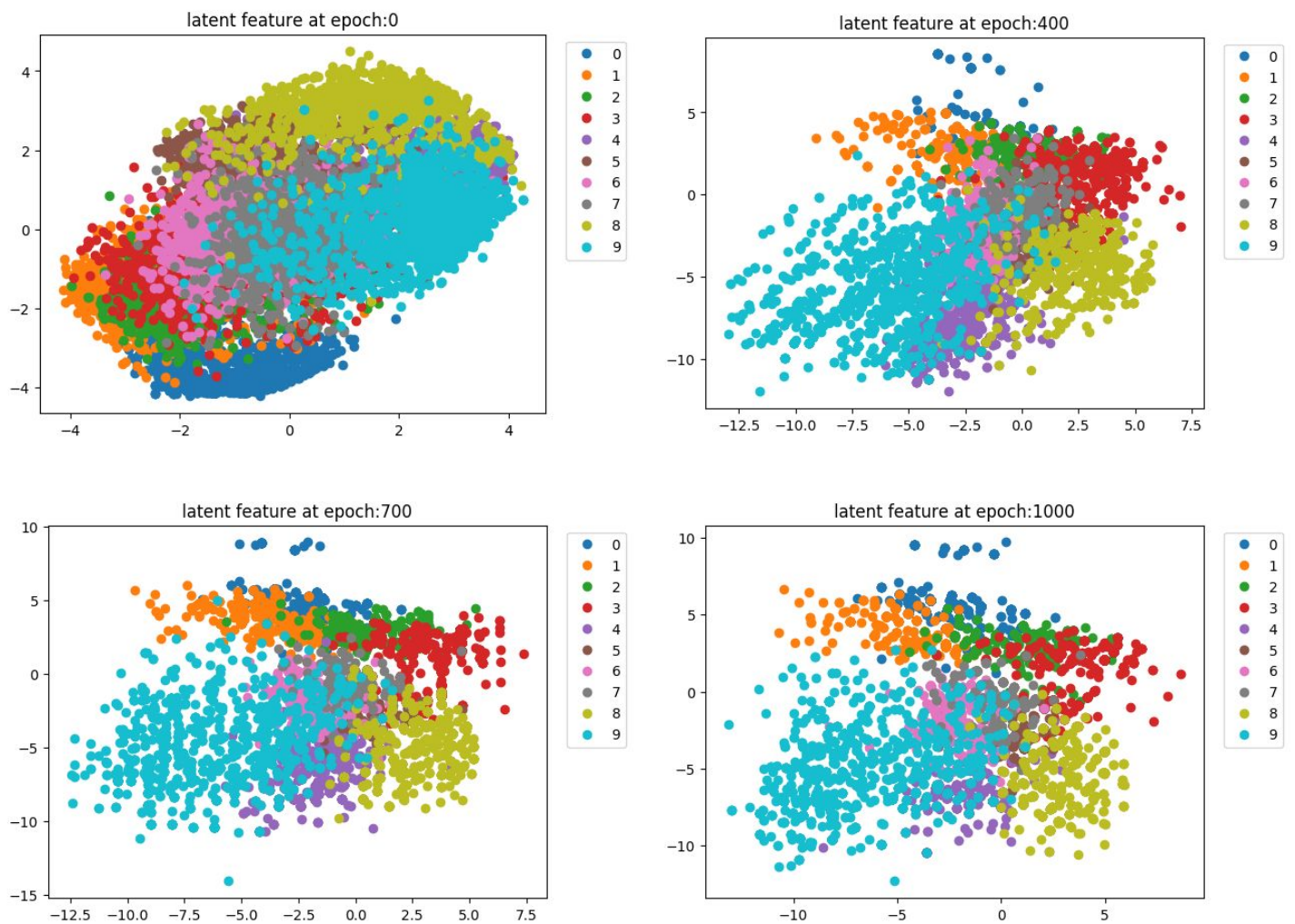
- zero init weights

fail to calculate the gradient using zero init weights



## Latent space visualization

As training goes on. Different class of input gradually groups together.



## Confusion matrix

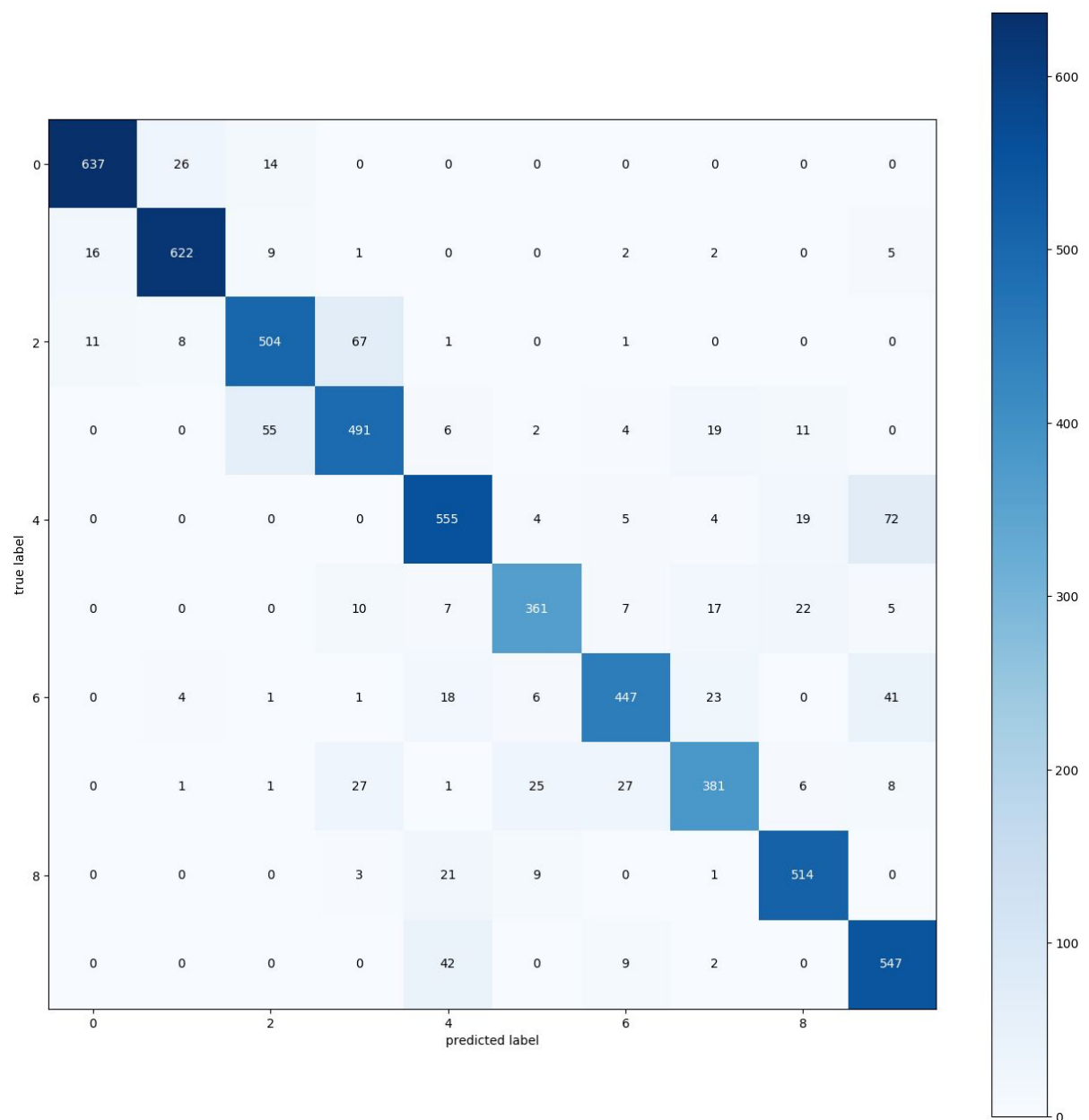
most class are correctly classified.

number 2 have a few cases being recognized as 3

number 3 have a few cases being recognized as 2

number 4 have a few cases being recognized as 9

number 9 have a few cases being recognized as 4



## Problem2

### Image preprocess

crop every face in bounding box, and resize to 80x80x3 to have consistent input, and keep the detail of every image.

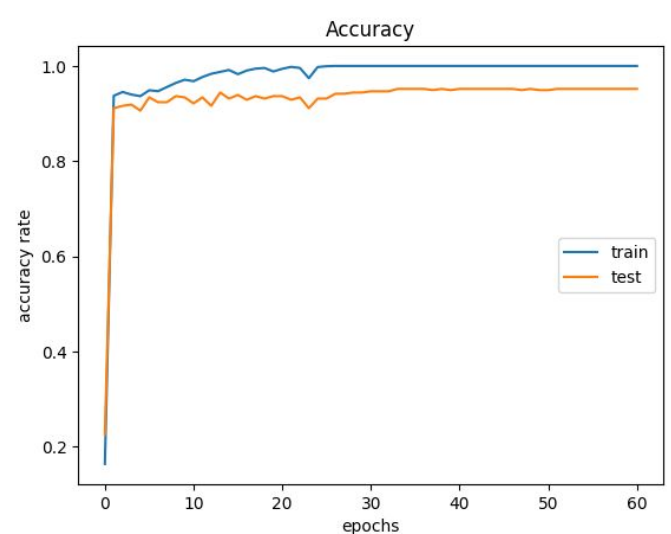
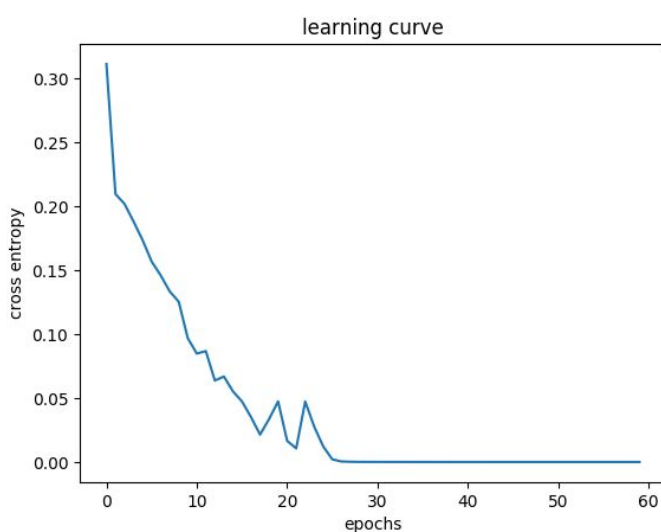


### Network structure & Hyper parameter

- learning rate (0.001)
- batch size(8)
- epoch (60)

#### setting 1

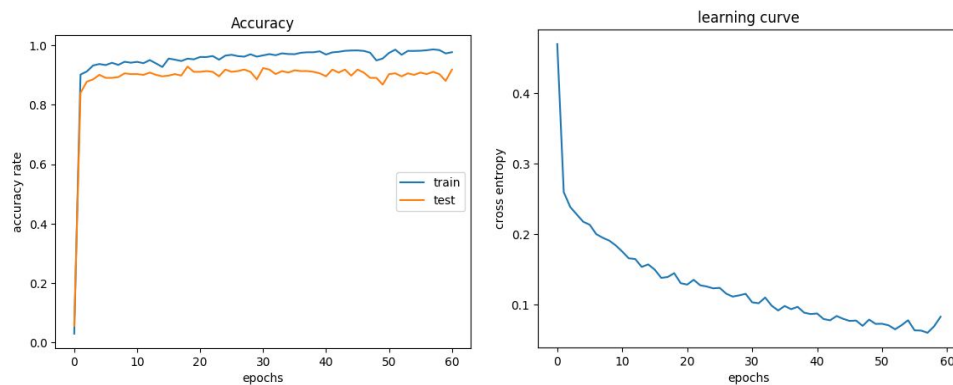
- input(80x80)
- conv2D(input channel: 3, output channel: 6, kernel size: 5, stride: 1)
- maxpool2D(stride: 2)
- conv2D(input channel: 6, output channel: 16, kernel size: 5, stride: 1)
- fully connected(120)
- relu(120)
- fully connected(84)
- relu(84)
- fully connected output (3)



## setting 2

change convolution stride size to 3, need longer training time to converge.

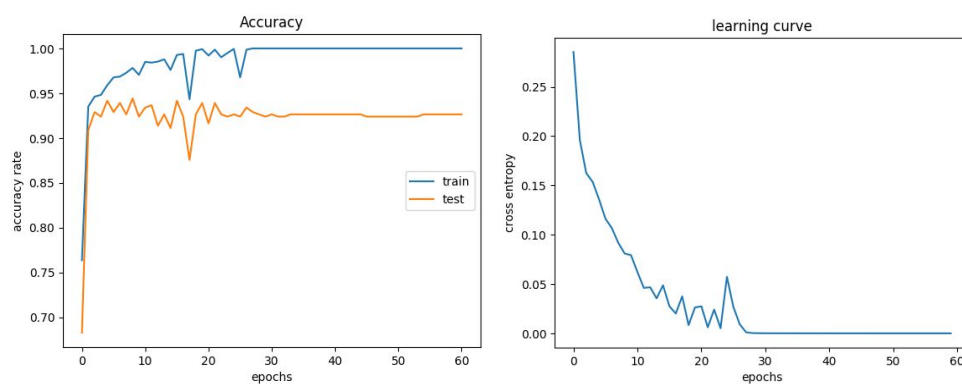
- input(80x80)
- conv2D(input channel: 3, output channel: 6, kernel size: 5, stride: 3)
- maxpool2D(stride: 2)
- conv2D(input channel: 6, output channel: 16, kernel size: 5, stride: 3)
- fully connected(120)
- relu(120)
- fully connected(84)
- relu(84)
- fully connected output (3)



## setting 3

change kernel size to 3, result is similar to setting 1.

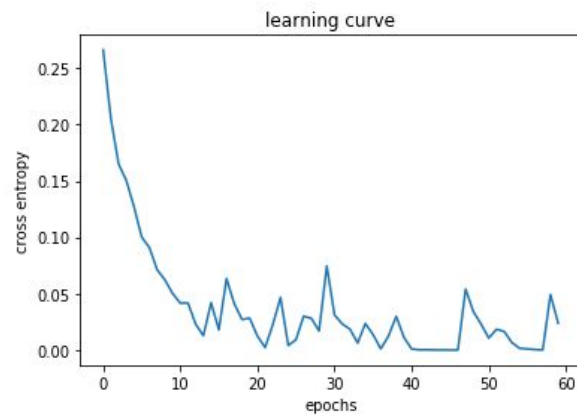
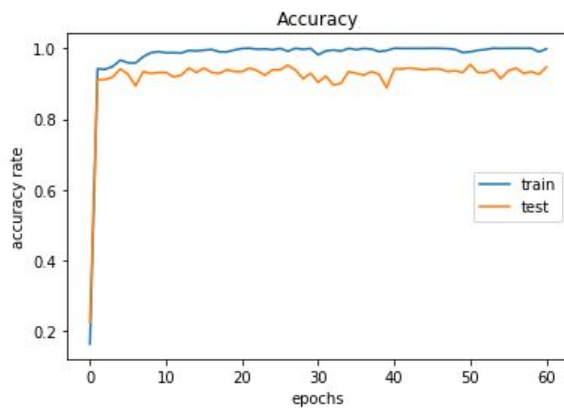
- input(80x80)
- conv2D(input channel: 3, output channel: 6, kernel size: 3, stride: 1)
- maxpool2D(stride: 2)
- conv2D(input channel: 6, output channel: 16, kernel size: 3, stride: 1)
- fully connected(120)
- relu(120)
- fully connected(84)
- relu(84)
- fully connected output (3)



## setting 4

add batch normalization, result is similar to setting 1. But no overfit.

- input(80x80)
- BatchNormalization
- conv2D(input channel: 3, output channel: 6, kernel size: 5, stride: 1)
- maxpool2D(stride: 2)
- BatchNormalization
- conv2D(input channel: 6, output channel: 16, kernel size: 5, stride: 1)
- fully connected(120)
- relu(120)
- fully connected(84)
- relu(84)
- fully connected output (3)

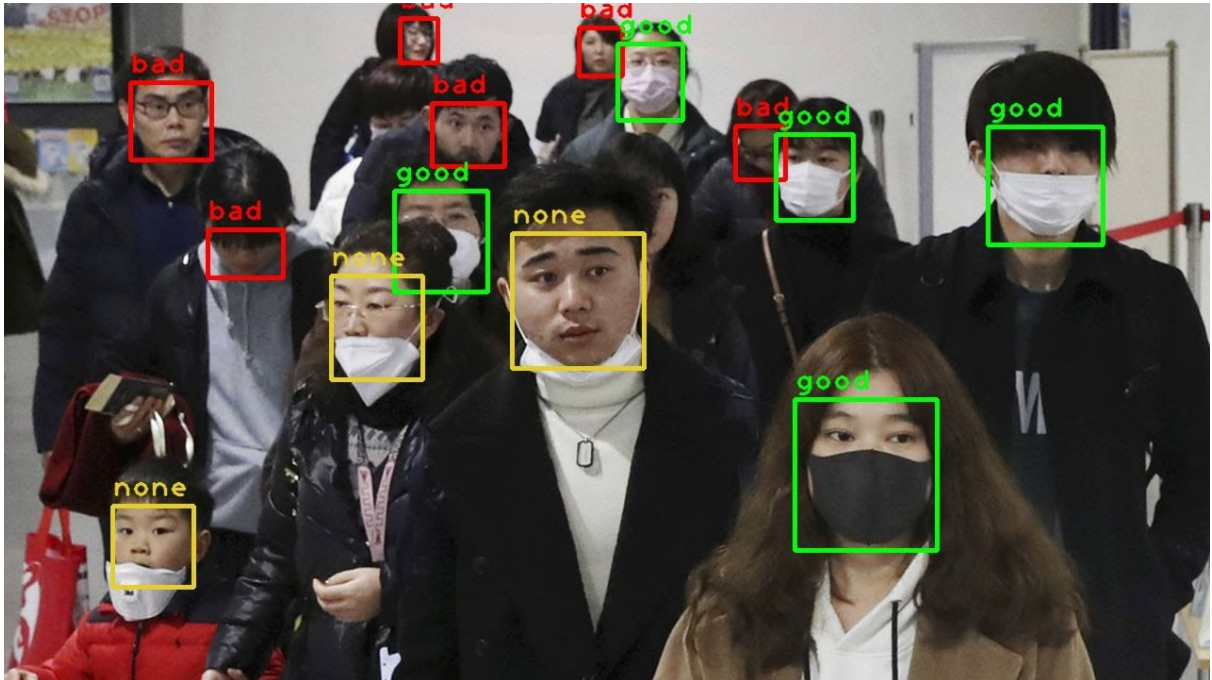




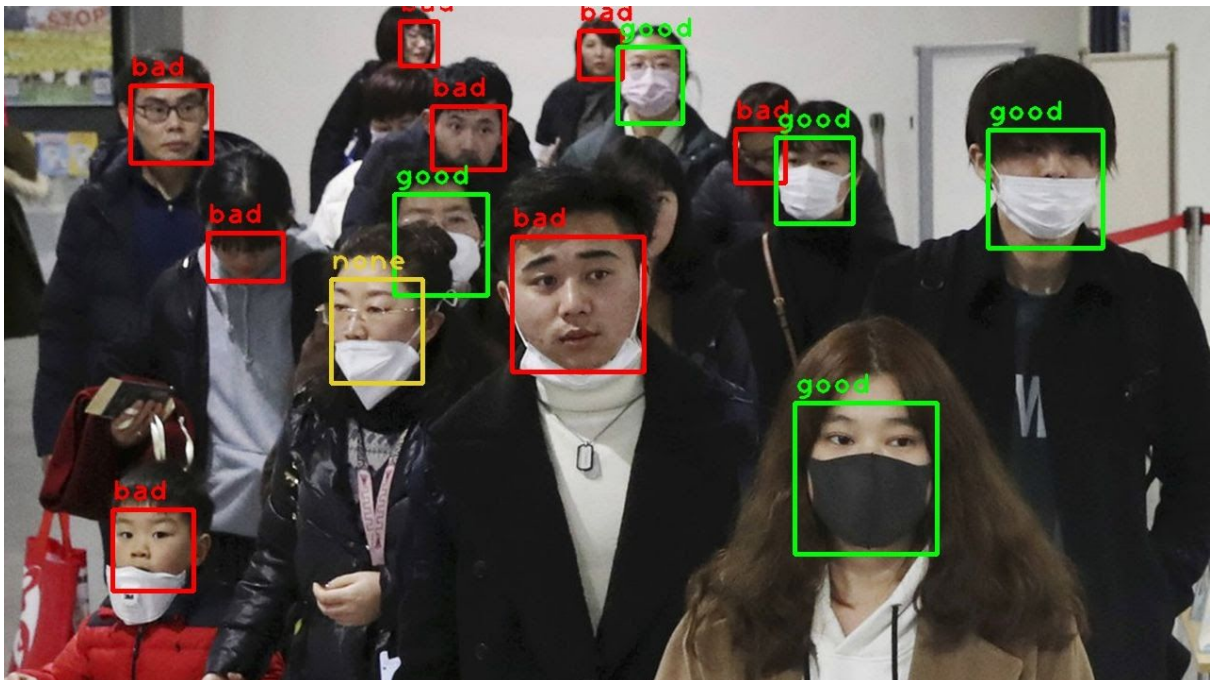
## Visualized example

visualize example of test dataset using model trained by setting 1

- label



- model predict

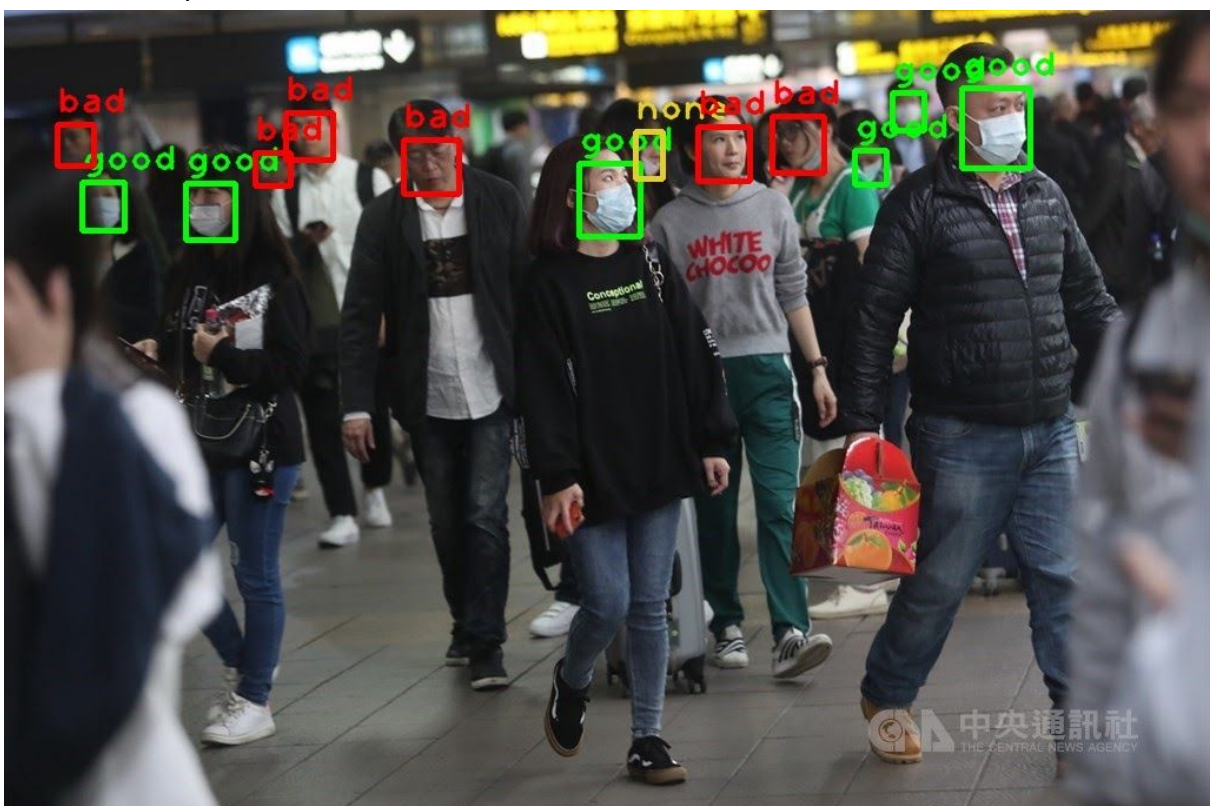




- label



- model predict





- label



- model predict



## Accuracy

accuracy of model trained by setting 1

class	Train accuracy	Test accuracy
Good	1	0.9788
None	1	0.5
Bad	1	0.9775

None class has lowest accuracy because model usually recognized none as good or bad. Sometimes none class looks like there is a mask or looks like the top of the shirts.

