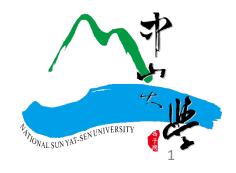
# Assignment 4a

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## Outline

- 1. Fashion-MNIST with Keras
  - Model Fitting
  - Save Checkpoints

### Fashion MNIST



- 70,000 images in 10 classes
- Each class has 7,000 images
- Each grayscale image is 28x28
- 60,000 images for training
- 10,000 images for testing

0	T-shirt/top
1	Trouser
2	Pullover
3	Dress
4	Coat
5	Sandal
6	Shirt
7	Sneaker
8	Bag
9	Ankle boot

- 1. Open assignment\_4a\_template.ipynb
- 2. Use show\_image\_grids() to see samples of Fashion-MNIST

  Understand your training data is very important for the success of machine learning.
- 3. Normalize training and test data

```
train_images = train_images / 255.0
test_images = test_images / 255.0
```

After normalization, the value of each pixel is between [0,1]

4. Create a Keras Sequential Model

```
model = keras.Sequential([
          keras.layers.Flatten(input_shape=input_shape),
          keras.layers.Dense(128, activation=tf.nn.relu),
          keras.layers.Dense(output_shape, activation=tf.nn.softmax)
])
```

You need to determine input\_shape and output\_shape.

- 5. Use model.summary() to see the network structure of the model
- 6. Compile the Keras model

7. [Optional] Create TensorBoard callback

```
tb_callback = TensorBoard(log_dir='log_fashion')
```

8. Create checkpoint feedback

```
checkpoint_path = "ckpt_fashion/cp-{epoch:04d}.ckpt"
cp_callback = ModelCheckpoint(checkpoint_path, save_weights_only=True,
verbose=1, period=5)
```

9. Fit the model

```
model.fit(train_images, train_labels, epochs=20, callbacks =
[cp_callback, tb_callback])
```

10. Compute and print the test accuracy

```
test_loss, test_acc = model.evaluate(test_images, test_labels)
```

11. [Optional] Type the following command in Anaconda Prompt (make sure that log\_fashion/ is in the current directory)

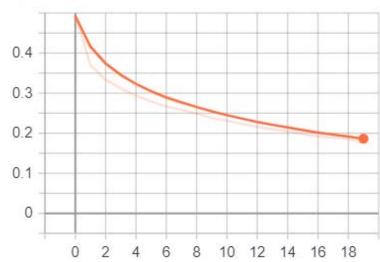
```
tensorboard.exe --logdir log_fashion
```

Navigate your web browser to localhost: 6006 to view the TensorBoard



TensorBoard will display the following two figures.

epoch\_loss



12. Compute and print the test accuracy using ckpt saved at different training epochs:

model.load\_weights(weight\_path)

loss, acc = model.evaluate(test\_images, test\_labels)

where weight\_path is one of the following

ckpt\_fashion/cp-0005.ckpt

ckpt\_fashion/cp-0010.ckpt

ckpt\_fashion/cp-0015.ckpt

ckpt\_fashion/cp-0020.ckpt

13. Load the ckpt saved at epoch = 20.

```
Use prob = model.predict(test_images) to get the probability.
prob.shape is (10000,10)
prob[i,:] is the probability of test_images[i,:,:]
Compute predictions based on prob. For example,
```

```
if prob[i,:] is [0, 0.4, 0.6, 0, 0, 0, 0, 0, 0, 0],
```

then predictions[i] should be 2 (the index corresponding to the largest probability). You can use np.argmax() to compute predictions.

Compute and print the test accuracy by comparing predictions with test\_labels. If predictions[i] equals test\_labels[i], then this means the prediction of ith test image is correct, otherwise the prediction is incorrect. Hint: Use the == operator.

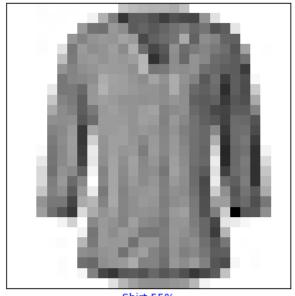
此步驟不需要使用迴圈,如使用迴圈會被扣分。

此步驟得到之 test accuracy 應該與上一頁用 model.evaluate()得到的 test accuracy 一致。

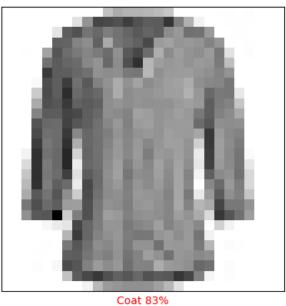
14. Display the prediction result of test\_image[4] and np.fliplr(test\_images[4]) as in the following two figures (左圖是 test\_image[4] 預測結果,右圖是對 test\_image[4] 左右翻轉後之預測結果).

Use prob = model.predict(test\_images[4:5]). Note that model.predict() is assumed to predict multiple images at once, and

```
test_images[4:5].shape is (1,28,28)
test_images[4].shape is (28,28)
```



Shirt 55% (Shirt is the true label)



Coat 83% (Shirt is the true label)

```
plt.grid(False)
plt.xticks([])
plt.yticks([])
plt.imshow(img, cmap=plt.cm.binary)
plt.show()
```

- Use plt.xlabel(description) to display to xlabel, where description contains the words we wish to add to the image.
- You need to use prob and class\_names to generate description.

#### Questions for Assignment 4a

1. What is the difference between the following two options in model.compile()?

```
loss='categorical_crossentropy'
loss='sparse_categorical_crossentropy'
```

Use an example to illustrate the difference.