

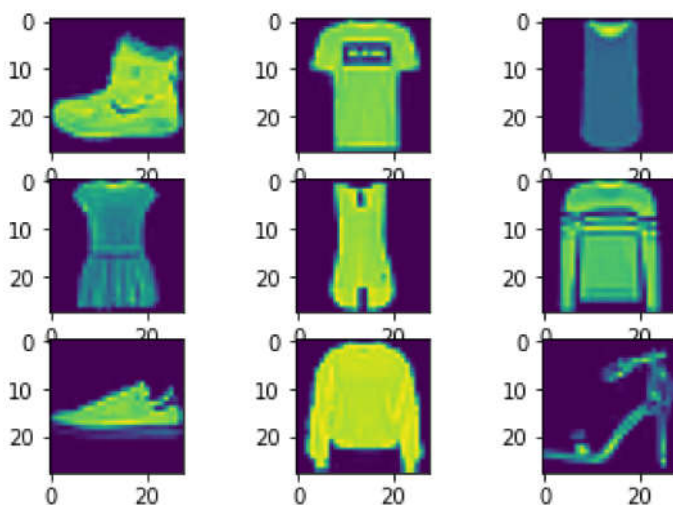
<https://github.com/DoanAI/Fashion.git>

```
from keras.datasets import fashion_mnist
import matplotlib.pyplot as plt
```

```
(x_train, y_train), (x_test, y_test) = fashion_mnist.load_data()
print(x_train.shape)
print(x_test.shape)
```

```
Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datasets
32768/29515 [=====] - 0s 0us/step
40960/29515 [=====] - 0s 0us/step
Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datasets
26427392/26421880 [=====] - 0s 0us/step
26435584/26421880 [=====] - 0s 0us/step
Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datasets
16384/5148 [=====]
Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datasets
4423680/4422102 [=====] - 0s 0us/step
4431872/4422102 [=====] - 0s 0us/step
(60000, 28, 28)
(10000, 28, 28)
```

```
for i in range(9):
    plt.subplot(330+i+1)    # 330 mean: 3 hang 3 cot
    plt.imshow(x_train[i])
plt.show()
```



```
x = x_test
from tensorflow.keras.utils import to_categorical
x_train = x_train.reshape((x_train.shape[0], x_train.shape[1], x_train.shape[2], 1))
x_test = x_test.reshape((x_test.shape[0], x_test.shape[1], x_test.shape[2], 1))
```

```
print(x_train.shape, y_train.shape)
print(x_test.shape, y_test.shape)
```

```
(60000, 28, 28, 1) (60000,)
(10000, 28, 28, 1) (10000,)
```

```
x_train = x_train.astype('float32')/255
x_test = x_test.astype('float32')/255
```

Create Model

```
from keras.models import Sequential
from tensorflow.keras.layers import Flatten, MaxPooling2D, Conv2D
model = Sequential()
```

```
from tensorflow.keras.layers import Dense, Activation, Dropout
model.add(Conv2D(32, (3,3), activation='relu', input_shape= (28,28,1)))

model.add(MaxPooling2D((2,2)))

model.add(Conv2D(48, (3,3), activation='relu'))

model.add(MaxPooling2D((2,2)))

model.add(Dropout(0.5))

model.add(Flatten())

model.add(Dense(500, activation='relu'))

model.add(Dense(10, activation='softmax'))
model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
=====		
conv2d (Conv2D)	(None, 26, 26, 32)	320
max_pooling2d (MaxPooling2D)	(None, 13, 13, 32)	0
conv2d_1 (Conv2D)	(None, 11, 11, 48)	13872
max_pooling2d_1 (MaxPooling2D)	(None, 5, 5, 48)	0
dropout (Dropout)	(None, 5, 5, 48)	0
flatten (Flatten)	(None, 1200)	0
dense (Dense)	(None, 500)	600500

dense_1 (Dense) (None, 10) 5010

```
=====
Total params: 619,702
Trainable params: 619,702
Non-trainable params: 0
=====
```

```
model.compile(optimizer='adam', loss='sparse_categorical_crossentropy', metrics=['accuracy'])
```

```
history = model.fit(x_train, y_train, epochs=10, batch_size = 128, verbose= 2, validation_data=(x_test, y_test))
```

```
Epoch 1/10
422/422 - 41s - loss: 0.5863 - accuracy: 0.7853 - val_loss: 0.4044 - val_accuracy: 0.8516
Epoch 2/10
422/422 - 42s - loss: 0.4057 - accuracy: 0.8516 - val_loss: 0.3283 - val_accuracy: 0.8701
Epoch 3/10
422/422 - 43s - loss: 0.3542 - accuracy: 0.8701 - val_loss: 0.3142 - val_accuracy: 0.8809
Epoch 4/10
422/422 - 40s - loss: 0.3200 - accuracy: 0.8809 - val_loss: 0.2915 - val_accuracy: 0.8897
Epoch 5/10
422/422 - 40s - loss: 0.2981 - accuracy: 0.8897 - val_loss: 0.2724 - val_accuracy: 0.8968
Epoch 6/10
422/422 - 39s - loss: 0.2792 - accuracy: 0.8968 - val_loss: 0.2660 - val_accuracy: 0.9016
Epoch 7/10
422/422 - 39s - loss: 0.2651 - accuracy: 0.9016 - val_loss: 0.2633 - val_accuracy: 0.9046
Epoch 8/10
422/422 - 39s - loss: 0.2535 - accuracy: 0.9046 - val_loss: 0.2457 - val_accuracy: 0.9084
Epoch 9/10
422/422 - 39s - loss: 0.2446 - accuracy: 0.9084 - val_loss: 0.2481 - val_accuracy: 0.9121
Epoch 10/10
422/422 - 39s - loss: 0.2348 - accuracy: 0.9121 - val_loss: 0.2388 - val_accuracy: 0.9121
```



```
score = model.evaluate(x_test, y_test, verbose=0)
print('Sai so kiem tra la:', score[0])
print('Do chinh xac kiem tra la:', score[1]*100)
```

```
Sai so kiem tra la: 0.25306054949760437
Do chinh xac kiem tra la: 90.61999917030334
```

```
model.save('CNN_FASHION_MNIST.h5')
from keras.models import load_model
model1 = load_model('CNN_FASHION_MNIST.h5')
```

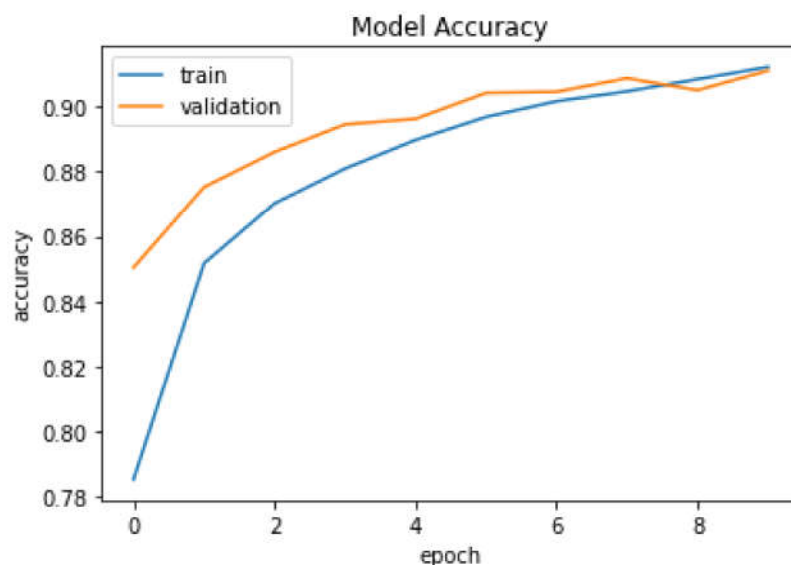
```
plt.plot(history.history['accuracy'])
plt.plot(history.history['val_accuracy'])
plt.title('Model Accuracy')
plt.ylabel('accuracy')
plt.xlabel('epoch')
plt.legend(['train', 'validation'], loc='upper-left')
```

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:6: MatplotlibDeprecationWarning:

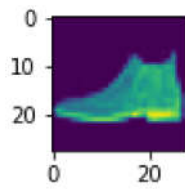
best
upper right
upper left
lower left
lower right
right
center left
center right
lower center
upper center
center

This will raise an exception in 3.3.

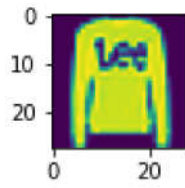
<matplotlib.legend.Legend at 0x7f7e2572ff50>



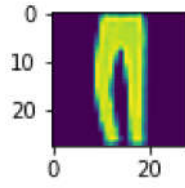
```
import numpy as np
y_pred = model.predict(x_test)
for i in range(9):
    plt.subplot(330+i+1)    # 330 mean: 3 hang 3 cot
    plt.imshow(x[i])
    plt.show()
    print(np.round(y_pred[i]))
```



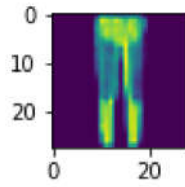
[0. 0. 0. 0. 0. 0. 0. 0. 0. 1.]



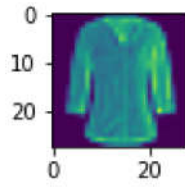
[0. 0. 1. 0. 0. 0. 0. 0. 0. 0.]



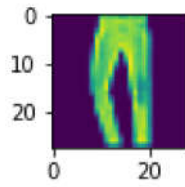
[0. 1. 0. 0. 0. 0. 0. 0. 0. 0.]



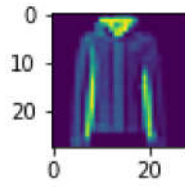
[0. 1. 0. 0. 0. 0. 0. 0. 0. 0.]



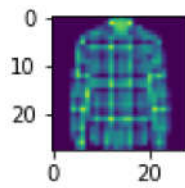
[0. 0. 0. 0. 0. 0. 1. 0. 0. 0.]



[0. 1. 0. 0. 0. 0. 0. 0. 0. 0.]



[0. 0. 0. 0. 1. 0. 0. 0. 0. 0.]



[0. 0. 0. 0. 0. 0. 1. 0. 0. 0.]

