

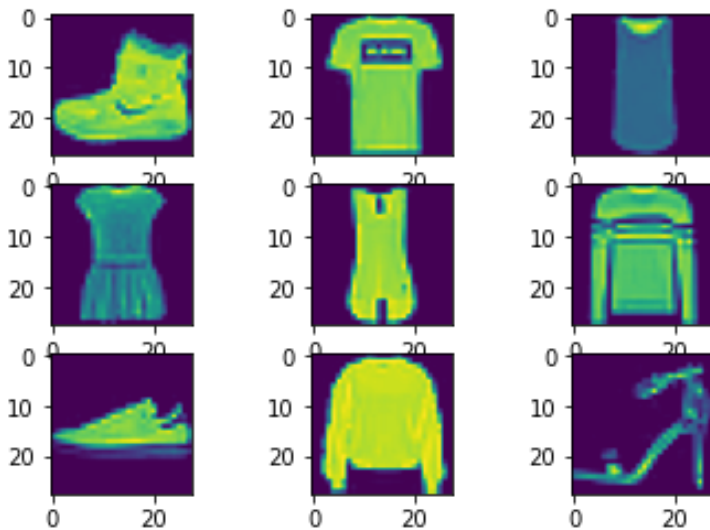
## CNN\_fashion

```
from keras.datasets import fashion_mnist
import matplotlib.pyplot as plt
from tensorflow.keras.utils import to_categorical
```

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

```
(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
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
```

```
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_s


```

```

Epoch 1/10
422/422 - 15s - loss: 0.5899 - accuracy: 0.7848 - val_loss: 0.4024 - val_accuracy: 0.
Epoch 2/10
422/422 - 2s - loss: 0.4136 - accuracy: 0.8513 - val_loss: 0.3494 - val_accuracy: 0.8
Epoch 3/10
422/422 - 2s - loss: 0.3621 - accuracy: 0.8671 - val_loss: 0.3236 - val_accuracy: 0.8
Epoch 4/10
422/422 - 2s - loss: 0.3321 - accuracy: 0.8777 - val_loss: 0.3105 - val_accuracy: 0.8

```

```
Epoch 5/10
422/422 - 2s - loss: 0.3093 - accuracy: 0.8867 - val_loss: 0.2802 - val_accuracy: 0.8
Epoch 6/10
422/422 - 2s - loss: 0.2867 - accuracy: 0.8934 - val_loss: 0.2633 - val_accuracy: 0.9
Epoch 7/10
422/422 - 2s - loss: 0.2768 - accuracy: 0.8966 - val_loss: 0.2553 - val_accuracy: 0.9
Epoch 8/10
422/422 - 2s - loss: 0.2598 - accuracy: 0.9030 - val_loss: 0.2464 - val_accuracy: 0.9
Epoch 9/10
422/422 - 2s - loss: 0.2519 - accuracy: 0.9053 - val_loss: 0.2391 - val_accuracy: 0.9
Epoch 10/10
422/422 - 2s - loss: 0.2421 - accuracy: 0.9095 - val_loss: 0.2361 - val_accuracy: 0.9
```



```
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: 0.25060924887657166
Do chinh xac kiem tra la: 90.6000018119812
```

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

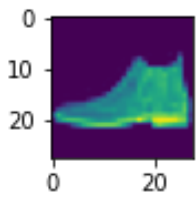
```
plt.plot(history.history['accuracy'])
plt.plot(history.history['val_accuracy'])
plt.title('Model Accuracy')
plt.ylabel('accuracy')
plt.xlabel('epoch')
plt.legend(['train', 'validation'], loc='upper-left')
```

/usr/local/lib/python3.7/dist-packages/ipykernel\_launcher.py:6: MatplotlibDeprecation

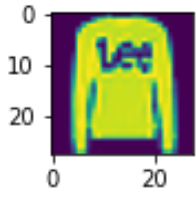
best  
upper right  
upper left  
lower left  
lower right  
right  
center left  
center right  
lower center  
upper center  
center

```
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]))
```

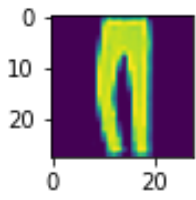




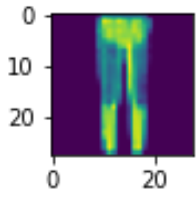
[0. 0. 0. 0. 0. 0. 0. 0. 0. 1.]



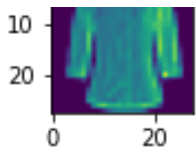
[0. 0. 1. 0. 0. 0. 0. 0. 0. 0.]



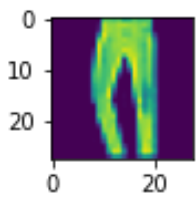
[0. 1. 0. 0. 0. 0. 0. 0. 0. 0.]



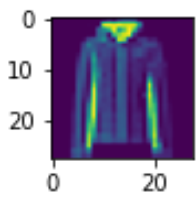
[0. 0. 0. 0. 0. 0. 0. 0. 0. 0.]



[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.]

