

## ▼ Build a DNN using Keras with RELU and ADAM

### ▼ Load tensorflow

```
!pip install -U tensorflow --quiet
```

### ▼ Collect Fashion mnist data from tf.keras.datasets

```
import tensorflow as tf
import pandas as pd
import numpy as np
tf.__version__
```

```
↳ '2.0.0'
```

### ▼ Change train and test labels into one-hot vectors

```
(x_train, trainY),(x_test, testY) = tf.keras.datasets.fashion_mnist.load_data()
```

```
print("Number of samples in Training are x_train and y_train ",(x_train.shape,trainY.shape))
```

```
↳ Number of samples in Training are x_train and y_train ((60000, 28, 28), (60000,))
```

```
print("Number of samples in Test are x_test and y_test",(x_test.shape,testY.shape))
```

```
↳ Number of samples in Test are x_test and y_test ((10000, 28, 28), (10000,))
```

```
x_train.dtype, x_test.dtype
```

```
↳ (dtype('uint8'), dtype('uint8'))
```

```
trainX = x_train.astype('float32')
```

```
testX = x_test.astype('float32')
```

```
trainX.dtype, testX.dtype
```

```
↳ (dtype('float32'), dtype('float32'))
```

```
set(trainY)
```

```
↳ {0, 1, 2, 3, 4, 5, 6, 7, 8, 9}
```

```
pd.value_counts(trainY)
```

```
↳
```

9	6000
8	6000
7	6000
6	6000
5	6000
4	6000
3	6000
2	6000
1	6000
0	6000

```
dtype: int64
```

```
pd.value_counts(testY)
```

```
↳
```

7	1000
6	1000
5	1000
4	1000
3	1000
2	1000
9	1000
1	1000
8	1000
0	1000

```
dtype: int64
```

```
trainX[0]
```

```
↳
```

```

array([[ 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.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,
        0.,  1.,  0.,  0., 13., 73.,  0.,  0.,  1.,  4.,  0.,
        0.,  0.,  0.,  1.,  1.,  0.],
       [ 0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,
        0.,  3.,  0., 36.,136.,127., 62., 54.,  0.,  0.,  0.,
        1.,  3.,  4.,  0.,  0.,  3.],
       [ 0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,
        0.,  6.,  0.,102.,204.,176.,134.,144.,123.,23.,  0.,
        0.,  0.,  0.,12.,10.,  0.],
       [ 0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,
        0.,  0.,  0.,155.,236.,207.,178.,107.,156.,161.,109.,
        64.,23.,77.,130.,72.,15.],
       [ 0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,
        1.,  0., 69.,207.,223.,218.,216.,216.,163.,127.,121.,
        122.,146.,141.,88.,172.,66.],
       [ 0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,  1.,  1.,
        1.,  0.,200.,232.,232.,233.,229.,223.,223.,215.,213.,
        164.,127.,123.,196.,229.,  0.],
       [ 0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,
        0.,  0.,183.,225.,216.,223.,228.,235.,227.,224.,222.,
        224.,221.,223.,245.,173.,  0.],
       [ 0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,
        0.,  0.,193.,228.,218.,213.,198.,180.,212.,210.,211.,
        213.,223.,220.,243.,202.,  0.],
       [ 0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,  1.,  3.,
        0.,12.,219.,220.,212.,218.,192.,169.,227.,208.,218.,
        224.,212.,226.,197.,209.,52.],
       [ 0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,  6.,
        0.,99.,244.,222.,220.,218.,203.,198.,221.,215.,213.,
        222.,220.,245.,119.,167.,56.],
       [ 0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,  4.,  0.,
        0.,55.,236.,228.,230.,228.,240.,232.,213.,218.,223.,
        234.,217.,217.,209.,92.,  0.],
       [ 0.,  0.,  1.,  4.,  6.,  7.,  2.,  0.,  0.,  0.,  0.,
        0.,237.,226.,217.,223.,222.,219.,222.,221.,216.,223.,
        229.,215.,218.,255.,77.,  0.],
       [ 0.,  3.,  0.,  0.,  0.,  0.,  0.,  0.,  0., 62.,145.,
        204.,228.,207.,213.,221.,218.,208.,211.,218.,224.,223.,
        219.,215.,224.,244.,159.,  0.],
       [ 0.,  0.,  0.,  0.,18.,44.,82.,107.,189.,228.,220.,
        222.,217.,226.,200.,205.,211.,230.,224.,234.,176.,188.,
        250.,248.,233.,238.,215.,  0.],
       [ 0.,57.,187.,208.,224.,221.,224.,208.,204.,214.,208.,
        209.,200.,159.,245.,193.,206.,223.,255.,255.,221.,234.,
        221.,211.,220.,232.,246.,  0.],
       [ 3.,202.,228.,224.,221.,211.,211.,214.,205.,205.,205.,
        220.,240.,80.,150.,255.,229.,221.,188.,154.,191.,210.,
        204.,209.,222.,228.,225.,  0.],
       -

```

```
[ 98., 233., 198., 210., 222., 229., 229., 234., 249., 220., 194.,
 215., 217., 241., 65., 73., 106., 117., 168., 219., 221., 215.,
 217., 223., 223., 224., 229., 29.],
[ 75., 204., 212., 204., 193., 205., 211., 225., 216., 185., 197.,
 206., 198., 213., 240., 195., 227., 245., 239., 223., 218., 212.,
 209., 222., 220., 221., 230., 67.],
[ 48., 203., 183., 194., 213., 197., 185., 190., 194., 192., 202.,
 214., 219., 221., 220., 236., 225., 216., 199., 206., 186., 181.,
 177., 172., 181., 205., 206., 115.],
[ 0., 122., 219., 193., 179., 171., 183., 196., 204., 210., 213.,
 207., 211., 210., 200., 196., 194., 191., 195., 191., 198., 192.,
 176., 156., 167., 177., 210., 92.],
[ 0., 0., 74., 189., 212., 191., 175., 172., 175., 181., 185.,
 188., 189., 188., 193., 198., 204., 209., 210., 210., 211., 188.,
 188., 194., 192., 216., 170., 0.],
[ 2., 0., 0., 0., 66., 200., 222., 237., 239., 242., 246.,
 243., 244., 221., 220., 193., 191., 179., 182., 182., 181., 176.,
 166., 168., 99., 58., 0., 0.],
[ 0., 0., 0., 0., 0., 0., 0., 40., 61., 44., 72.,
 41., 35., 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 11 dtype=float32)
```

```
trainX = trainX / 255
```

```
trainX[0]
```



```

array([[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.          , 0.00392157, 0.          , 0.          ,
        0.05098039, 0.28627452, 0.          , 0.          , 0.00392157,
        0.01568628, 0.          , 0.          , 0.          , 0.          ,
        0.00392157, 0.00392157, 0.          ],
       [0.          , 0.          , 0.          , 0.          , 0.          ,
        0.          , 0.          , 0.          , 0.          , 0.          ,
        0.          , 0.          , 0.01176471, 0.          , 0.14117648,
        0.53333336, 0.49803922, 0.24313726, 0.21176471, 0.          ,
        0.          , 0.          , 0.00392157, 0.01176471, 0.01568628,
        0.          , 0.          , 0.01176471],
       [0.          , 0.          , 0.          , 0.          , 0.          ,
        0.          , 0.          , 0.          , 0.          , 0.          ,
        0.          , 0.          , 0.02352941, 0.          , 0.4          ,
        0.8          , 0.6901961 , 0.5254902 , 0.5647059 , 0.48235294,
        0.09019608, 0.          , 0.          , 0.          , 0.          ,
        0.04705882, 0.03921569, 0.          ],
       [0.          , 0.          , 0.          , 0.          , 0.          ,
        0.          , 0.          , 0.          , 0.          , 0.          ,
        0.          , 0.          , 0.          , 0.          , 0.60784316,
        0.9254902 , 0.8117647 , 0.69803923, 0.41960785, 0.6117647 ,
        0.6313726 , 0.42745098, 0.2509804 , 0.09019608, 0.3019608 ,
        0.50980395, 0.28235295, 0.05882353],
       [0.          , 0.          , 0.          , 0.          , 0.          ,
        0.          , 0.          , 0.          , 0.          , 0.          ,
        0.          , 0.00392157, 0.          , 0.27058825, 0.8117647 ,
        0.8745098 , 0.85490197, 0.84705883, 0.84705883, 0.6392157 ,
        0.49803922, 0.4745098 , 0.47843137, 0.57254905, 0.5529412 ,
        0.34509805, 0.6745098 , 0.25882354],
       [0.          , 0.          , 0.          , 0.          , 0.          ,
        0.          , 0.          , 0.          , 0.          , 0.00392157,
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        0.9098039 , 0.9137255 , 0.8980392 , 0.8745098 , 0.8745098 ,
        0.84313726, 0.8352941 , 0.6431373 , 0.49803922, 0.48235294,
        0.76862746, 0.8980392 , 0.          ],
       [0.          , 0.          , 0.          , 0.          , 0.          ,
        0.          , 0.          , 0.          , 0.          , 0.          ,
        0.          , 0.          , 0.          , 0.          , 0.          ,
        0.          , 0.          , 0.          , 0.7176471 , 0.88235295,

```

```

0.84705883, 0.8745098 , 0.89411765, 0.92156863, 0.8901961 ,
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0.9607843 , 0.6784314 , 0.      ],
[0.      , 0.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      , 0.      , 0.      ,
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0.85490197, 0.8352941 , 0.7764706 , 0.7058824 , 0.83137256,
0.8235294 , 0.827451 , 0.8352941 , 0.8745098 , 0.8627451 ,
0.9529412 , 0.7921569 , 0.      ],
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0.      , 0.      , 0.      , 0.      , 0.00392157,
0.01176471, 0.      , 0.04705882, 0.85882354, 0.8627451 ,
0.83137256, 0.85490197, 0.7529412 , 0.6627451 , 0.8901961 ,
0.8156863 , 0.85490197, 0.8784314 , 0.83137256, 0.8862745 ,
0.77254903, 0.81960785, 0.20392157],
[0.      , 0.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      , 0.      , 0.      ,
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0.8627451 , 0.85490197, 0.79607844, 0.7764706 , 0.8666667 ,
0.84313726, 0.8352941 , 0.87058824, 0.8627451 , 0.9607843 ,
0.46666667, 0.654902 , 0.21960784],
[0.      , 0.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      , 0.      , 0.01568628,
0.      , 0.      , 0.21568628, 0.9254902 , 0.89411765,
0.9019608 , 0.89411765, 0.9411765 , 0.9098039 , 0.8352941 ,
0.85490197, 0.8745098 , 0.91764706, 0.8509804 , 0.8509804 ,
0.81960785, 0.36078432, 0.      ],
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0.02745098, 0.00784314, 0.      , 0.      , 0.      ,
0.      , 0.      , 0.92941177, 0.8862745 , 0.8509804 ,
0.8745098 , 0.87058824, 0.85882354, 0.87058824, 0.8666667 ,
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1.      , 0.3019608 , 0.      ],
[0.      , 0.01176471, 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      , 0.      , 0.24313726,
0.5686275 , 0.8      , 0.89411765, 0.8117647 , 0.8352941 ,
0.8666667 , 0.85490197, 0.8156863 , 0.827451 , 0.85490197,
0.8784314 , 0.8745098 , 0.85882354, 0.84313726, 0.8784314 ,
0.95686275, 0.62352943, 0.      ],
[0.      , 0.      , 0.      , 0.      , 0.07058824,
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0.8627451 , 0.87058824, 0.8509804 , 0.8862745 , 0.78431374,
0.8039216 , 0.827451 , 0.9019608 , 0.8784314 , 0.91764706,
0.6901961 , 0.7372549 , 0.98039216, 0.972549 , 0.9137255 ,
0.93333334, 0.84313726, 0.      ],
[0.      , 0.22352941, 0.73333335, 0.8156863 , 0.8784314 ,
0.8666667 , 0.8784314 , 0.8156863 , 0.8      , 0.8392157 ,
0.8156863 , 0.81960785, 0.78431374, 0.62352943, 0.9607843 ,
0.75686276, 0.80784315, 0.8745098 , 1.      , 1.      ,
0.8666667 , 0.91764706, 0.8666667 , 0.827451 , 0.8627451 ,
0.9098039 , 0.9647059 , 0.      ],
[0.01176471, 0.7921569 , 0.89411765, 0.8784314 , 0.8666667 ,
0.827451 , 0.827451 , 0.8392157 , 0.8039216 , 0.8039216 ,
0.8039216 , 0.8627451 , 0.9411765 , 0.3137255 , 0.5882353 ,
1.      , 0.8980392 , 0.8666667 , 0.7372549 , 0.6039216 ,
0.7490196 , 0.8235294 , 0.8      , 0.81960785, 0.87058824,
0.89411765, 0.88235295, 0.      ],
[0.38431373, 0.9137255 , 0.7764706 , 0.8235294 , 0.87058824,

```

```
testX = testX / 255
```

## ▼ We have total of 10 classes in target variables

```
trainY = tf.keras.utils.to_categorical(trainY, num_classes=10)
testY = tf.keras.utils.to_categorical(testY, num_classes=10)
```

```
trainY[0]
```

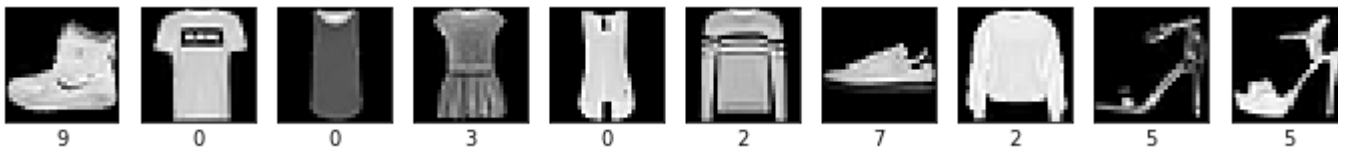
```
↳ array([0., 0., 0., 0., 0., 0., 0., 0., 0., 1.], dtype=float32)
```

```
np.argmax(trainY[0])
```

```
↳ 9
```

```
from matplotlib import pyplot as plt
plt.figure(figsize=(12,12))
for i in range(10):
    plt.subplot(1,10,i+1)
    plt.xticks([])
    plt.yticks([])
    plt.grid(False)
    plt.imshow(trainX[i], cmap='gray')
    plt.xlabel(np.argmax(trainY[i]))
plt.show()
```

```
↳
```



## Build the Graph

### ▼ Initialize model, reshape & normalize data

```
#Clear out model from current memory
tf.keras.backend.clear_session()
```

```
#Initialize Sequential model
model = tf.keras.models.Sequential()
```

```
#Reshape data from 2D to 1D -> 28x28 to 784
model.add(tf.keras.layers.Reshape((784,), input_shape=(28,28,)))
```

```
#Normalize the data
model.add(tf.keras.layers.BatchNormalization())
```

### ▼ Add two fully connected layers with 200 and 100 neurons respectively with relu activation



```
#Add 1st hidden layer
model.add(tf.keras.layers.Dense(200, activation='relu'))
model.add(tf.keras.layers.BatchNormalization())
# Dropout
model.add(tf.keras.layers.Dropout(0.25))

#Add 2nd hidden layer
model.add(tf.keras.layers.Dense(100, activation='relu'))
model.add(tf.keras.layers.BatchNormalization())
model.add(tf.keras.layers.Dropout(0.25))
```

▼ Add the output layer with a fully connected layer with 10 neurons with softmax activation, categorical\_crossentropy loss and adam optimizer and train the network. And, r

```
#Add OUTPUT layer
model.add(tf.keras.layers.Dense(10, kernel_initializer='he_normal',
                                activation='softmax'))

#Compile the model
model.compile(optimizer='adam', loss='categorical_crossentropy',
              metrics=['accuracy'])

from datetime import datetime
logdir = "logs/scalars/" + datetime.now().strftime("%Y%m%d-%H%M%S")
tensorboard_callback = tf.keras.callbacks.TensorBoard(log_dir=logdir)

model.summary()
```



Model: "sequential"

Layer (type)	Output Shape	Param #
reshape (Reshape)	(None, 784)	0
batch_normalization (Batch Normalization)	(None, 784)	3136
dense (Dense)	(None, 200)	157000
batch_normalization_1 (Batch Normalization)	(None, 200)	800
dropout (Dropout)	(None, 200)	0
dense_1 (Dense)	(None, 100)	20100
batch_normalization_2 (Batch Normalization)	(None, 100)	400
dropout_1 (Dropout)	(None, 100)	0
dense_2 (Dense)	(None, 10)	1010
Total params: 182,446		
Trainable params: 180,278		
Non-trainable params: 2,168		

#Modelcheckpoint callback

```
ckpt = tf.keras.callbacks.ModelCheckpoint('mnist_v1.hdf5', save_best_only=True,
                                          monitor='val_loss', mode='min')
```

```
model.fit(trainX,trainY,
          validation_data=(testX,testY),
          epochs=10,
          batch_size=32, callbacks=[ckpt,tensorboard_callback])
```



Train on 60000 samples, validate on 10000 samples

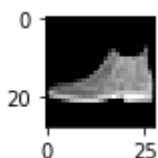
```
Epoch 1/10
60000/60000 [=====] - 15s 254us/sample - loss: 0.5566 - accurac
Epoch 2/10
60000/60000 [=====] - 13s 215us/sample - loss: 0.4295 - accurac
Epoch 3/10
60000/60000 [=====] - 14s 230us/sample - loss: 0.3919 - accurac
Epoch 4/10
60000/60000 [=====] - 14s 228us/sample - loss: 0.3737 - accurac
Epoch 5/10
60000/60000 [=====] - 14s 228us/sample - loss: 0.3569 - accurac
Epoch 6/10
60000/60000 [=====] - 17s 277us/sample - loss: 0.3374 - accurac
Epoch 7/10
60000/60000 [=====] - 16s 272us/sample - loss: 0.3278 - accurac
Epoch 8/10
60000/60000 [=====] - 16s 270us/sample - loss: 0.3225 - accurac
Epoch 9/10
60000/60000 [=====] - 17s 292us/sample - loss: 0.3114 - accurac
Epoch 10/10
60000/60000 [=====] - 16s 274us/sample - loss: 0.3032 - accurac
<tensorflow.python.keras.callbacks.History at 0x7f0c9df5e940>
```

```
y_pred = model.predict(testX)
print(y_pred)
```

```
[ 1.1615418e-06  1.0253641e-06  1.7382665e-06 ...  1.0170623e-01
  2.9786319e-05  8.7610346e-01]
[ 8.9308203e-05  3.5240062e-09  9.9870336e-01 ...  7.1629977e-08
  9.9528403e-09  5.6840744e-08]
[ 1.5153137e-06  9.9999607e-01  1.6575723e-07 ...  7.7957178e-09
  2.5276262e-07  5.2019207e-09]
...
[ 1.4396167e-03  4.4202224e-07  3.3199985e-04 ...  1.1025061e-05
  9.9482441e-01  3.1092497e-07]
[ 1.7323366e-06  9.9984884e-01  2.2473982e-07 ...  1.6708621e-07
  6.4237577e-07  4.7100551e-07]
[ 5.6212690e-07  5.7720405e-07  6.5057570e-06 ...  3.6024649e-02
  6.8448817e-06  5.0523122e-05]]
```

```
#Lets print the image as well
import matplotlib.pyplot as plt
plt.figure(figsize=(1,1))
plt.imshow(testX[0],cmap='gray')
```

```
<matplotlib.image.AxesImage at 0x7f0c8dc067f0>
```



```
np.argmax(testY[0])
```

 9

```
np.argmax(y_pred[0])
```

 9