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**Supervised learning** 

**Group: 3AI-S1** 

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**Assignment 3** 

#### INTRODUCTION:

Link to the full code:

https://colab.research.google.com/drive/1Z1kRe3YIYV K5ZQCAUpxLGX5mx8pJdwe?usp=sharing

# FIRST TEST: we will try testing the number of epochs:

1. Trying 10 epochs

```
2. from tensorflow.python.keras.activations import softmax
3. from tensorflow.python.keras.backend import conv2d
4. model1 =Sequential()
5. #first convolution layer
6. model1.add(Conv2D(32, (3, 3), activation='relu', input shape=(28,
8. model1.add(Conv2D(64, (3, 3), activation='relu'))
9. model1.add(MaxPooling2D((2, 2), strides=(2,2)))
10.
11.
        model1.add(Flatten())
12.
       model1.add(Dense(64, activation='relu'))
       model1.add(Dense(10, activation='softmax'))
from tensorflow.keras.optimizers import SGD
13.
14.
15.
       sqd = SGD(learning rate=0.001, momentum=0.0)
16.
        model1.compile(optimizer= sqd,
17.
                       loss=tf.keras.losses.SparseCategoricalCrossent
  ropy(),
18.
                       metrics=['accuracy'])
19.
20.
       history = model1.fit(train X, train labels, epochs=10,
                              validation data=(test X, test labels),sh
   uffle = True, batch size = 32)
22.
        test loss, test acc = model1.evaluate(test X, test labels,
   verbose=1)
```

- 1. final accuracy if the model :95.40%
- 2. accuracy of the first 5 epochs:
  - 1. Epoch 1/10 :loss: 2.1509 accuracy: 0.3941 val\_loss: 1.7631 val\_accuracy: 0.7074
  - 2. Epoch 2/10 :loss: 0.9498 accuracy: 0.7906 val\_loss: 0.5131 val\_accuracy: 0.8688
  - 3. Epoch 3/10 : loss: 0.4294 accuracy: 0.8808 val\_loss: 0.3451 val\_accuracy: 0.9044
  - 4. Epoch 4/10 :loss: 0.3248 accuracy: 0.9079 val\_loss: 0.2746 val\_accuracy: 0.9217
  - 5. Epoch 5/10 :loss: 0.2718 accuracy: 0.9218 val\_loss: 0.2330 val\_accuracy: 0.9329
- 3. Average training time for each epoch: 16 s

- 4. Average training time for each epoch :16 s
- 5. Learning rate = 0.001
- 6. Number of parameters: 121,930
- 7. Used optimizer: SGD, LR = 0.001, momentum = 0.0
- 8. Layers:
  - 1. Cnn layer followed by relu activation layer and a 2D maxpooling layer with kernel 2\*2 and stride 2\*2
  - 2. Another CNN layer followed by relu activation and a 2D maxpooling layer with kernel 2\*2 and stride 2\*2
  - 3. Flattening layer
  - 4. Fully connected layer with output 64 and relu activation
  - 5. Fully connected layer with output 10 and softmax activation

## 23. Trying 25 epochs:

```
24.
        model1 =Sequential()
25.
        model1.add(Conv2D(32, (3, 3), activation='relu', input shape
  =(28, 28, 1))
26.
        model1.add(MaxPooling2D((2, 2), strides=(2,2)))
27.
        model1.add(Conv2D(64, (3, 3), activation='relu'))
        model1.add(MaxPooling2D((2, 2), strides=(2,2)))
29.
        model1.add(Flatten())
        model1.add(Dense(64, activation='relu'))
32.
        model1.add(Dense(10, activation='softmax'))
33.
34.
        sgd = SGD(learning rate=0.001, momentum=0.0)
35.
36.
        model1.compile(optimizer= sqd,
37.
                       loss=tf.keras.losses.SparseCategoricalCrossent
  ropy(),
38.
                       metrics=['accuracy'])
39.
        history = model1.fit(train X, train labels, epochs=25,
41.
  uffle = True, batch size = 32)
         test loss, test acc = model1.evaluate(test X, test labels,
   verbose=1)
```

- 1. Final accuracy: 97.64 %
- 2. Accuracy of the 5 first epochs:
  - 1. Epoch 1/25 :loss: 2.0716 accuracy: 0.4557 val\_loss: 1.4673 val\_accuracy: 0.7564
  - 2. Epoch 2/25 :loss: 0.7506 accuracy: 0.8267 val\_loss: 0.4291 val\_accuracy: 0.8834
  - 3. Epoch 3/25: loss: 0.3755 accuracy: 0.8927 val\_loss: 0.3154 val\_accuracy: 0.9091
  - 4. Epoch 4/25: loss: 0.3016 accuracy: 0.9117 val\_loss: 0.2641 val\_accuracy: 0.9248
  - 5. Epoch 5/25 :loss: 0.2623 accuracy: 0.9231 val loss: 0.2322 val accuracy: 0.9323

- 3. Learning rate = 0.001
- 4. Average training time :  $10 \sim 11$  seconds
- 5. Number of parameters: 121,930
- 6. Average test time = 4s/step
- 7. Used optimizer: SGD, LR = 0.001, momentum = 0.0
- 8. Layers:
  - 1. Cnn layer followed by relu activation layer and a 2D maxpooling layer with kernel 2\*2 and stride 2\*2
  - 2. Another CNN layer followed by relu activation and a 2D maxpooling layer with kernel 2\*2 and stride 2\*2
  - 3. Flattening layer
  - 4. Fully connected layer with output 64 and relu activation
  - 5. Fully connected layer with output 10 and softmax activation

## 1.3. Trying 40 epochs

```
2. model1 =Sequential()
3.
4. model1.add(Conv2D(32, (3, 3), activation='relu', input shape=(28, 28
   , 1)))
5. model1.add(MaxPooling2D((2, 2), strides=(2,2)))
6. model1.add(Conv2D(64, (3, 3), activation='relu'))
7. modell.add(MaxPooling2D((2, 2), strides=(2, 2)))
9. model1.add(Flatten())
     model1.add(Dense(64, activation='relu'))
11. model1.add(Dense(10, activation='softmax'))
12.
13. sgd = SGD(learning rate=0.001, momentum=0.0)
14.
15. model1.compile(optimizer= sgd,
16.
                    loss=tf.keras.losses.SparseCategoricalCrossentropy(
                    metrics=['accuracy'])
18.
     earlystopping = callbacks.EarlyStopping(monitor ="val loss",
19.
                                              mode ="min", patience = 5
20.
                                               restore best weights = Tr
21.
22.
     time callback = TimeHistory()
23.
     history = model1.fit(train X, train labels, epochs=40,
24.
                          validation data=(test X, test labels), shuffle
    = True, batch size = 32 , callbacks = [time callback, earlystopping])
25. times = time callback.times
```

```
26. test_loss, test_acc = model1.evaluate(test_X, test_labels, verbo
se=1)
```

- 1. final accuracy: 98.27%
- 2. accuracy of the first 5 epochs:
  - 1. Epoch 1/40 :loss: 1.9742 accuracy: 0.4747 val\_loss: 1.2076 val\_accuracy: 0.7127
  - 2. Epoch 2/40 :loss: 0.7109 accuracy: 0.8074 val loss: 0.4496 val accuracy: 0.8745
  - 3. Epoch 3/40: loss: 0.4003 accuracy: 0.8854 val\_loss: 0.3318 val\_accuracy: 0.9051
  - 4. Epoch 4/40:loss: 0.3158 accuracy: 0.9082 val\_loss: 0.2719 val\_accuracy: 0.9229
  - 5. Epoch 5/40: loss: 0.2705 accuracy: 0.9207 val\_loss: 0.2422 val\_accuracy: 0.9303
- 3. Average train time: [13.28525161743164, 10.632070064544678, 10.308592081069946, 11.849564552307129, 12.000101089477539, 10.950177907943726, 10.19032907485962, 9.944242477416992, 10.437545776367188, 10.010447025299072, 10.574686527252197, 10.394061803817749, 10.083669185638428, 9.943841934204102, 10.371716260910034, 10.339104890823364, 10.3578941822052, 10.079188823699951, 10.77367901802063, 10.375660419464111, 10.312971591949463, 10.345721960067749, 10.087323665618896, 10.092099905014038, 10.476072311401367, 10.338266134262085, 10.10399079322815, 10.3533189706802368, 10.408555746078491, 10.375573873519897, 10.026330709457397, 9.979304075241089, 10.386029243469238, 10.372284412384033, 10.53375768661499, 10.353919267654419, 10.202246904373169, 10.498692274093628, 10.572432041168213, 10.520873308181763]
- 4. Average test time : 10~ 11 seconds
- 5. Number of parameters: 121,930
- 6. Learning rate = 0.001
- 7. Used optimizer: SGD, LR = 0.001, momentum = 0.0
- 8. Layers:
  - 1. Cnn layer followed by relu activation layer and a 2D maxpooling layer with kernel 2\*2 and stride 2\*2
  - 2. Another CNN layer followed by relu activation and a 2D maxpooling layer with kernel 2\*2 and stride 2\*2
  - 3. Flattening layer
  - 4. Fully connected layer with output 64 and relu activation
  - 5. Fully connected layer with output 10 and softmax activation

#### FINAL OBSERVATION:

Here we compared the number of epochs to choose the optimal value. We compared 10, 25, and 40. And comparing accuracies of them we find that 40 epochs gives an accuracy that is equal to 98.27% so we will use and test the other parameters with 40 epochs, but I will set an early stopping factor that will stop the process when the model starts overfitting.

We can also observe that the accuracy of the model increases by increasing the number of epochs

- 2. TESTING LEARNING RATES:
- 1. learning rate = 0.01

```
model1.add(Conv2D(32, (3, 3), activation='relu', input shape=(28, 28, 1
model1.add(MaxPooling2D((2, 2), strides=(2,2)))
model1.add(Conv2D(64, (3, 3), activation='relu'))
model1.add(MaxPooling2D((2, 2), strides=(2,2)))
model1.add(Flatten())
model1.add(Dense(64, activation='relu'))
model1.add(Dense(10, activation='softmax'))
sgd = SGD(learning rate=0.01, momentum=0.0)
model1.compile(optimizer= sgd,
              loss=tf.keras.losses.SparseCategoricalCrossentropy(),
              metrics=['accuracy'])
earlystopping = callbacks. EarlyStopping (monitor = "val loss",
                                        mode ="min", patience = 5,
                                         restore best weights = True)
time callback = TimeHistory()
history = model1.fit(train X, train labels, epochs=40,
                    validation data=(test X, test labels), shuffle = Tru
e,batch size = 32 , callbacks = [time callback, earlystopping])
times = time callback.times
test loss, test acc = model1.evaluate(test X, test labels, verbose=1)
```

- 1. Final accuracy of the model :98.93 %
- 2. Accuracy if the first 5 epochs:
  - 1. Epoch 1/40: loss: 0.5620 accuracy: 0.8344 val loss: 0.1737 val accuracy: 0.9492
  - 2. Epoch 2/40: loss: 0.1492 accuracy: 0.9554 val loss: 0.1009 val accuracy: 0.9699
  - 3. Epoch 3/40 :loss: 0.1030 accuracy: 0.9688 val loss: 0.0723 val accuracy: 0.9765
  - 4. Epoch 4/40: loss: 0.0827 accuracy: 0.9742 val loss: 0.0676 val accuracy: 0.9791
  - 5. Epoch 5/40: loss: 0.0701 accuracy: 0.9786 val\_loss: 0.0528 val\_accuracy: 0.9831
- 3. Number of parameters:121.930
- 4. Learning rate = 0.01
- 5. Average train time = [10.66630744934082, 10.488824844360352, 10.095211267471313, 10.666344404220581, 10.631511688232422, 10.509984016418457, 10.522321939468384, 10.4005765914917, 10.063866376876831, 10.131177425384521, 10.495679378509521, 10.164026021957397, 10.49898624420166, 10.532265424728394, 10.149816751480103, 10.598557233810425, 10.082297086715698, 10.162006616592407, 10.152749061584473, 10.657643795013428, 10.105109691619873, 10.536373138427734, 10.449999809265137, 10.120105981826782, 10.465376615524292, 10.19348406791687]
- 6. Average test time =  $10^{\sim} 11$
- 7. Used optimizer: SGD, LR = 0.01, momentum = 0.0
- 8. Layers:
- 1. Cnn layer followed by relu activation layer and a 2D maxpooling layer with kernel 2\*2 and stride 2\*2

- 2. Another CNN layer followed by relu activation and a 2D maxpooling layer with kernel 2\*2 and stride 2\*2
- 3. Flattening layer
- 4. Fully connected layer with output 64 and relu activation
- 5. Fully connected layer with output 10 and softmax activation.

# 2.2. Learning rate = 0.1

```
model1 =Sequential()
modell.add(Conv2D(32, (3, 3), activation='relu', input shape=(28, 28, 1
)))
model1.add(MaxPooling2D((2, 2),strides=(2,2)))
model1.add(Conv2D(64, (3, 3), activation='relu'))
model1.add(MaxPooling2D((2, 2),strides=(2,2)))
model1.add(Flatten())
model1.add(Dense(64, activation='relu'))
model1.add(Dense(10, activation='softmax'))
sqd = SGD(learning rate=0.1, momentum=0.0)
model1.compile(optimizer= sgd,
              loss=tf.keras.losses.SparseCategoricalCrossentropy(),
              metrics=['accuracy'])
earlystopping = callbacks.EarlyStopping(monitor ="val loss",
                                        mode ="min", patience = 5,
                                        restore best weights = True)
time callback = TimeHistory()
history = model1.fit(train X, train labels, epochs=40,
                    validation data=(test X, test labels), shuffle = Tru
e,batch size = 32 , callbacks =[time callback,earlystopping])
times = time callback.times
test loss, test acc = model1.evaluate(test X, test labels, verbose=1)
```

- 1. Final accuracy of the model: 99.07
- 2. Accuracy of the first 5 epochs:
  - 1. Epoch 1/40:loss: 0.1779 accuracy: 0.9438 val\_loss: 0.0575 val\_accuracy: 0.9803
  - 2. Epoch 2/40:loss: 0.0553 accuracy: 0.9830 val\_loss: 0.0391 val\_accuracy: 0.9871
  - 3. Epoch 3/40:loss: 0.0384 accuracy: 0.9879 val\_loss: 0.0394 val\_accuracy: 0.9864
  - 4. Epoch 4/40: loss: 0.0296 accuracy: 0.9911 val\_loss: 0.0359 val\_accuracy: 0.9883
  - 5. Epoch 5/40:loss: 0.0237 accuracy: 0.9924 val\_loss: 0.0302 val\_accuracy: 0.9907
- 3. Average train time: [10.947302103042603, 10.109866380691528, 11.027601480484009, 10.799046277999878, 10.092986822128296, 10.398900985717773, 10.477643251419067, 10.040547370910645, 10.456888198852539, 10.107140302658081, 10.492445945739746]
- 4. Average test time: 10 ~ 11 seconds

- 5. Number of parameters: 121,930
- 6. Learning rate: 0.1
- 7. Used optimizer: SGD, LR = 0.01, momentum = 0.0
- 8. Layers:
  - 1. Cnn layer followed by relu activation layer and a 2D maxpooling layer with kernel 2\*2 and stride 2\*2
- 2. Another CNN layer followed by relu activation and a 2D maxpooling layer with kernel 2\*2 and stride 2\*2
  - 3. Flattening layer
  - 4. Fully connected layer with output 64 and relu activation
  - 5. Fully connected layer with output 10 and softmax activation.

## 2.3. Learning rate = 0.003:

```
model1 =Sequential()
modell.add(Conv2D(32, (3, 3), activation='relu', input shape=(28, 28, 1
)))
model1.add(MaxPooling2D((2, 2),strides=(2,2)))
model1.add(Conv2D(64, (3, 3), activation='relu'))
model1.add(MaxPooling2D((2, 2),strides=(2,2)))
model1.add(Flatten())
model1.add(Dense(64, activation='relu'))
model1.add(Dense(10, activation='softmax'))
sqd = SGD(learning rate=0.003, momentum=0.0)
model1.compile(optimizer= sgd,
              loss=tf.keras.losses.SparseCategoricalCrossentropy(),
              metrics=['accuracy'])
earlystopping = callbacks.EarlyStopping(monitor ="val loss",
                                        mode ="min", patience = 5,
                                        restore best weights = True)
time callback = TimeHistory()
history = model1.fit(train X, train labels, epochs=40,
                    validation data=(test X, test labels), shuffle = Tru
e,batch size = 32 , callbacks =[time callback,earlystopping])
times = time callback.times
```

- 1. Final accuracy of the model: 98.81 %
- 2. Accuracy of the first 5 epochs:
  - 1. Epoch 1/40:loss: 1.1887 accuracy: 0.6259 val\_loss: 0.3600 val\_accuracy: 0.8965

- 2. Epoch 2/40 :loss: 0.3083 accuracy: 0.9075 val\_loss: 0.2367 val\_accuracy: 0.9315
- 3. Epoch 3/40:loss: 0.2218 accuracy: 0.9345 val\_loss: 0.1715 val\_accuracy: 0.9505
- 4. Epoch 4/40: loss: 0.1720 accuracy: 0.9491 val\_loss: 0.1393 val\_accuracy: 0.9578
- 5. Epoch 5/40: loss: 0.1413 accuracy: 0.9584 val\_loss: 0.1137 val\_accuracy: 0.9679
- 3. Number of parameters: 121,930
- 4. Learning rate: 0.003
- 5. Average train time: [10.648336410522461, 10.470914840698242, 10.317973613739014, 10.250239372253418, 10.331140041351318, 10.368586778640747, 9.954809427261353, 10.307901382446289, 10.501755475997925, 9.986394882202148, 10.352505922317505, 10.800374746322632, 10.25293779373169, 10.060513973236084, 10.007049322128296, 10.453199863433838, 10.130411148071289, 10.090394258499146, 9.967403888702393, 10.494374513626099, 9.993850231170654, 10.417911529541016, 12.686509132385254, 10.030638933181763, 9.98386836051941, 10.323788404464722, 9.921845197677612, 10.339212656021118, 9.969069719314575, 9.973387718200684, 10.396514892578125, 10.425524234771729, 10.17065167427063, 10.054506063461304, 10.036363363265991, 10.488706111907959, 10.535224199295044, 10.08469271659851, 10.498995542526245, 10.511680364608765]
- 6. Average test time: 10 ~11 seconds
- 7. Used optimizer: SGD, LR = 0.003, momentum = 0.0
- 8. Layers:
  - Cnn layer followed by relu activation layer and a 2D maxpooling layer with kernel 2\*2 and stride 2\*2
  - Another CNN layer followed by relu activation and a 2D maxpooling layer with kernel 2\*2 and stride 2\*2
  - Flattening layer
  - Fully connected layer with output 64 and relu activation
  - Fully connected layer with output 10 and softmax activation.

## 2.4. Learning rate = 0.005

```
model1 =Sequential()
# testing learning rate = 0.005
model1.add(Conv2D(32, (3, 3), activation='relu', input_shape=(28, 28, 1)))
model1.add(MaxPooling2D((2, 2), strides=(2,2)))
model1.add(Conv2D(64, (3, 3), activation='relu'))
model1.add(MaxPooling2D((2, 2), strides=(2,2)))

model1.add(Flatten())
model1.add(Dense(64, activation='relu'))
model1.add(Dense(10, activation='relu'))
sgd = SGD(learning rate=0.005, momentum=0.0)
```

- 1. Final accuracy of the model: 98. %90
- 2. Accuracy of the first 5 epochs:
  - 1. Epoch 1/40:loss: 0.7580 accuracy: 0.7807 val\_loss: 0.3076 val\_accuracy: 0.9096
  - 2. Epoch 2/40: loss: 0.2151 accuracy: 0.9357 val\_loss: 0.1513 val\_accuracy: 0.9544
  - 3. Epoch 3/40:loss: 0.1489 accuracy: 0.9556 val\_loss: 0.1075 val\_accuracy: 0.9667
  - 4. Epoch 4/40:loss: 0.1177 accuracy: 0.9650 val\_loss: 0.0930 val\_accuracy: 0.9723
  - 5. Epoch 5/40:loss: 0.0997 accuracy: 0.9708 val loss: 0.0787 val accuracy: 0.9759
- 3. Learning rate = 0.005
- 4. Used optimizer: SGD, LR = 0.003, momentum = 0.0
- 5. Number of parameters: 121.930
- Average train time: [6.4069600105285645, 5.524645805358887, 6.771639585494995, 5.5406951904296875, 5.589333772659302, 5.54652214050293, 5.541147708892822, 5.558661460876465, 5.595424175262451, 5.544501066207886, 5.569594860076904, 5.555826902389526, 5.561650991439819, 5.577897071838379, 5.6011059284210205, 5.523564100265503, 5.623640537261963, 5.718517303466797, 5.6122376918792725, 5.6008710861206055, 5.5956079959869385, 5.583516597747803, 5.628449440002441, 5.507747650146484, 5.574894428253174, 5.632693290710449, 5.52852988243103, 5.612900495529175, 5.74003005027771, 5.595170736312866, 5.553348779678345, 5.582526206970215, 5.595680236816406, 5.547945022583008, 5.68924880027771, 5.60068154335022, 5.749489784240723, 5.590587139129639, 5.681508302688599, 5.658182621002197]
- 7. Average test time: 5 ~ 6 seconds
- 8. Layers:
  - Cnn layer followed by relu activation layer and a 2D maxpooling layer with kernel 2\*2 and stride 2\*2
  - Another CNN layer followed by relu activation and a 2D maxpooling layer with kernel 2\*2 and stride 2\*2
  - Flattening layer
  - Fully connected layer with output 64 and relu activation
  - Fully connected layer with output 10 and softmax activation.

# Final observation:

we tested here 4 different learning rates: 0.01, 0.1, 0.003, 0.005, we can observe that the learning rate 0.1 gives the highest accuracy = 99.07, so we will use 0.1 as a learning rate when using SGD as an optimizer.

#### 3. TESTING DIFFERENT Architectures:

#### 3.1. 1 CNN AND 1 FC

```
model1 =Sequential()
model1.add(Conv2D(32, (3, 3), activation='relu', input shape=(28, 28, 1
model1.add(MaxPooling2D((2, 2), strides=(2,2)))
model1.add(Flatten())
model1.add(Dense(10, activation='softmax'))
sgd = SGD(learning rate=0.1, momentum=0.0)
model1.compile(optimizer= sgd,
              loss=tf.keras.losses.SparseCategoricalCrossentropy(),
              metrics=['accuracy'])
earlystopping = callbacks.EarlyStopping(monitor ="val loss",
                                        mode ="min", patience = 5,
                                         restore best weights = True)
time callback = TimeHistory()
history = model1.fit(train X, train labels, epochs=40,
                    validation_data=(test_X, test_labels), shuffle = Tru
e,batch size = 32 , callbacks = [time callback,earlystopping])
times = time callback.times
test loss, test acc = model1.evaluate(test X, test labels, verbose=1)
```

- 1. Final accuracy of the model: 98.45
- 2. Accuracy of the first 5 epochs:
  - 1. Epoch 1/40:loss: 0.2722 accuracy: 0.9195 val\_loss: 0.1197 val\_accuracy: 0.9660
  - 2. Epoch 2/40:loss: 0.0975 accuracy: 0.9725 val\_loss: 0.0758 val\_accuracy: 0.9757
  - 3. Epoch 3/40:loss: 0.0733 accuracy: 0.9784 val\_loss: 0.0609 val\_accuracy: 0.9810
  - 4. Epoch 4/40:loss: 0.0611 accuracy: 0.9822 val\_loss: 0.0622 val\_accuracy: 0.9800
  - 5. Epoch 5/40:loss: 0.0536 accuracy: 0.9838 val loss: 0.0590 val accuracy: 0.9806
- 3. Number of parameters: 54,410
- 4. Average train time: [8.484246253967285, 7.416532516479492, 7.487035036087036, 7.415877103805542, 7.5231146812438965, 7.9372618198394775, 7.915063381195068, 7.440031051635742, 7.917072534561157, 7.481266021728516, 7.968287706375122, 7.526801347732544, 8.035681009292603]
- 5. Average test time: 7~8 seconds

- 6. Learning rate = 0.1
- 7. Used optimizer: SGD, LR = 0.1, momentum = 0.0
- 8. Layers:
  - A CNN layer with a relu activation function and a maxpooling2D with kernel size 2\*2 and strides 2\*2
  - Flattening layer
  - A fully connected layer with 10 outputs and softmax activation

#### 3.2. 2CNN AND 3FC

```
model1 =Sequential()
# testing learning rate = 0.1
model1.add(Conv2D(32, (3, 3), activation='relu', input shape=(28, 28, 1
)))
model1.add(MaxPooling2D((2, 2), strides=(2,2)))
model1.add(Conv2D(64, (3, 3), activation='relu'))
model1.add(MaxPooling2D((2, 2),strides=(2,2)))
model1.add(Flatten())
model1.add(Dense(64, activation='relu'))
model1.add(Dense(32, activation='relu'))
model1.add(Dense(10, activation='softmax'))
sgd = SGD(learning rate=0.1, momentum=0.0)
model1.compile(optimizer= sgd,
              loss=tf.keras.losses.SparseCategoricalCrossentropy(),
              metrics=['accuracy'])
earlystopping = callbacks.EarlyStopping(monitor ="val loss",
                                        mode ="min", patience = 5,
                                        restore best weights = True)
time callback = TimeHistory()
history = model1.fit(train X, train labels, epochs=40,
                    validation data=(test X, test labels), shuffle = Tru
e,batch size = 32 , callbacks =[time callback,earlystopping])
times = time callback.times
test loss, test acc = model1.evaluate(test X, test labels, verbose=1)
```

- 1. Layers:
  - 1. 2 CNN layers with a relu activation function and followed by maxpooling 2D with kernel size 2\*2 and strides 2\*2
  - 2. Flattening layer
  - 3. 2 Fully connected layers with relu activation functions
  - 4. Another fully connected layer with softmax activation function
- 2. Final accuracy of the model: 99.18
- 3. Accuracy of the first 5 epochs:

- 1. Epoch 1/40:loss: 0.2088 accuracy: 0.9340 val\_loss: 0.0618 val\_accuracy: 0.9802
- 2. Epoch 2/40:loss: 0.0580 accuracy: 0.9822 val\_loss: 0.0420 val\_accuracy: 0.9871
- 3. Epoch 3/40:loss: 0.0390 accuracy: 0.9882 val\_loss: 0.0343 val\_accuracy: 0.9895
- 4. Epoch 4/40:loss: 0.0295 accuracy: 0.9906 val\_loss: 0.0355 val\_accuracy: 0.9886
- 5. Epoch 5/40 loss: 0.0243 accuracy: 0.9922 val\_loss: 0.0297 val\_accuracy: 0.9898
- 4. Average train time: [6.328695058822632, 5.764863014221191, 5.6650331020355225, 5.74968147277832, 5.780593156814575, 5.768248081207275, 5.766294717788696, 5.742184400558472, 5.763494253158569, 5.757632493972778, 5.755081653594971, 5.807462215423584, 5.757145643234253, 5.727740287780762]
- 5. Average test time: 5~7 seconds
- 6. Learning rate = 0.1
- 7. Used optimizer: SGD, LR = 0.1, momentum = 0.0
- 8. Number of parameters: 123,690

#### 3.3. 3CNN AND 3FC:

```
model1 =Sequential()
# testing learning rate = 0.1 , trying 3CNN and 3FC
modell.add(Conv2D(32, (3, 3), activation='relu', input shape=(28, 28, 1
)))
model1.add(MaxPooling2D((2, 2), strides=(2,2)))
model1.add(Conv2D(64, (3, 3), activation='relu'))
model1.add(MaxPooling2D((2, 2),strides=(2,2)))
model1.add(Conv2D(32, (3, 3), activation='relu'))
model1.add(MaxPooling2D((2, 2),strides=(2,2)))
model1.add(Flatten())
model1.add(Dense(64, activation='relu'))
model1.add(Dense(32, activation='relu'))
model1.add(Dense(10, activation='softmax'))
sgd = SGD(learning rate=0.1, momentum=0.0)
model1.compile(optimizer= sgd,
              loss=tf.keras.losses.SparseCategoricalCrossentropy(),
earlystopping = callbacks.EarlyStopping(monitor ="val loss",
                                        mode ="min", patience = 5,
                                        restore_best_weights = True)
time callback = TimeHistory()
history = model1.fit(train X, train labels, epochs=40,
                    validation data=(test X, test labels), shuffle = Tru
e,batch size = 32 , callbacks =[time callback,earlystopping])
```

```
times = time_callback.times
test_loss, test_acc = model1.evaluate(test_X, test_labels, verbose=1)
```

- 1. Layers:
  - 1. 3 CNN layers with a relu activation function and followed by maxpooling 2D with kernel size 2\*2 and strides 2\*2
  - 2. Flattening layer
  - 3. 2 Fully connected layers with relu activation functions
  - 4. Another fully connected layer with softmax activation
- 2. Accuracy of the final epoch: 98.59%
- 3. Accuracy of the first 5 epochs:
  - 1. Epoch 1/40:loss: 0.3402 accuracy: 0.8899 val\_loss: 0.0895 val\_accuracy: 0.9743
  - 2. Epoch 2/40:loss: 0.0864 accuracy: 0.9741 val\_loss: 0.0806 val\_accuracy: 0.9752
  - 3. Epoch 3/40:loss: 0.0609 accuracy: 0.9813 val loss: 0.0530 val accuracy: 0.9854
  - 4. Epoch 4/40:loss: 0.0492 accuracy: 0.9849 val loss: 0.0613 val accuracy: 0.9808
  - 5. Epoch 5/40:loss: 0.0412 accuracy: 0.9872 val\_loss: 0.0562 val\_accuracy: 0.9831
- 4. Number of parameters:1,802
- 5. Learning rate =0.1
- 6. Used optimizer: SGD, LR = 0.1, momentum = 0.0
- 7. Average train time: [6.6867616176605225, 6.2012739181518555, 6.1085498332977295, 6.056229114532471, 6.767364740371704, 6.054199695587158, 6.107012987136841, 6.086791038513184, 6.102833032608032, 6.042593002319336, 6.016862630844116]
- 8. Average test time: 6 seconds

#### 3.4. 3CNN AND 2FC

```
model1 =Sequential()
model1.add(Conv2D(32, (3, 3), activation='relu', input shape=(28, 28, 1
))))
model1.add(MaxPooling2D((2, 2),strides=(2,2)))
model1.add(Conv2D(64, (3, 3), activation='relu'))
model1.add(MaxPooling2D((2, 2),strides=(2,2)))
model1.add(Conv2D(32, (3, 3), activation='relu'))
model1.add(MaxPooling2D((2, 2), strides=(2,2)))
model1.add(Flatten())
model1.add(Dense(32, activation='relu'))
model1.add(Dense(10, activation='softmax'))
sgd = SGD(learning_rate=0.1, momentum=0.0)
model1.compile(optimizer= sgd,
              loss=tf.keras.losses.SparseCategoricalCrossentropy(),
              metrics=['accuracy'])
earlystopping = callbacks.EarlyStopping(monitor ="val loss",
                                        mode ="min", patience = 5,
```

- 1. Layers:
  - 3 CNN layers with a relu activation function and followed by maxpooling 2D with kernel size 2\*2 and strides 2\*2
  - Flattening layer
  - Fully connected layer with relu activation
  - Fully connected layer with softmax
- 2. Accuracy of the final epoch :98.58
- 3. Number of parameters: 38,666
- 4. Accuracy of the first 5 epochs:
  - 1. Epoch 1/40:loss: 0.3089 accuracy: 0.9020 val loss: 0.1018 val accuracy: 0.9687
  - 2. Epoch 2/40:loss: 0.0844 accuracy: 0.9740 val\_loss: 0.0607 val\_accuracy: 0.9811
  - 3. Epoch 3/40:loss: 0.0614 accuracy: 0.9814 val\_loss: 0.0586 val\_accuracy: 0.9834
  - 4. Epoch 4/40:loss: 0.0483 accuracy: 0.9850 val loss: 0.0603 val accuracy: 0.9841
  - 5. Epoch 5/40:loss: 0.0414 accuracy: 0.9872 val loss: 0.0607 val accuracy: 0.9832
- 5. Average train time: [9.056811332702637, 6.00836968421936, 5.844502925872803, 6.525131464004517, 5.908543109893799, 5.901589393615723, 5.83287239074707, 5.832604169845581, 5.852531909942627, 5.828242301940918, 6.560439586639404, 6.2201774120330811
- 6. Average test time: 5~ 6 seconds
- 7. Learning rate =0.1
- 8. Used optimizer: SGD, LR = 0.1, momentum = 0.0

#### FINAL OBSERVATION:

WE can notice that the model that includes 2 convolution layers and 3 fully connected layers has given higher accuracy than the other 3 tested models as it equals 99.18 so we will keep working with that model.

In addition, it is noticeable that models that gave higher accuracies have included bigger number of fully connected layers

# 4- TESTING DIFFERENT BATCH SIZE

#### 4.1. Batch size = 64

```
5. model1 =Sequential()
6. # testing learning rate = 0.1
7. # 2 CNN and 3 FC
8. # batch size = 64
```

```
9. model1.add(Conv2D(32, (3, 3), activation='relu', input shape=(28
   , 28, 1)))
      model1.add(MaxPooling2D((2, 2), strides=(2,2)))
11.
       model1.add(Conv2D(64, (3, 3), activation='relu'))
      model1.add(MaxPooling2D((2, 2), strides=(2,2)))
12.
13.
14.
      model1.add(Flatten())
15.
      model1.add(Dense(32, activation='relu'))
16.
17.
      model1.add(Dense(10, activation='softmax'))
18.
19.
       sgd = SGD(learning rate=0.1, momentum=0.0)
20.
21.
      model1.compile(optimizer= sgd,
22.
                     loss=tf.keras.losses.SparseCategoricalCrossent
  ropy(),
23.
                     metrics=['accuracy'])
24.
       earlystopping = callbacks.EarlyStopping(monitor = "val loss",
25.
                                                mode ="min", patienc
26.
                                                restore best weights
27.
28.
      time callback = TimeHistory()
29.
      history = model1.fit(train X, train labels, epochs=40,
                           validation data=(test X, test labels),sh
30.
  uffle = True,batch size = 64 , callbacks =[time callback,earlyst
  opping])
31.
       times = time callback.times
       test loss, test acc = model1.evaluate(test X, test labels,
  verbose=1)
```

- 1. accuracy of the first 5 layers:
  - Epoch 1/40:loss: 0.2903 accuracy: 0.9075 val\_loss: 0.0708 val\_accuracy: 0.9785
  - Epoch 2/40:loss: 0.0686 accuracy: 0.9784 val\_loss: 0.0447 val\_accuracy: 0.9858
  - Epoch 3/40:loss: 0.0479 accuracy: 0.9852 val loss: 0.0413 val accuracy: 0.9865
  - Epoch 4/40:loss: 0.0379 accuracy: 0.9879 val\_loss: 0.0328 val\_accuracy: 0.9885
  - Epoch 5/40:loss: 0.0299 accuracy: 0.9905 val\_loss: 0.0310 val\_accuracy: 0.9886
- 2. final accuracy of the model: 99.19 %
- 3. layers:
  - 2 CNN layers with a relu activation function and followed by maxpooling 2D with kernel size 2\*2 and strides 2\*2
  - Flattening layer
  - 2 Fully connected layers with relu activation functions
  - Another fully connected layer with softmax activation function
- 4. Average train time: [4.016432762145996, 3.4067440032958984, 3.7209880352020264, 3.364039897918701, 3.3387527465820312, 3.359880208969116, 3.801670789718628,

```
3.6276445388793945, 3.6455094814300537, 3.5915675163269043, 3.55423903465271, 3.357398748397827, 3.322221040725708, 3.3238909244537354, 3.3913235664367676, 3.455738067626953, 3.7189836502075195, 3.399812698364258]
```

- 5. Average test time: 3 ~ 4 seconds6. Number of parameters: 123,690
- 7. Learning rate = 0.1
- 8. Used optimizer: SGD, LR = 0.1, momentum = 0.0

#### 4.2. Batch size = 96

```
5. model1 =Sequential()
6. # testing learning rate = 0.1
8. # batch size = 96
9. model1.add(Conv2D(32, (3, 3), activation='relu', input shape=(28
   , 28, 1)))
10.
       model1.add(MaxPooling2D((2, 2), strides=(2, 2)))
11.
12.
       model1.add(MaxPooling2D((2, 2), strides=(2,2)))
13.
14.
       model1.add(Flatten())
15.
       model1.add(Dense(64, activation='relu'))
16.
     model1.add(Dense(32, activation='relu'))
17.
19.
       sgd = SGD(learning rate=0.1, momentum=0.0)
20.
21.
       model1.compile(optimizer= sgd,
22.
                     loss=tf.keras.losses.SparseCategoricalCrossent
  ropy(),
23.
                     metrics=['accuracy'])
24.
       earlystopping = callbacks. EarlyStopping (monitor = "val loss",
25.
                                                mode ="min", patienc
26.
                                                restore best weights
27.
28.
       time callback = TimeHistory()
29.
       history = model1.fit(train X, train labels, epochs=40,
                            validation data=(test X, test labels),sh
  uffle = True, batch size = 96 , callbacks = [time callback, earlyst
  opping])
       times = time callback.times
       test loss, test acc = model1.evaluate(test X, test labels,
 verbose=1)
```

- 1. accuracy of the first 5 layers:
- Epoch 1/4: loss: 0.3860 accuracy: 0.8763 val loss: 0.1010 val accuracy: 0.9678

- Epoch 2/40:loss: 0.0862 accuracy: 0.9729 val\_loss: 0.0884 val\_accuracy: 0.9701
- Epoch 3/40:loss: 0.0609 accuracy: 0.9807 val loss: 0.0547 val accuracy: 0.9820
- Epoch 4/40:loss: 0.0464 accuracy: 0.9859 val\_loss: 0.0449 val\_accuracy: 0.9850
- Epoch 5/40:loss: 0.0374 accuracy: 0.9881 val\_loss: 0.0329 val\_accuracy: 0.9894
  - 2. final accuracy of the model: 99.03 %
  - 3. layers:
    - 2 CNN layers with a relu activation function and followed by maxpooling 2D with kernel size 2\*2 and strides 2\*2
    - Flattening layer
    - 2 Fully connected layers with relu activation functions
    - Another fully connected layer with softmax activation function
  - Average train time: [2.959050178527832, 2.2978005409240723, 2.601702928543091, 2.311624765396118, 2.168905019760132, 2.2645761966705322, 2.32275128364563, 2.289583683013916, 2.210627317428589, 2.1708009243011475, 2.188220739364624, 2.693422794342041, 2.355470895767212]
  - 5. Average test time: 2 seconds
  - 6. Number of parameters: 123,690
  - 7. Learning rate = 0.1
  - 8. Used optimizer: SGD, LR = 0.1, momentum = 0.0

#### **Full observation:**

We can notice that batch size 64 gives higher accuracy than 32 and 96 this accuracy is equal to 99.19%, so we will keep testing with batch size =64

- 5- Testing different activation functions:
  - 5.1. Sigmoid activation function:

```
6. model1 =Sequential()
8. # 2 CNN and 3 FC
9. # batch size = 64 , activation function will be sigmoid
10. model1.add(Conv2D(32, (3, 3), activation='sigmoid', input sh
  ape=(28, 28, 1))
11. model1.add(MaxPooling2D((2, 2), strides=(2,2)))
12.
     model1.add(Conv2D(64, (3, 3), activation='sigmoid'))
13. model1.add(MaxPooling2D((2, 2), strides=(2,2)))
14.
15.
     model1.add(Flatten())
     model1.add(Dense(64, activation='sigmoid'))
16.
     model1.add(Dense(32, activation='sigmoid'))
17.
     model1.add(Dense(10, activation='softmax'))
18.
19.
     sgd = SGD(learning rate=0.1, momentum=0.0)
```

```
21.
22.
     model1.compile(optimizer= sqd,
23.
                    loss=tf.keras.losses.SparseCategoricalCrossent
  ropy(),
24.
                    metrics=['accuracy'])
25.
     earlystopping = callbacks.EarlyStopping(monitor = "val loss",
26.
                                               mode ="min", patienc
27.
                                               restore best weights
28.
29.
     time callback = TimeHistory()
30.
     history = model1.fit(train X, train labels, epochs=40,
                          validation data=(test X, test labels),sh
  uffle = True, batch size = 64 , callbacks = [time callback, earlys
   topping])
     times = time callback.times
     test loss, test acc = model1.evaluate(test X, test labels,
```

- 1. accuracy of the final epoch: 98.48
- 2. accuracy of the first 5 epochs:
  - 1. Epoch 1/40:loss: 2.3064 accuracy: 0.1068 val loss: 2.3096 val accuracy: 0.1028
  - 2. Epoch 2/40:loss: 2.3038 accuracy: 0.1078 val\_loss: 2.3020 val\_accuracy: 0.1135
  - 3. Epoch 3/40:loss: 2.3020 accuracy: 0.1124 val loss: 2.3026 val accuracy: 0.1135
  - 4. Epoch 4/40:loss: 2.2940 accuracy: 0.1297 val\_loss: 2.2776 val\_accuracy: 0.2497
  - 5. Epoch 5/40:loss: 1.7290 accuracy: 0.4425 val loss: 0.9315 val accuracy: 0.7372
- 3. Number of parameters: 123,690
- 4. Learning rate = 0.1
- 5. Used optimizer: SGD, LR = 0.1, momentum = 0.0
- 6. layers:
  - 2 CNN layers with a relu activation function and followed by maxpooling 2D with kernel size 2\*2 and strides 2\*2
  - Flattening layer
  - 2 Fully connected layers with relu activation functions
  - Another fully connected layer with softmax activation function
- Average train time: [4.367655992507935, 3.720479726791382, 3.338361978530884, 3.6639647483825684, 3.739867687225342, 3.9349794387817383, 3.557138204574585, 3.3655545711517334, 3.6426117420196533, 3.6802406311035156, 3.6767983436584473, 3.5736868381500244, 3.6751344203948975, 3.640183687210083, 3.734370708465576, 3.7161905765533447, 4.379518747329712, 3.3885579109191895, 3.7565677165985107, 3.66469407081604, 3.3001341819763184, 3.515160322189331, 3.6699423789978027, 4.0349321365356445, 4.885350704193115, 5.046709060668945, 3.7519640922546387, 4.367229223251343, 3.5672640800476074, 3.3498902320861816, 3.328296661376953, 3.40008544921875, 3.343167304992676, 3.4531607627868652, 4.099081993103027, 4.829409837722778, 3.874272584915161, 4.423378944396973, 3.916652202606201, 3.6958553791046143]
- 8. Average test time: 3 ~ 4 seconds

#### 5.2. Tanh activation function

```
6. model1 =Sequential()
7. # testing learning rate = 0.1
8. # 2 CNN and 3 FC
9. # batch size = 64 , activation function will be tanh
10. model1.add(Conv2D(32, (3, 3), activation='tanh', input shape
  =(28, 28, 1))
11. model1.add(MaxPooling2D((2, 2), strides=(2,2)))
12.
     model1.add(Conv2D(64, (3, 3), activation='tanh'))
     model1.add(MaxPooling2D((2, 2), strides=(2,2)))
14.
15.
     model1.add(Flatten())
16.
     model1.add(Dense(64, activation='tanh'))
17.
     model1.add(Dense(32, activation='tanh'))
18.
     model1.add(Dense(10, activation='softmax'))
20.
     sgd = SGD(learning rate=0.1, momentum=0.0)
21.
22.
     model1.compile(optimizer= sgd,
23.
                    loss=tf.keras.losses.SparseCategoricalCrossent
  ropy(),
24.
25.
     earlystopping = callbacks.EarlyStopping(monitor ="val loss",
26.
                                              mode ="min", patienc
27.
                                              restore best weights
     time callback = TimeHistory()
     history = model1.fit(train X, train labels, epochs=40,
30.
31.
                          validation data=(test X, test labels),sh
  uffle = True, batch size = 64 , callbacks = [time callback, earlys
  topping])
     times = time callback.times
     test loss, test acc = model1.evaluate(test X, test labels,
  verbose=1)
```

- 1. accuracy of the first five epochs:
  - 1. Epoch 1/40:loss: 0.2344 accuracy: 0.9365 val\_loss: 0.0736 val\_accuracy: 0.9785
  - 2. Epoch 2/40:loss: 0.0673 accuracy: 0.9805 val loss: 0.0517 val accuracy: 0.9852
  - 3. Epoch 3/40:loss: 0.0467 accuracy: 0.9862 val\_loss: 0.0413 val\_accuracy: 0.9868
  - 4. Epoch 4/40:loss: 0.0348 accuracy: 0.9902 val loss: 0.0374 val accuracy: 0.9877
  - 5. Epoch 5/40:loss: 0.0278 accuracy: 0.9920 val loss: 0.0365 val accuracy: 0.9891
- 2. Accuracy of the final epoch: 99.06
- 3. Number of parameters: 123,690
- Average train time: [4.3000712394714355, 3.316728115081787, 3.315232038497925, 3.4011199474334717, 3.6583423614501953, 3.651688814163208, 3.6783761978149414,

- 3.6783487796783447, 3.642878293991089, 3.6715195178985596, 3.7345173358917236, 4.851062297821045, 3.7517940998077393, 4.050658464431763, 3.4767770767211914, 3.6623072624206543, 3.714305877685547, 3.3614184856414795]
- 5. Average test time: 3 ~ 4 seconds
- 6. Learning rate = 0.1
- 7. Used optimizer: SGD, LR = 0.1, momentum = 0.0
- 8. layers:
  - 2 CNN layers with a relu activation function and followed by maxpooling 2D with kernel size 2\*2 and strides 2\*2
  - Flattening layer
  - 2 Fully connected layers with relu activation functions
  - Another fully connected layer with softmax activation function

# 5.3 testing on softplus:

```
6. model1 =Sequential()
7. # testing learning rate = 0.1
8. # 2 CNN and 3 FC
9. # batch size = 64 , activation function will be softplus
10. model1.add(Conv2D(32, (3, 3), activation='softplus', input s
  hape=(28, 28, 1))
11.
     model1.add(MaxPooling2D((2, 2), strides=(2,2)))
     model1.add(Conv2D(64, (3, 3), activation='softplus'))
13.
     model1.add(MaxPooling2D((2, 2), strides=(2,2)))
14.
15.
     model1.add(Flatten())
16.
     model1.add(Dense(64, activation='softplus'))
17.
     model1.add(Dense(32, activation='softplus'))
     model1.add(Dense(10, activation='softmax'))
18.
19.
20.
     sqd = SGD(learning rate=0.1, momentum=0.0)
21.
22.
     model1.compile(optimizer= sgd,
23.
                   loss=tf.keras.losses.SparseCategoricalCrossent
  ropy(),
24.
                   metrics=['accuracy'])
     earlystopping = callbacks. EarlyStopping (monitor = "val loss",
26.
                                              mode ="min", patienc
27.
                                              restore best weights
28.
29.
     time callback = TimeHistory()
30.
     history = model1.fit(train X, train labels, epochs=40,
31.
                          validation data=(test X, test labels),sh
  uffle = True, batch size = 64 , callbacks = [time callback, earlys
  topping])
```

```
32. times = time_callback.times
33. test_loss, test_acc = model1.evaluate(test_X, test_labels, verbose=1)
```

- 1. accuracy of the first 5 epochs:
  - Epoch 1/40:loss: 2.0021 accuracy: 0.2448 val\_loss: 0.5331 val\_accuracy: 0.8260
  - Epoch 2/40:loss: 0.1802 accuracy: 0.9446 val\_loss: 0.1132 val\_accuracy: 0.9635
  - Epoch 3/40:loss: 0.1028 accuracy: 0.9694 val\_loss: 0.0788 val\_accuracy: 0.9753
  - Epoch 4/40:loss: 0.0729 accuracy: 0.9773 val\_loss: 0.0977 val\_accuracy: 0.9674
  - Epoch 5/40:loss: 0.0582 accuracy: 0.9818 val\_loss: 0.0748 val\_accuracy: 0.9755
- 2. Accuracy of the final epoch: 98.68
- 3. Number of parameters: 123,690
- 4. Learning rate = 0.1
- 5. Average train time: [4.391051292419434, 3.716292142868042,
  - 3.4012131690979004, 3.3829143047332764, 3.4249937534332275,
  - 3.7440598011016846, 3.762845993041992, 3.670835494995117,
  - 3.705263137817383, 3.3708441257476807, 3.508181095123291,
  - 3.6963117122650146, 3.4666929244995117, 3.392502546310425,
  - 3.381305456161499, 3.38926362991333]
- 6. Average test time: 3 ~ 4 seconds
- 7. Layers:
  - 2 CNN layers with a relu activation function and followed by maxpooling 2D with kernel size 2\*2 and strides 2\*2
  - Flattening layer
  - 2 Fully connected layers with relu activation functions
  - Another fully connected layer with softmax activation function
- 8. Used optimizer: SGD, LR = 0.1, momentum = 0.0

## FINAL OBSERVATION:

All of the activation functions gave a kind of close accuracies but the heighest accuracy was given by relu activation function with an accuracy = 99.19 % so we will keep testing with it

# 6. TESTING WITH DIFFERENT OPTIMIZERS: 6.1. ADAM OPTIMIZER:

```
7. model1 =Sequential()
8. # testing learning rate = 0.1
9. # 2 CNN and 3 FC
10. # batch size = 64 , # activation function = relu , adam optimizer
```

```
11. model1.add(Conv2D(32, (3, 3), activation='relu', input shape=(28,
    28, 1)))
12. model1.add(MaxPooling2D((2, 2), strides=(2,2)))
13. model1.add(Conv2D(64, (3, 3), activation='relu'))
14. model1.add(MaxPooling2D((2, 2), strides=(2,2)))
15.
16. model1.add(Flatten())
17. model1.add(Dense(64, activation='relu'))
18. model1.add(Dense(32, activation='relu'))
19. model1.add(Dense(10, activation='softmax'))
20.
21. model1.compile(optimizer= 'adam',
22.
                  loss=tf.keras.losses.SparseCategoricalCrossentropy(
  ),
23.
                  metrics=['accuracy'])
24. earlystopping = callbacks. EarlyStopping (monitor = "val loss",
25.
                                              mode ="min", patience = 5
26.
                                              restore best weights = Tr
27.
28. time callback = TimeHistory()
29. history = model1.fit(train X, train labels, epochs=40,
30.
                        validation data=(test X, test labels), shuffle
    = True, batch size = 64 , callbacks = [time callback, earlystopping]
31. times = time callback.times
32. test loss, test acc = model1.evaluate(test X, test labels, verbo
   se=1)
      1. accuracy of the first 5 epochs:
         1. Epoch 1/40:loss: 0.2007 - accuracy: 0.9396 - val loss: 0.0568 - val accuracy:
```

- Epoch 1/40:loss: 0.2007 accuracy: 0.9396 val\_loss: 0.0568 val\_accuracy: 0.9834
- 2. Epoch 2/40:loss: 0.0585 accuracy: 0.9821 val\_loss: 0.0425 val\_accuracy: 0.9859
- 3.Epoch 3/40:loss: 0.0410 accuracy: 0.9870 val\_loss: 0.0349 val\_accuracy: 0.9887
- 4. Epoch 4/40:loss: 0.0307 accuracy: 0.9901 val loss: 0.0353 val accuracy: 0.9869
- 5. Epoch 5/40:loss: 0.0250 accuracy: 0.9920 val\_loss: 0.0359 val\_accuracy: 0.9891

# 2. Accuracy of the final epoch: 99.03

## 3. Layers:

- 2 CNN layers with a relu activation function and followed by maxpooling 2D with kernel size 2\*2 and strides 2\*2
- Flattening layer
- 2 Fully connected layers with relu activation functions
- Another fully connected layer with softmax activation function

4- average train time: [5.872049570083618, 3.538182497024536, 3.7079007625579834, 3.6989691257476807, 3.4229421615600586, 3.741835832595825, 3.7338030338287354, 3.412816286087036, 3.443467378616333, 3.4417223930358887, 3.822333574295044, 3.717013120651245]

5. Number of parameters: 123,690

6. Used optimizer: adam

## Adedelta optimizer:

```
model1 =Sequential()
modell.add(Conv2D(32, (3, 3), activation='relu', input shape=(28, 28, 1
)))
model1.add(MaxPooling2D((2, 2),strides=(2,2)))
model1.add(Conv2D(64, (3, 3), activation='relu'))
model1.add(MaxPooling2D((2, 2),strides=(2,2)))
model1.add(Flatten())
model1.add(Dense(64, activation='relu'))
model1.add(Dense(32, activation='relu'))
model1.add(Dense(10, activation='softmax'))
sgd = SGD(learning rate=0.1, momentum=0.0)
model1.compile(optimizer= 'adadelta',
              loss=tf.keras.losses.SparseCategoricalCrossentropy(),
              metrics=['accuracy'])
earlystopping = callbacks. EarlyStopping (monitor = "val loss",
                                         mode ="min", patience = 5,
                                         restore best weights = True)
time callback = TimeHistory()
history = model1.fit(train X, train labels, epochs=40,
                    validation data=(test X, test labels), shuffle = Tru
e, batch size = 64 , callbacks = [time callback, earlystopping])
times = time callback.times
```

- 1. Accuracy: 90.65
- 2. Epoch 1/40:loss: 2.2913 accuracy: 0.1049 val\_loss: 2.2710 val\_accuracy: 0.1155
- 3. Epoch 2/40:loss: 2.2543 accuracy: 0.1389 val loss: 2.2356 val accuracy: 0.1570
- 4. Epoch 3/40:loss: 2.2156 accuracy: 0.1815 val\_loss: 2.1923 val\_accuracy: 0.2108
- 5. Epoch 4/40:loss: 2.1684 accuracy: 0.2344 val loss: 2.1422 val accuracy: 0.2428
- 6. Epoch 5/40:loss: 2.1142 accuracy: 0.2675 val\_loss: 2.0842 val\_accuracy: 0.2810
- 7. Optimizer: adadelta

#### FINAL OBSERVATION:

Adam optimizer gives higher accuracy than adadelta and sgd, accuracy is =99.03, so will keep using the model with adam optimizer.

## ADD A DROPOUT LAYER:

We try using a dropout layer at 2 different places and 2 different rates, we will have 4 probabilities , 2 of them will contain the rates 50% and 75% after the convolution layers once and the another after the 2 fully connected layers and before the final one , it went that way :

Position of dropout layer	Drop rate	Accuracy
After the 2 conv2D layers	50%	99.30
Before the last FC layer	50%	99.02
After the 2 conv2D layers	75%	99.42
Before the last FC layer	75%	98.73

# Final observation:

After testing all possibilities we can see that the highest accuracy we can get is 99.42 with drop rate = 75% after the fully connected layer.