

모듈 임포트

```
1 import numpy as np
2 import pandas as pd
3 import matplotlib.pyplot as plt # 그래프를 그리기 위한 import
4
5 import tensorflow as tf
6
7 from tensorflow import keras
8 from tensorflow.keras import optimizers
9 from tensorflow.keras.layers import Dense
10 from tensorflow.keras.layers import Flatten
11 from tensorflow.keras.layers import Conv2D, MaxPooling2D, Input, Reshape # CNN을 위한 import
12
13 import time
```

데이터

데이터 다운로드

```
1 from google.colab import drive
2 drive.mount("/content/drive") # 구글 드라이브를 마운트한다.
```

Mounted at /content/drive

```
1 !cp -r /content/drive/MyDrive/cau_temp/number_data ./ # 데이터 파일(number_data)을 가져온다.
```

```
1 %cd number_data/
```

/content/number_data

```
1 raw_train = pd.read_csv("train.csv") # raw_train에 train.csv내용 넣기
2 raw_train.head()
```

	id	digit	letter	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26	27	28	29	30	31	32	33	34	35	36	...	744	745	746	747	748	749	750	751																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																	
0	1	5	L	1	1	1	4	3	0	0	4	4	3	0	4	3	3	3	4	4	0	0	1	1	3	4	0	4	2	0	4	0	1	3	1	0	4	1	1	3	...	4	3	4	1	3	0	0																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																		
1	2	0	B	0	4	0	0	4	1	1	1	4	2	0	3	4	0	0	2	3	4	0	3	4	3	0	2	2	1	4	2	3	3	4	1	2	4	2	0	3	...	4	2	3	0	0	0	0																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																		
2	3	4	L	1	1	2	2	1	1	1	0	2	1	3	2	2	2	4	1	1	4	1	0	1	3	4	2	2	2	4	1	1	2	0	3	0	2	3	4	0	...	3	0	4	0	3	0	2																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																		
3	4	9	D	1	2	0	2	0	4	0	3	4	3	1	0	3	2	2	0	3	4	1	0	4	1	2	2	3	2	2	0	2	0	3	0	3	2	4	0	0	...	0	3	0	1	4	1	3																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																		
4	5	6	A	3	0	2	4	0	3	0	4	2	4	2	1	4	1	1	4	4	0	2	3	4	4	3	3	3	3	4	1	0	3	0	3	0	0	0	1	1	...	2	1	3	2	1	4	2																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																		
5 rows × 787 columns																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																																		

```
1 raw_test = pd.read_csv("test.csv") # raw_test에 test.csv내용 넣기
2 raw_test.head()
```

	id	letter	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26	27	28	29	30	31	32	33	34	35	36	37	...	744	745	746	747	748	749	750	751																												
0	2049	L	0	4	0	2	4	2	3	1	0	0	1	0	1	3	4	4	0	0	2	4	4	1	3	3	2	2	4	1	0	1	2	2	1	2	2	1	4	0	...	1	3	1	1	3	3	4																													
1	2050	C	4	1	4	0	1	1	0	2	2	1	0	3	0	1	1	4	1	2	0	2	2	0	4	3	4	0	2	4	4	2	1	2	4	0	4	2	0	2	...	3	4	2	6	2	2	0																													
2	2051	S	0	4	0	1	3	2	3	0	2	1	2	0	1	0	3	0	1	4	3	0	0	3	0	4	1	0	3	2	0	4	1	2	0	0	1	3	0	2	...	0	4	4	3	4	1	4																													
3	2052	K	2	1	3	3	3	4	3	0	0	2	3	2	3	4	4	4	0	1	4	2	2	0	1	4	3	1	3	0	2	3	2	4	3	1	1	4	0	0	...	0	4	1	1	2	3	2																													
4	2053	W	1	0	1	1	2	2	1	4	1	1	4	3	4	1	2	1	4	3	3	4	0	4	4	2	0	0	0	0	3	4	0	1	4	2	2	2	1	4	...	4	1	3	2	1	2	1																													
5 rows × 786 columns																																																																													

```
1 data1 = raw_train.to_numpy() # data1은 train 파일 내용
2 print(data1.shape)
3 print(data1[:5])
```

```
(2048, 787)
[[1 5 'L' ... 4 3 4]
 [2 0 'B' ... 2 1 2]
 [3 4 'L' ... 0 2 2]
 [4 9 'D' ... 0 1 1]
 [5 6 'A' ... 3 1 2]]
```

```
1 data2 = raw_test.to_numpy() #data2는 test 파일
2 print(data2.shape)
3 print(data2[:5])
```

```
(20480, 786)
[[2049 'L' 0 ... 4 1 4]
 [2050 'C' 4 ... 2 1 2]
 [2051 'S' 0 ... 0 1 4]
 [2052 'K' 2 ... 4 4 4]
 [2053 'W' 1 ... 2 3 4]]
```

데이터 분리하기

```
1 # 각 이름에 맞게 내용 넣기
2 raw_train_x=(data1[:,3:]) # 모든 행 , 0 1 2 3 -> 4번째 열부터 끝까지 -> 2828 좌표의 모든 숫자 저장
3 raw_train_y=(data1[:,1]) # 모든 행, 0 1 -> 2번째 열만 -> digit 저장
4 raw_train_z=(data1[:,2]) # 모든 행, 0 1 2 -> 3번째 열만 -> letter 저장
5 raw_test_x=(data2[:,2:]) # 모든행, 0 1 2 -> 3번째 열부터 끝까지 -> 28,28 좌표의 모든 숫자 저장
6 raw_test_y=(data2[:,1]) # 모든 행, 0 1 -> 2번째 열만 -> letter 저장
7
8
9 print(raw_train_x)
10 print(raw_train_y)
11 print(raw_train_z)
12 print(raw_test_x)
13 print(raw_test_y)
14
```

```
[[1 1 1 ... 4 3 4]
```

```
[0 4 0 ... 2 1 2]
[1 1 2 ... 0 2 2]
...
[4 0 4 ... 2 0 0]
[2 3 3 ... 4 3 1]
[4 2 2 ... 4 3 4]]
[5 0 4 ... 9 0 5]
['L' 'B' 'L' ... 'A' 'Z' 'Z']
[[0 4 0 ... 4 1 4]
 [4 1 4 ... 2 1 2]
 [0 4 0 ... 0 1 4]
 ...
 [4 2 1 ... 3 1 1]
 [1 1 2 ... 4 4 2]
 [2 1 0 ... 2 2 0]]
['L' 'C' 'S' ... 'B' 'K' 'S']
```

```
1 print(raw_train_x.shape)
2 print(raw_train_y.shape)
3 print(raw_test_x.shape)
4 print(raw_test_y.shape)
```

```
(2048, 784)
(2048,)
(20480, 784)
(20480,)
```

▶ 데이터 사용하기

데이터를 사용하기 위해 모델의 모양과 type을 바꾸어준다.

(784개의 나열 -> 28, 28 좌표의 형태 / object -> uint8)

```
1 raw_train_x = raw_train_x.reshape((2048, 28, 28)) # 28, 28 좌표로 형태를 바꿈
2 raw_test_x = raw_test_x.reshape((20480, 28, 28))
3 print(raw_train_x.shape)
4 print(raw_test_x.shape)
```

```
(2048, 28, 28)
(20480, 28, 28)
```

```
1 from numpy import uint8
2 raw_train_x=raw_train_x.astype('uint8') #타입을 숫자로 바꾸기
3 raw_test_x=raw_test_x.astype('uint8')
4 raw_train_y=raw_train_y.astype('uint8') #타입을 숫자로 바꾸기
```

```
1 raw_train_x

array([[ [1, 1, 1, ..., 2, 0, 4],
        [0, 1, 3, ..., 4, 1, 3],
        [2, 0, 4, ..., 3, 3, 3],
        ...,
        [2, 1, 2, ..., 4, 1, 0],
        [3, 3, 3, ..., 3, 3, 0],
        [3, 2, 2, ..., 4, 3, 4]],

       [ [0, 4, 0, ..., 1, 4, 2],
        [3, 3, 4, ..., 3, 4, 2],
        [1, 4, 2, ..., 4, 0, 4],
        ...,
        [2, 4, 4, ..., 0, 0, 1],
        [3, 1, 4, ..., 2, 2, 4],
        [2, 1, 1, ..., 2, 1, 2]],

       [ [1, 1, 2, ..., 2, 4, 1],
        [1, 2, 0, ..., 2, 4, 0],
        [4, 2, 0, ..., 1, 3, 3],
        ...,
        [3, 0, 1, ..., 2, 2, 1],
        [0, 3, 2, ..., 4, 2, 3],
        [4, 4, 4, ..., 0, 2, 2]],

       ...,

       [ [4, 0, 4, ..., 3, 2, 3],
        [4, 3, 1, ..., 4, 4, 4],
        [3, 2, 0, ..., 1, 3, 0],
        ...,
        [4, 0, 4, ..., 0, 2, 0],
        [4, 3, 1, ..., 0, 4, 4],
        [1, 3, 3, ..., 2, 0, 0]],

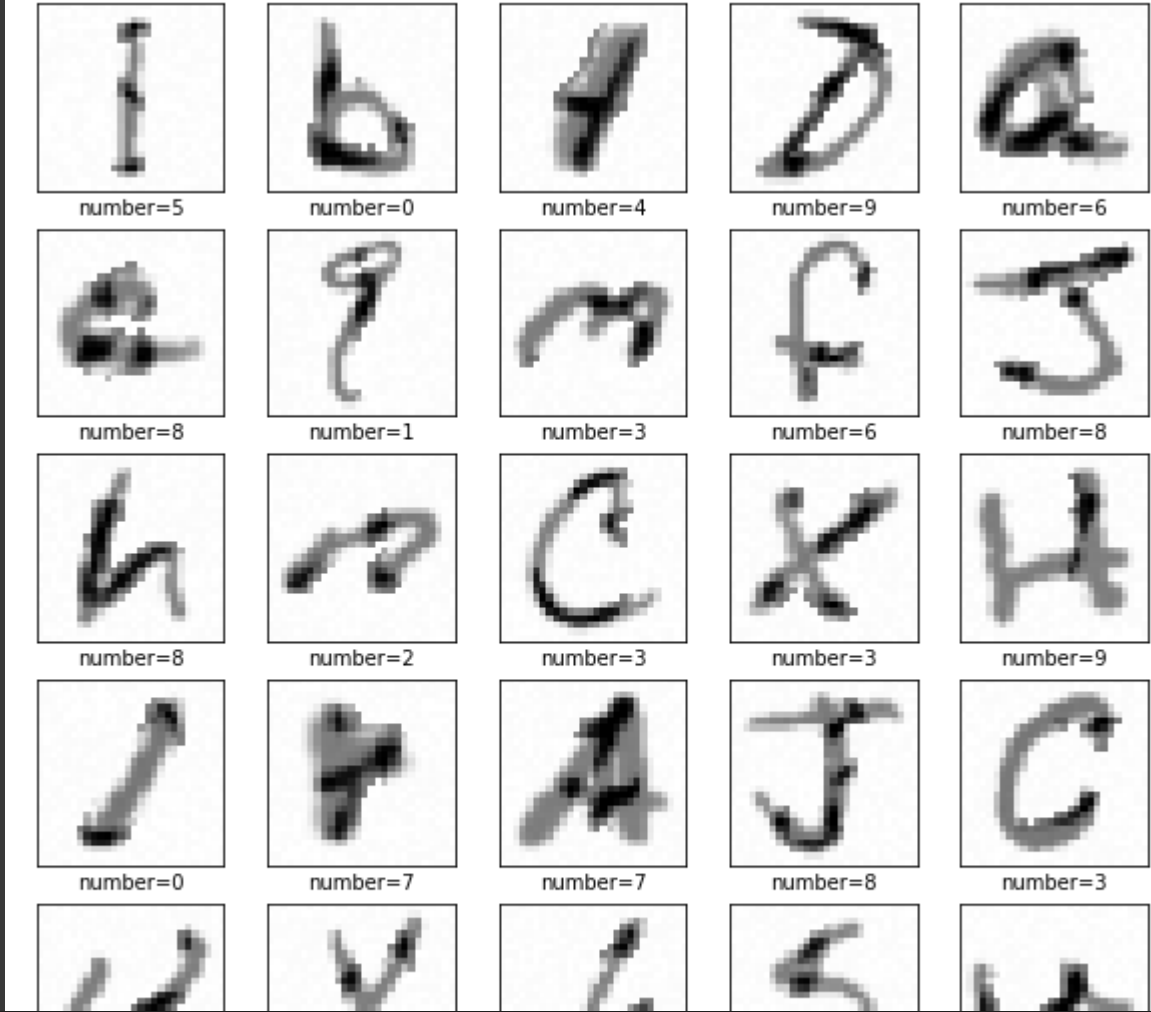
       [ [2, 3, 3, ..., 0, 1, 2],
        [4, 1, 4, ..., 4, 0, 0],
        [3, 1, 3, ..., 3, 3, 3],
        ...,
        [1, 1, 4, ..., 1, 0, 0],
        [4, 4, 0, ..., 0, 1, 0],
        [1, 0, 1, ..., 4, 3, 1]],

       [ [4, 2, 2, ..., 3, 1, 2],
        [0, 2, 4, ..., 0, 3, 3],
        [4, 1, 4, ..., 4, 4, 4],
        ...,
        [0, 0, 2, ..., 3, 2, 2],
        [2, 3, 2, ..., 0, 1, 0],
        [0, 3, 3, ..., 4, 3, 4]]], dtype=uint8)
```

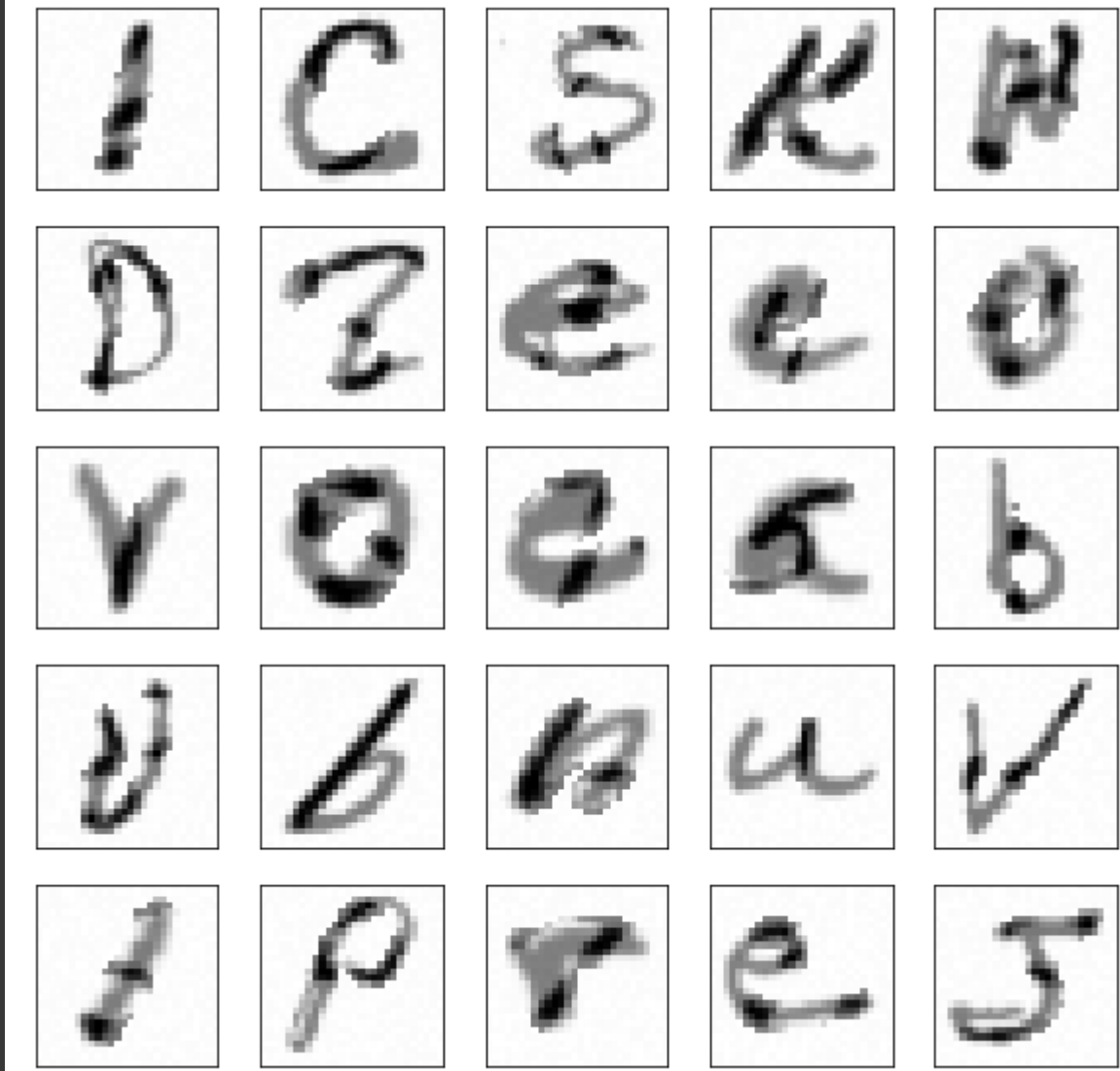
▶ 그림으로 나타내보기

```
1 # ADD
2 number_class_names =  ['number=0', 'number=1', 'number=2', 'number=3', 'number=4', 'number=5', 'number=6', 'number=7', 'number=8', 'number=9']
```

```
1 # train의 데이터
2 plt.figure(figsize=(10,10))
3 for i in range(25):
4     plt.subplot(5,5,i+1) # 한줄에 5개, 5줄로 그린다. 그중에 i+1 번째칸에 그린다.
5     plt.xticks([])      # 이미지 그릴때 가로축의 눈금 그리지 않는다.
6     plt.yticks([])      # 이미지 그릴때 세로축의 눈금 그리지 않는다.
7     plt.grid(False)     # 이미지 내의 눈금을 그리지 않는다.
8     plt.imshow(raw_train_x[i], cmap=plt.cm.binary)
9     plt.xlabel(number_class_names[raw_train_y[i]])
10 plt.show()
```



```
1 #test의 데이터
2 plt.figure(figsize=(10,10))
3 for i in range(25):
4     plt.subplot(5,5,i+1) # 한줄에 5개, 5줄로 그린다. 그중에 i+1 번째칸에 그린다.
5     plt.xticks([])      # 이미지 그릴때 가로축의 눈금 그리지 않는다.
6     plt.yticks([])      # 이미지 그릴때 세로축의 눈금 그리지 않는다.
7     plt.grid(False)     # 이미지 내의 눈금을 그리지 않는다.
8     plt.imshow(raw_test_x[i], cmap=plt.cm.binary)
9 plt.show()
```



Normalization

```
1 print(np.max(raw_train_x[:,]))
2 print(np.max(raw_test_x[:,]))
3
4 train_x = raw_train_x/255 # 0~255까지의 수치를 255로 나누어서 0~1의 수치로 변환한다.
5 test_x = raw_test_x/255
6
7 train_y = raw_train_y
8 test_y = raw_test_y
9
10 print(np.max(train_x[:,]))
11 print(np.max(test_x[:,]))
12 print(train_y)
13
255
255
1.0
1.0
[5 0 4 ... 9 0 5]
```

로스 실시간으로 나타내기

```
1 # copy from https://gist.github.com/stared/dfb4dfaf6d9a8501cd1cc8b8cb806d2e
2
3 from IPython.display import clear_output
4 from tensorflow.keras.callbacks import Callback
5
6 class PlotLosses(Callback):
7
8     def on_train_begin(self, logs={}):
9
10         self.i = 0
11         self.x = []
12         self.losses = []
13         self.val_losses = []
14
15         self.fig = plt.figure()
16
17         self.logs = []
18
19
20     def on_epoch_end(self, epoch, logs={}):
21
```

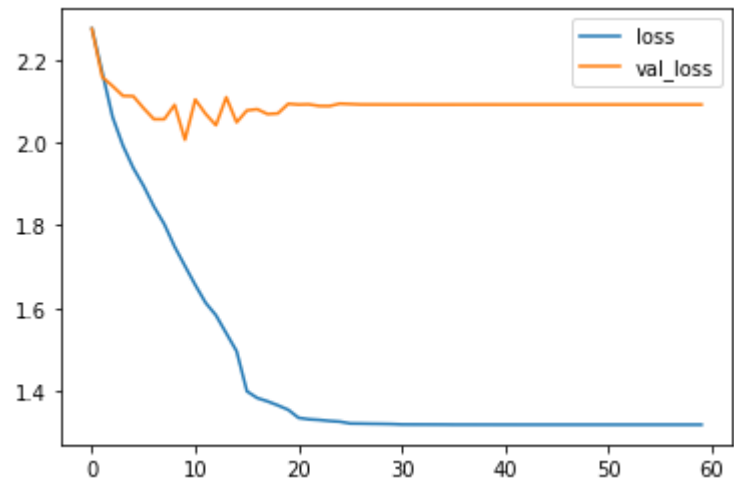
```
22     self.logs.append(logs)
23     self.x.append(self.i)
24     self.losses.append(logs.get('loss'))
25     self.val_losses.append(logs.get('val_loss'))
26     self.i += 1
27
28     clear_output(wait=True)
29     plt.plot(self.x, self.losses, label="loss")
30     plt.plot(self.x, self.val_losses, label="val_loss")
31     plt.legend()
32     plt.show():
33     print("loss = ", self.losses[-1], ", val_loss = ", self.val_losses[-1])
34
35
```

모델

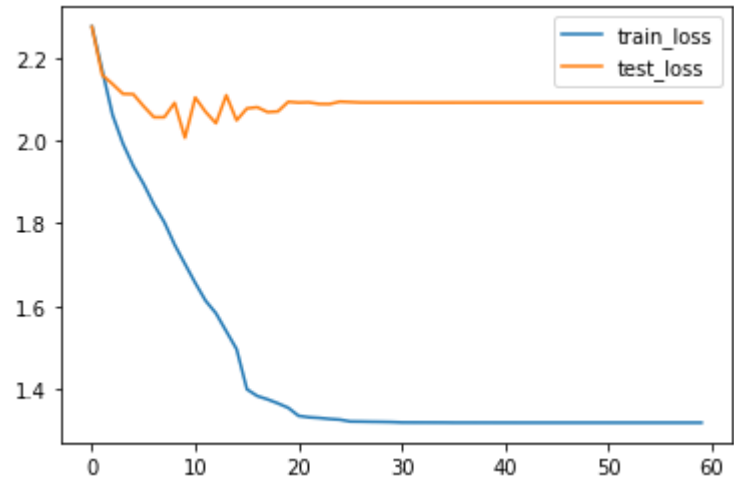
DNN

Dense 10을 20으로 변경

```
1 model = keras.Sequential()
2
3 model.add(Flatten())
4 model.add(Dense(20, activation='relu'))
5 model.add(Dense(20, activation='relu'))
6 model.add(Dense(20, activation='relu'))
7 model.add(Dense(10, activation='softmax'))
8
9 model.compile(optimizer="adam", loss="sparse_categorical_crossentropy", metrics=["accuracy"]) # 정수형 자료의 입출력이므로 loss 에서 SCC (Sparse categorical entropy ) 사용
10
11 #-----모델 저장, early stopping, 학습을 조정
12 from tensorflow.keras.callbacks import ModelCheckpoint, EarlyStopping, ReduceLR0nPlateau
13
14
15 model_check_point = ModelCheckpoint('best_model.h5', monitor='val_loss', mode='min', save_best_only=True)
16 plot_losses = PlotLosses()
17 early_stopping = EarlyStopping(monitor='val_loss', mode='min', verbose=1, patience=50)
18 reduce_lr = ReduceLR0nPlateau(monitor='val_loss', factor=0.2, patience=5, min_learning_rate=0.001)
19
20 callbacks = [model_check_point, plot_losses, early_stopping, reduce_lr]
21 #-----
22
23 start_time = time.time()
24 history = model.fit(train_x, train_y, validation_split=0.1, epochs=100, verbose=0, callbacks=callbacks, batch_size=16) # validation data set을 위해 10% 만큼 분리하여 사용
25
26
27 plt.plot(history.history['loss'], label='train_loss')
28 plt.plot(history.history['val_loss'], label='test_loss')
29 plt.legend()
30 plt.show()
31 loss, acc = model.evaluate(train_x, train_y)
32 print("loss=",loss)
33 print("acc=",acc)
34
35 print("elapsed : {}".format(time.time() - start_time))
```



loss = 1.3192002773284912 , val_loss = 2.091294050216675
Epoch 00060: early stopping

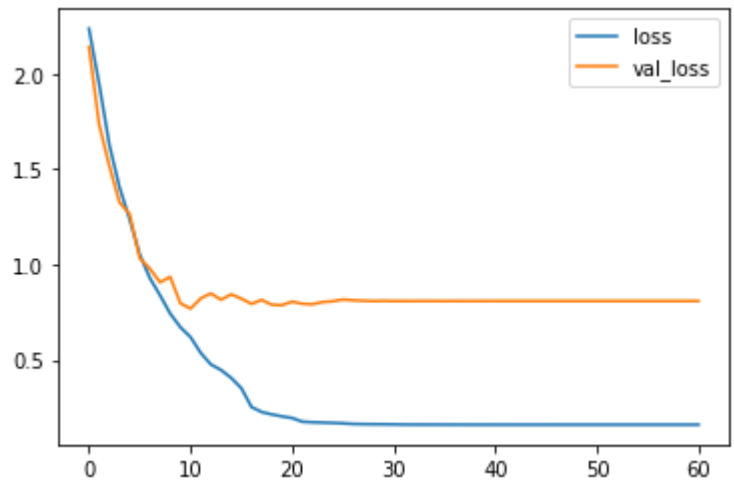


64/64 [=====] - 0s 2ms/step - loss: 1.3965 - accuracy: 0.5347
loss= 1.396485686302185
acc= 0.53466796875
elapsed : 23.639520406723022

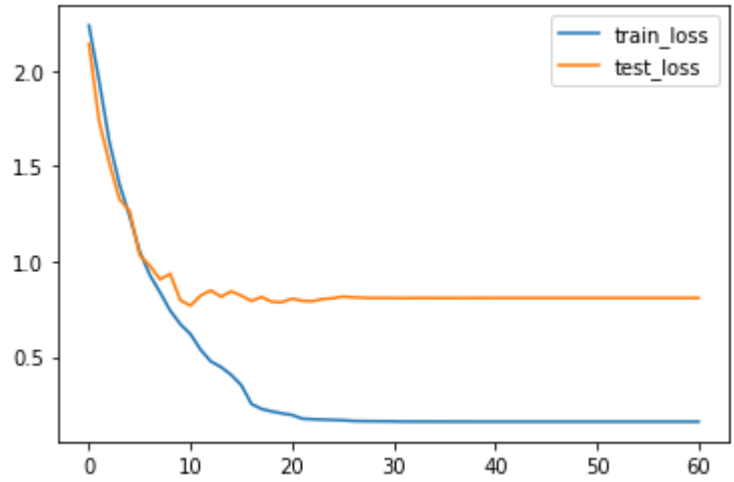
CNN

```
1 model = keras.Sequential()
2
3 model.add(Input((28,28)))
4 model.add(Reshape((28,28,1))) # ADDED
5 model.add(Conv2D(32, (3, 3), padding="same", activation="relu")) # ADDED # 입출력 이미지 사이즈가 같으므로 padding 에는 same
6 model.add(MaxPooling2D((2, 2))) # ADDED
7 model.add(Conv2D(64, (3, 3), padding="same", activation="relu")) # ADDED
8 model.add(MaxPooling2D((2, 2))) # ADDED
9
10 model.add(Flatten())
11 model.add(Dense(20, activation='relu'))
12 model.add(Dense(20, activation='relu'))
13 model.add(Dense(20, activation='relu'))
14 model.add(Dense(10, activation='softmax'))
15
16 model.compile(optimizer="adam", loss="sparse_categorical_crossentropy", metrics=["accuracy"])
17
18 from tensorflow.keras.callbacks import ModelCheckpoint, EarlyStopping, ReduceLR0nPlateau
19
```

```
20
21 model_check_point = ModelCheckpoint('best_model.h5', monitor='val_loss', mode='min', save_best_only=True)
22 plot_losses = PlotLosses()
23 early_stopping = EarlyStopping(monitor='val_loss', mode='min', verbose=1, patience=50)
24 reduce_lr = ReduceLROnPlateau(monitor='val_loss', factor=0.2, patience=5, min_learning_rate=0.001)
25
26 callbacks = [model_check_point, plot_losses, early_stopping, reduce_lr]
27
28 start_time = time.time()
29 history = model.fit(train_x, train_y, validation_split=0.1, epochs=100, verbose=0, callbacks=callbacks, batch_size=16)
30
31
32 plt.plot(history.history['loss'], label='train_loss')
33 plt.plot(history.history['val_loss'], label='test_loss')
34 plt.legend()
35 plt.show()
36 loss, acc = model.evaluate(train_x, train_y)
37 print("loss=",loss)
38 print("acc=",acc)
39
40 print("elapsed : {}".format(time.time() - start_time))
```



loss = 0.16055725514888763 , val_loss = 0.8088459968566895
Epoch 00061: early stopping

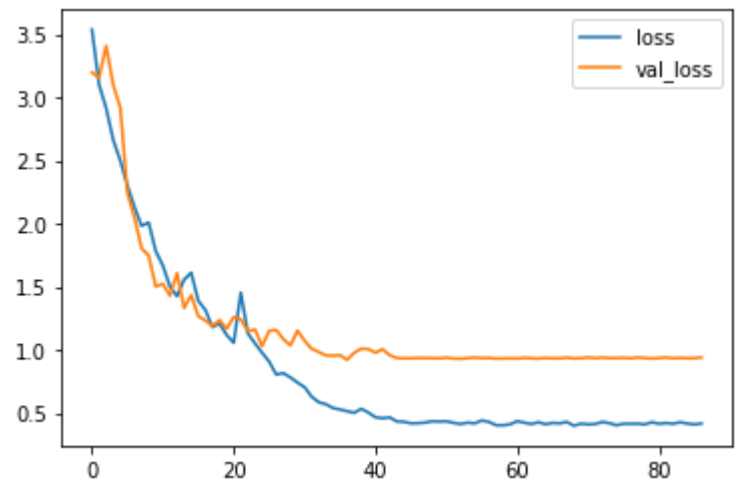


64/64 [=====] - 1s 13ms/step - loss: 0.2254 - accuracy: 0.9463
loss= 0.22544939815998077
acc= 0.9462890625
elapsed : 162.6628782749176

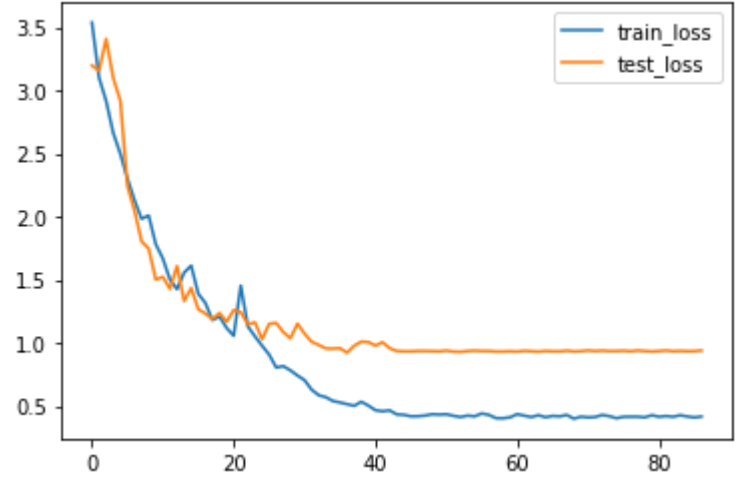
overfitting

- Dropout
- BatchNormalization
- regularizer

```
1 from tensorflow.keras.layers import Dropout
2 from tensorflow.keras.layers import BatchNormalization
3 from tensorflow.keras.regularizers import l1, l2, L1L2
4
5 model = keras.Sequential()
6
7 model.add(Input((28,28)))
8 model.add(Reshape((28,28,1)))
9 model.add(Conv2D(32, (3, 3), padding="same", activation="relu"))
10 model.add(BatchNormalization())
11 model.add(MaxPooling2D((2, 2)))
12 model.add(Conv2D(64, (3, 3), padding="same", activation="relu"))
13 model.add(BatchNormalization())
14 model.add(MaxPooling2D((2, 2)))
15
16 model.add(Flatten())
17 model.add(Dense(20, activation='relu', kernel_regularizer=l2()))
18 model.add(BatchNormalization())
19 model.add(Dropout(0.2))
20 model.add(Dense(20, activation='relu', kernel_regularizer=l2()))
21 model.add(BatchNormalization())
22 model.add(Dropout(0.2))
23 model.add(Dense(20, activation='relu', kernel_regularizer=l2()))
24 model.add(BatchNormalization())
25 model.add(Dropout(0.2))
26 model.add(Dense(10, activation='softmax'))
27
28 model.compile(optimizer="adam", loss="sparse_categorical_crossentropy", metrics=["accuracy"])
29 model.summary()
30
31
32 from tensorflow.keras.callbacks import ModelCheckpoint, EarlyStopping, ReduceLROnPlateau
33
34
35 model_check_point = ModelCheckpoint('best_model.h5', monitor='val_loss', mode='min', save_best_only=True)
36 plot_losses = PlotLosses()
37 early_stopping = EarlyStopping(monitor='val_loss', mode='min', verbose=1, patience=50)
38 reduce_lr = ReduceLROnPlateau(monitor='val_loss', factor=0.2, patience=5, min_learning_rate=0.001)
39
40 callbacks = [model_check_point, plot_losses, early_stopping, reduce_lr]
41
42 start_time = time.time()
43 history = model.fit(train_x, train_y, validation_split=0.1, epochs=100, verbose=0, callbacks=callbacks, batch_size=16)
44
45
46 plt.plot(history.history['loss'], label='train_loss')
47 plt.plot(history.history['val_loss'], label='test_loss')
48 plt.legend()
49 plt.show()
50 loss, acc = model.evaluate(train_x, train_y)
51 print("loss=",loss)
52 print("acc=",acc)
53
54 print("elapsed : {}".format(time.time() - start_time))
```



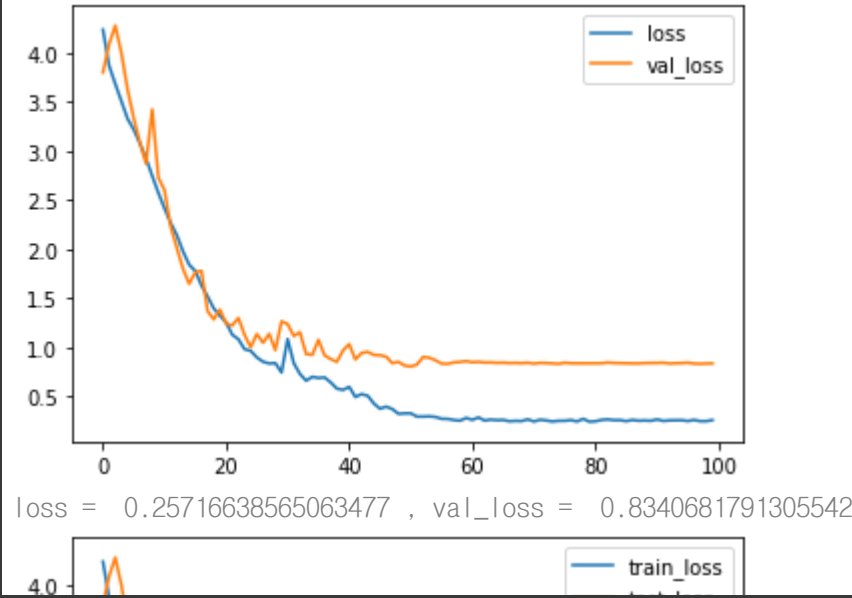
loss = 0.41743671894073486 , val_loss = 0.9403938055038452
Epoch 00087: early stopping



64/64 [=====] - 1s 16ms/step - loss: 0.3129 - accuracy: 0.9785
loss= 0.31285884976387024
acc= 0.978515625
elapsed : 311.6710584163666

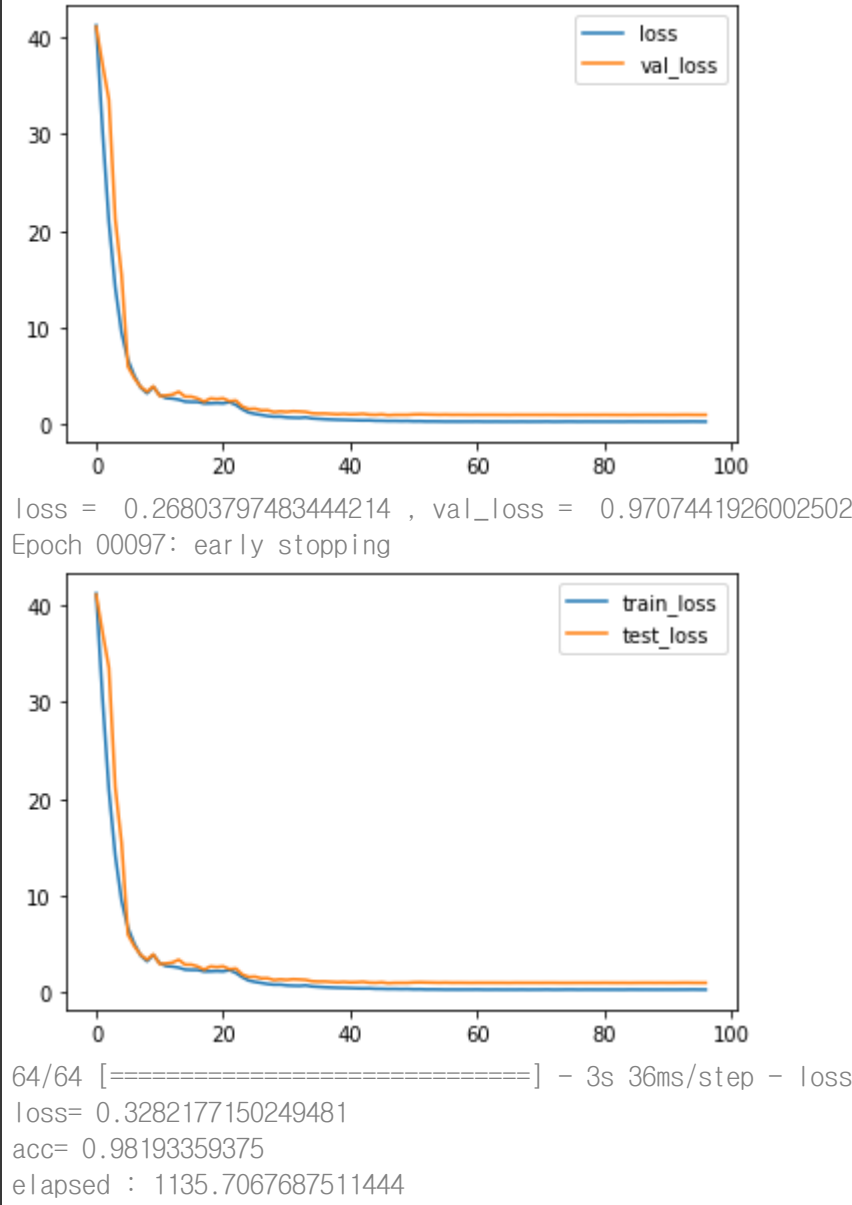
- 1번
- 노드 : 20 -> 30
- 히든 레이어 : 3개 -> 4개
- Conv2D : 3개 -> 4개

```
1 from tensorflow.keras.layers import Dropout
2 from tensorflow.keras.layers import BatchNormalization
3 from tensorflow.keras.regularizers import l1, l2, L1L2
4
5 model = keras.Sequential()
6
7 model.add(Input((28,28)))
8 model.add(Reshape((28,28,1)))
9 model.add(Conv2D(32, (3, 3), padding="same", activation="relu"))
10 model.add(BatchNormalization())
11 model.add(MaxPooling2D((2, 2)))
12 model.add(Conv2D(64, (3, 3), padding="same", activation="relu"))
13 model.add(BatchNormalization())
14 model.add(MaxPooling2D((2, 2)))
15 model.add(Conv2D(128, (3, 3), padding="same", activation="relu"))
16 model.add(BatchNormalization())
17 model.add(MaxPooling2D((2, 2)))
18
19 model.add(Flatten())
20 model.add(Dense(30, activation='relu', kernel_regularizer=l2()))
21 model.add(BatchNormalization())
22 model.add(Dropout(0.2))
23 model.add(Dense(30, activation='relu', kernel_regularizer=l2()))
24 model.add(BatchNormalization())
25 model.add(Dropout(0.2))
26 model.add(Dense(30, activation='relu', kernel_regularizer=l2()))
27 model.add(BatchNormalization())
28 model.add(Dropout(0.2))
29 model.add(Dense(30, activation='relu', kernel_regularizer=l2()))
30 model.add(BatchNormalization())
31 model.add(Dropout(0.2))
32 model.add(Dense(10, activation='softmax'))
33
34 model.compile(optimizer="adam", loss="sparse_categorical_crossentropy", metrics=["accuracy"])
35 model.summary()
36
37
38 from tensorflow.keras.callbacks import ModelCheckpoint, EarlyStopping, ReduceLR0nPlateau
39
40
41 model_check_point = ModelCheckpoint('best_model.h5', monitor='val_loss', mode='min', save_best_only=True)
42 plot_losses = PlotLosses()
43 early_stopping = EarlyStopping(monitor='val_loss', mode='min', verbose=1, patience=50)
44 reduce_lr = ReduceLR0nPlateau(monitor='val_loss', factor=0.2, patience=5, min_learning_rate=0.001)
45
46 callbacks = [model_check_point, plot_losses, early_stopping, reduce_lr]
47
48 start_time = time.time()
49 history = model.fit(train_x, train_y, validation_split=0.1, epochs=100, verbose=0, callbacks=callbacks, batch_size=16)
50
51
52 plt.plot(history.history['loss'], label='train_loss')
53 plt.plot(history.history['val_loss'], label='test_loss')
54 plt.legend()
55 plt.show()
56 loss, acc = model.evaluate(train_x, train_y)
57 print("loss=",loss)
58 print("acc=",acc)
59
60 print("elapsed : {}".format(time.time() - start_time))
```



- 2번
- 노드 : 30 -> 1024

```
1 from tensorflow.keras.layers import Dropout
2 from tensorflow.keras.layers import BatchNormalization
3 from tensorflow.keras.regularizers import l1, l2, L1L2
4
5 model = keras.Sequential()
6
7 model.add(Input((28,28)))
8 model.add(Reshape((28,28,1)))
9 model.add(Conv2D(32, (3, 3), padding="same", activation="relu"))
10 model.add(BatchNormalization())
11 model.add(MaxPooling2D((2, 2)))
12 model.add(Conv2D(64, (3, 3), padding="same", activation="relu"))
13 model.add(BatchNormalization())
14 model.add(MaxPooling2D((2, 2)))
15 model.add(Conv2D(128, (3, 3), padding="same", activation="relu"))
16 model.add(BatchNormalization())
17 model.add(MaxPooling2D((2, 2)))
18
19 model.add(Flatten())
20 model.add(Dense(1024, activation='relu', kernel_regularizer=l2()))
21 model.add(BatchNormalization())
22 model.add(Dropout(0.2))
23 model.add(Dense(1024, activation='relu', kernel_regularizer=l2()))
24 model.add(BatchNormalization())
25 model.add(Dropout(0.2))
26 model.add(Dense(1024, activation='relu', kernel_regularizer=l2()))
27 model.add(BatchNormalization())
28 model.add(Dropout(0.2))
29 model.add(Dense(1024, activation='relu', kernel_regularizer=l2()))
30 model.add(BatchNormalization())
31 model.add(Dropout(0.2))
32 model.add(Dense(10, activation='softmax'))
33
34 model.compile(optimizer="adam", loss="sparse_categorical_crossentropy", metrics=["accuracy"])
35 model.summary()
36
37
38 from tensorflow.keras.callbacks import ModelCheckpoint, EarlyStopping, ReduceLR0nPlateau
39
40
41 model_check_point = ModelCheckpoint('best_model.h5', monitor='val_loss', mode='min', save_best_only=True)
42 plot_losses = PlotLosses()
43 early_stopping = EarlyStopping(monitor='val_loss', mode='min', verbose=1, patience=50)
44 reduce_lr = ReduceLR0nPlateau(monitor='val_loss', factor=0.2, patience=5, min_learning_rate=0.001)
45
46 callbacks = [model_check_point, plot_losses, early_stopping, reduce_lr]
47
48 start_time = time.time()
49 history = model.fit(train_x, train_y, validation_split=0.1, epochs=100, verbose=0, callbacks=callbacks, batch_size=16)
50
51
52 plt.plot(history.history['loss'], label='train_loss')
53 plt.plot(history.history['val_loss'], label='test_loss')
54 plt.legend()
55 plt.show()
56 loss, acc = model.evaluate(train_x, train_y)
57 print("loss=",loss)
58 print("acc=",acc)
59
60 print("elapsed : {}".format(time.time() - start_time))
```

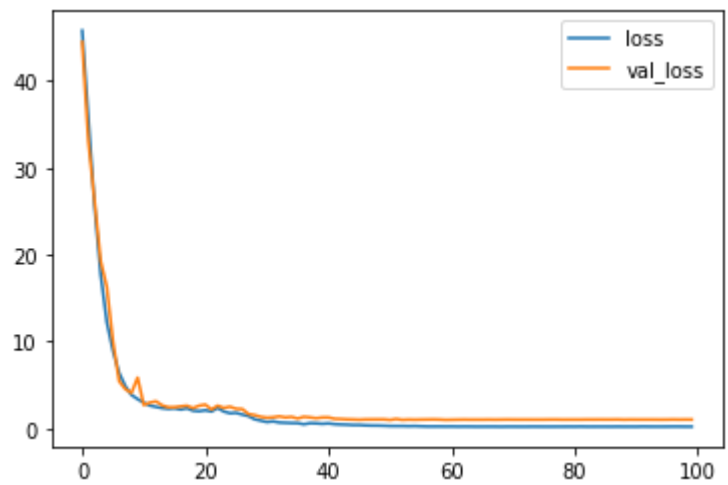


- 제출 결과

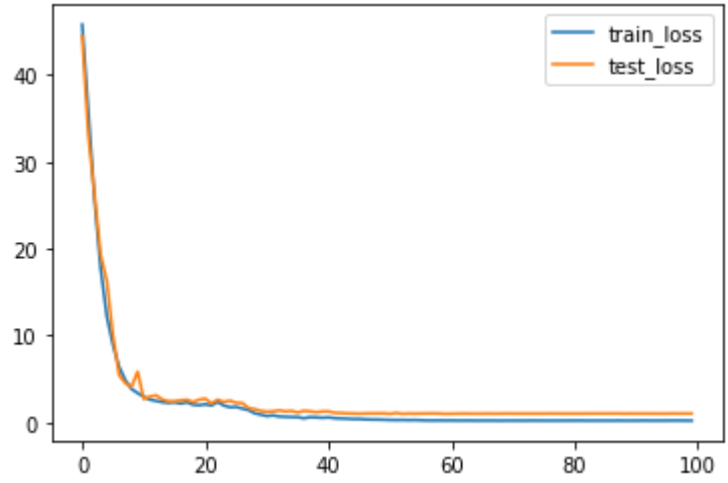
0.8235294118/ 0.8177155257

- 3번
- Conv2D : 3개 -> 4개
- 히든 레이어 : 4개 -> 5개

```
1 from tensorflow.keras.layers import Dropout
2 from tensorflow.keras.layers import BatchNormalization
3 from tensorflow.keras.regularizers import l1, l2, L1L2
4
5 model = keras.Sequential()
6
7 model.add(Input((28,28)))
8 model.add(Reshape((28,28,1)))
9 model.add(Conv2D(32, (3, 3), padding="same", activation="relu"))
10 model.add(BatchNormalization())
11 model.add(MaxPooling2D((2, 2)))
12 model.add(Conv2D(64, (3, 3), padding="same", activation="relu"))
13 model.add(BatchNormalization())
14 model.add(MaxPooling2D((2, 2)))
15 model.add(Conv2D(128, (3, 3), padding="same", activation="relu"))
16 model.add(BatchNormalization())
17 model.add(MaxPooling2D((2, 2)))
18 model.add(Conv2D(256, (3, 3), padding="same", activation="relu"))
19 model.add(BatchNormalization())
20 model.add(MaxPooling2D((2, 2)))
21
22 model.add(Flatten())
23 model.add(Dense(1024, activation='relu', kernel_regularizer=l2()))
24 model.add(BatchNormalization())
25 model.add(Dropout(0.2))
26 model.add(Dense(1024, activation='relu', kernel_regularizer=l2()))
27 model.add(BatchNormalization())
28 model.add(Dropout(0.2))
29 model.add(Dense(1024, activation='relu', kernel_regularizer=l2()))
30 model.add(BatchNormalization())
31 model.add(Dropout(0.2))
32 model.add(Dense(1024, activation='relu', kernel_regularizer=l2()))
33 model.add(BatchNormalization())
34 model.add(Dropout(0.2))
35 model.add(Dense(1024, activation='relu', kernel_regularizer=l2()))
36 model.add(BatchNormalization())
37 model.add(Dropout(0.2))
38 model.add(Dense(10, activation='softmax'))
39
40 model.compile(optimizer="adam", loss="sparse_categorical_crossentropy", metrics=["accuracy"])
41 model.summary()
42
43 start_time = time.time()
44
45 from tensorflow.keras.callbacks import ModelCheckpoint, EarlyStopping, ReduceLR0nPlateau
46
47
48 model_check_point = ModelCheckpoint('best_model.h5', monitor='val_loss', mode='min', save_best_only=True)
49 plot_losses = PlotLosses()
50 early_stopping = EarlyStopping(monitor='val_loss', mode='min', verbose=1, patience=50)
51 reduce_lr = ReduceLR0nPlateau(monitor='val_loss', factor=0.2, patience=5, min_learning_rate=0.001)
52
53 callbacks = [model_check_point, plot_losses, early_stopping, reduce_lr]
54
55 start_time = time.time()
56 history = model.fit(train_x, train_y, validation_split=0.1, epochs=100, verbose=0, callbacks=callbacks, batch_size=16)
57
58
59 plt.plot(history.history['loss'], label='train_loss')
60 plt.plot(history.history['val_loss'], label='test_loss')
61 plt.legend()
62 plt.show()
63 loss, acc = model.evaluate(train_x, train_y)
64 print("loss=", loss)
65 print("acc=", acc)
66
67
68 print("elapsed : {}".format(time.time() - start_time))
```



loss = 0.18398424983024597 , val_loss = 0.9994882941246033



64/64 [=====] - 3s 39ms/step - loss: 0.2555 - accuracy: 0.9839
loss= 0.25553351640701294
acc= 0.98388671875
elapsed : 1347.3061680793762

- 제출결과

0.8333333333/ 0.8136713356

예측해보기


```
1 submission_1 = pd.read_csv('/content/drive/MyDrive/cau_temp/number_data/submission.csv')
2 submission_1['digit'] = np.argmax(model.predict(test_x), axis=1)
3 submission_1.head(5)
```

	id	digit
0	2049	6
1	2050	9
2	2051	8
3	2052	0
4	2053	3

예측한 데이터 다운로드

```
1 from pandas import DataFrame
2 from google.colab import files # 파일을 다운받기 위해서 import
```

```
1 submission_1.to_csv('result8.csv')
```

```
1 files.download('result8.csv')
```