

1. 라이브러리 불러오기

```
In [52]: import tensorflow as tf
import numpy as np
from tensorflow.keras import datasets, layers, models
from keras.utils import plot_model
import matplotlib.pyplot as plt # !pip install matplotlib # 외부에서 설치
import pydot
from PIL import Image, ImageOps
```

2. 데이터 읽어오기

```
In [2]: # 훈련용(Train -> 질문 : 답), 평가용(test -> 질문 : 답) 데이터 분류해서 가져오기
# 데이터가 수정되면 안됨 -> 튜플로 가져오기
(train_images, train_labels), (test_images, test_labels) = datasets.mnist.load_data()
```

```
In [3]: # 데이터 확인
train_images.shape, train_labels.shape, test_images.shape, test_labels.shape
# ((60000, 28, 28) -> 흑백 이미지(채널이 없음 -> 60000, 28, 28, 3 -> 컬러), (6000, 28, 28, 3) -> 컬러)
# 생략된 채널을 넣어줘야 한다. -> 구조를 reshape 사용해서 조정
```

```
Out[3]: ((60000, 28, 28), (60000,), (10000, 28, 28), (10000,))
```

```
In [24]: # !pip install tensorflow_datasets
```

```

Collecting tensorflow_datasets
  Downloading tensorflow_datasets-4.9.4-py3-none-any.whl.metadata (9.2 kB)
Requirement already satisfied: absl-py in c:\users\hi\anaconda3\envs\p310_cnn\lib\site-packages (from tensorflow_datasets) (2.1.0)
Collecting click (from tensorflow_datasets)
  Downloading click-8.1.7-py3-none-any.whl.metadata (3.0 kB)
Collecting dm-tree (from tensorflow_datasets)
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Collecting etils>=0.9.0 (from etils[enp,epath,etree]>=0.9.0->tensorflow_datasets)
  Downloading etils-1.7.0-py3-none-any.whl.metadata (6.4 kB)
Requirement already satisfied: numpy in c:\users\hi\anaconda3\envs\p310_cnn\lib\site-packages (from tensorflow_datasets) (1.26.4)
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  Preparing metadata (setup.py): finished with status 'done'
Requirement already satisfied: protobuf>=3.20 in c:\users\hi\anaconda3\envs\p310_cnn\lib\site-packages (from tensorflow_datasets) (4.25.3)
Requirement already satisfied: psutil in c:\users\hi\anaconda3\envs\p310_cnn\lib\site-packages (from tensorflow_datasets) (5.9.8)
Requirement already satisfied: requests>=2.19.0 in c:\users\hi\anaconda3\envs\p310_cnn\lib\site-packages (from tensorflow_datasets) (2.31.0)
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Collecting tqdm (from tensorflow_datasets)
  Downloading tqdm-4.66.4-py3-none-any.whl.metadata (57 kB)
----- 0.0/57.6 kB ? eta -:--:--
----- 57.6/57.6 kB 3.0 MB/s eta 0:00:00
Requirement already satisfied: wrapt in c:\users\hi\anaconda3\envs\p310_cnn\lib\site-packages (from tensorflow_datasets) (1.16.0)
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Requirement already satisfied: certifi>=2017.4.17 in c:\users\hi\anaconda3\envs\p310_cnn\lib\site-packages (from requests>=2.19.0->tensorflow_datasets) (2024.2.2)
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```

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Downloading zipp-3.18.2-py3-none-any.whl (8.3 kB)
Building wheels for collected packages: promise
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c8a752867445bab994d2340724928aa3ab059c57c8db
Successfully built promise
Installing collected packages: dm-tree, zipp, tqdm, toml, protobuf, promise, impo
rtlib_resources, fsspec, etils, click, tensorflow-metadata, tensorflow_datasets
  Attempting uninstall: protobuf
    Found existing installation: protobuf 4.25.3
    Uninstalling protobuf-4.25.3:
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Successfully installed click-8.1.7 dm-tree-0.1.8 etils-1.7.0 fsspec-2024.5.0 impo
rtlib_resources-6.4.0 promise-2.3 protobuf-3.20.3 tensorflow-metadata-1.15.0 tens
orflow_datasets-4.9.4 toml-0.10.2 tqdm-4.66.4 zipp-3.18.2

WARNING: Failed to remove contents in a temporary directory 'C:\Users\hi\anacon
da3\envs\p310_cnn\Lib\site-packages\google\~upb'.
You can safely remove it manually.

```

```
In [4]: import tensorflow_datasets as tfds # tf에서 제공되는 ds
        tfds.list_builders()
```

```
Out[4]: ['abstract_reasoning',
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```

```

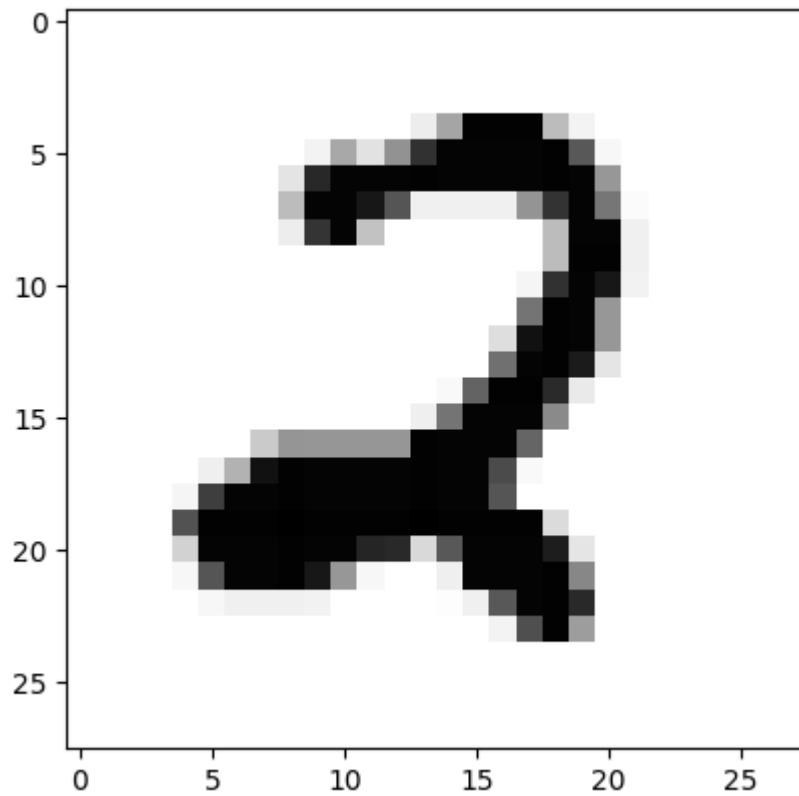
In [5]: # train_images, train_labels 데이터 확인
plt.imshow(train_images[365], cmap = 'Greys')

```

```

Out[5]: <matplotlib.image.AxesImage at 0x1a809523100>

```



```
In [6]: train_labels[1235]
```

```
Out[6]: 9
```

```
In [7]: # 전처리 작업  
train_images = train_images.reshape((60000, 28, 28, 1))
```

```
In [8]: test_images = test_images.reshape((10000, 28, 28, 1))
```

```
In [9]: train_images.shape, test_images.shape
```

```
Out[9]: ((60000, 28, 28, 1), (10000, 28, 28, 1))
```

```
In [10]: train_images[35556]
```

[illegible]

[illegible]

[illegible]

$$\begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

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```

In [11]: # 0-255를 0과 1 사이의 실수로 변경
# train_images = train_images / 255.
# test_images = test_images / 255. -> 1줄로 처리
train_images, test_images = train_images / 255. , test_images / 255.

```

```

In [12]: train_images[35556]

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[illegible][illegible]
$$\begin{bmatrix} 0. \\ 0. \\ 0. \\ 0. \end{bmatrix},$$

[illegible][illegible]
$$\begin{bmatrix} [0. &] \\ [0. &] \\ [0. &] \\ [0. &] \\ [0. &] \\ [0. &] \end{bmatrix},$$

$$\begin{bmatrix} 0. \\ 0. \\ 0. \\ 0. \\ 0. \\ 0. \\ 0. \\ 0. \end{bmatrix},$$

$$\begin{bmatrix} [0. &], \\ [0. &], \\ [0. &], \\ [0. &], \\ [0. &], \\ [0. &], \\ [0. &], \\ [0.16862745], \\ [0.96078431], \\ [0.99607843], \end{bmatrix}$$

$$\begin{bmatrix} 0. \\ 0. \\ 0. \\ 0. \\ 0. \\ 0. \\ 0. \\ 0.17647059 \\ 0.96078431 \\ 0.99607843 \\ 0.55294118 \\ 0. \end{bmatrix},$$

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3. 신경망 만들기

```
In [13]: # Keras 모델 설계
model = models.Sequential()
# 32개의 레이어, 3X3 합성곱 필터, activation function은 relu(음수는 0으로, 양수는
model.add(layers.Conv2D(32, (3,3), activation='relu', input_shape = (28, 28, 1)))
model.add(layers.MaxPooling2D((2, 2))) # 절반으로
model.add(layers.Conv2D(64, (3, 3), activation = 'relu')) # CNN의 경우 3배로 늘리
model.add(layers.MaxPooling2D((2, 2))) # 절반으로
model.add(layers.Conv2D(64, (3, 3), activation = 'relu')) # CNN의 경우 3배로 늘리
model.add(layers.MaxPooling2D((2, 2))) # 절반으로
```

```
C:\Users\hi\anaconda3\envs\p310_cnn\lib\site-packages\keras\src\layers\convolutional\base_conv.py:107: UserWarning: Do not pass an `input_shape`/`input_dim` argument to a layer. When using Sequential models, prefer using an `Input(shape)` object as the first layer in the model instead.
  super().__init__(activity_regularizer=activity_regularizer, **kwargs)
```

```
In [14]: model.summary()
```

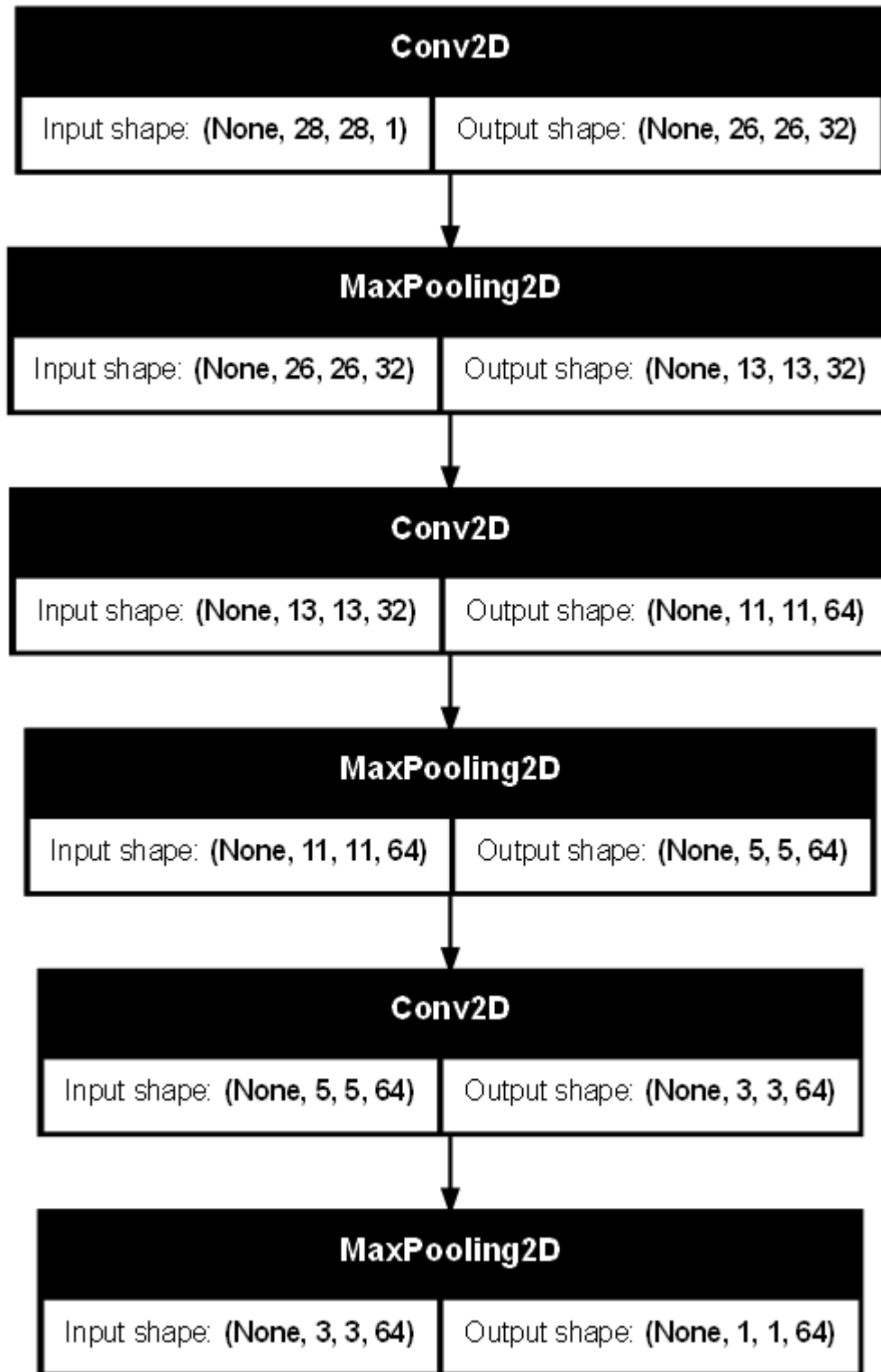
Model: "sequential"

Layer (type)	Output Shape
conv2d (Conv2D)	(None, 26, 26, 32)
max_pooling2d (MaxPooling2D)	(None, 13, 13, 32)
conv2d_1 (Conv2D)	(None, 11, 11, 64)
max_pooling2d_1 (MaxPooling2D)	(None, 5, 5, 64)
conv2d_2 (Conv2D)	(None, 3, 3, 64)
max_pooling2d_2 (MaxPooling2D)	(None, 1, 1, 64)

Total params: 55,744 (217.75 KB)
Trainable params: 55,744 (217.75 KB)
Non-trainable params: 0 (0.00 B)

```
In [16]: plot_model(model, show_shapes=True, dpi=80)
```


Out[16]:



```

In [18]: # 분류단계
model.add(layers.Flatten())
model.add(layers.Dense(64, activation='relu'))
model.add(layers.Dense(10, activation='softmax')) # 학습 시키0에서 9까지라서

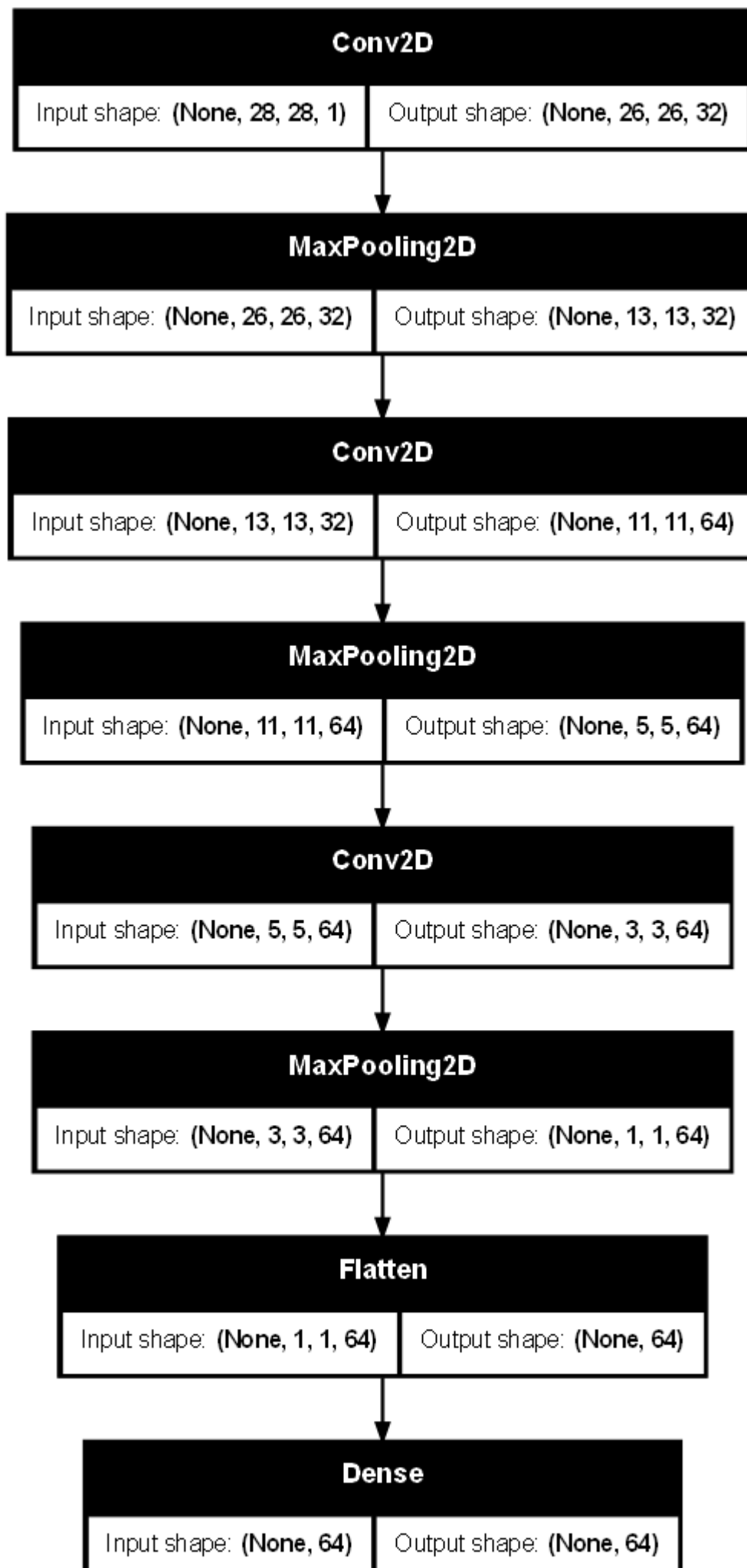
```

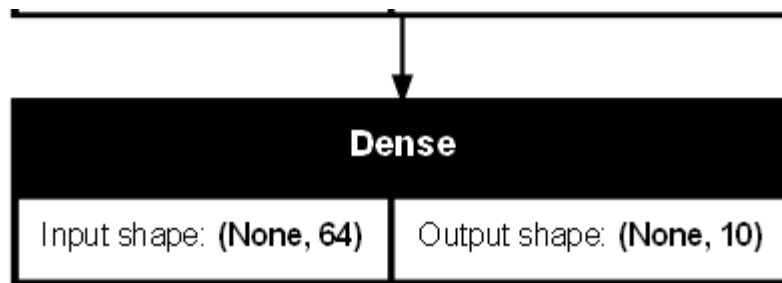
```

In [19]: plot_model(model, show_shapes=True, dpi=80)

```

Out[19]:





4. 모델 만들고 학습하기

```
In [21]: model.compile(  
    optimizer = 'adam',  
    loss = 'sparse_categorical_crossentropy',  
    metrics = ['accuracy']  
)
```

```
In [22]: # 모델 훈련 -> # fit = 훈련 / verbose = 학습 상황 표시  
history = model.fit(train_images, train_labels, batch_size = 16, epochs = 50,  
    verbose = 1, validation_data = (test_images, test_labels)  
)
```

Epoch 1/50
3750/3750 ————— 17s 4ms/step - accuracy: 0.8772 - loss: 0.3824 - val_accuracy: 0.9787 - val_loss: 0.0694

Epoch 2/50
3750/3750 ————— 15s 4ms/step - accuracy: 0.9782 - loss: 0.0723 - val_accuracy: 0.9794 - val_loss: 0.0674

Epoch 3/50
3750/3750 ————— 15s 4ms/step - accuracy: 0.9864 - loss: 0.0472 - val_accuracy: 0.9870 - val_loss: 0.0442

Epoch 4/50
3750/3750 ————— 16s 4ms/step - accuracy: 0.9883 - loss: 0.0355 - val_accuracy: 0.9858 - val_loss: 0.0523

Epoch 5/50
3750/3750 ————— 17s 4ms/step - accuracy: 0.9913 - loss: 0.0297 - val_accuracy: 0.9865 - val_loss: 0.0533

Epoch 6/50
3750/3750 ————— 15s 4ms/step - accuracy: 0.9930 - loss: 0.0215 - val_accuracy: 0.9872 - val_loss: 0.0463

Epoch 7/50
3750/3750 ————— 16s 4ms/step - accuracy: 0.9944 - loss: 0.0179 - val_accuracy: 0.9877 - val_loss: 0.0549

Epoch 8/50
3750/3750 ————— 15s 4ms/step - accuracy: 0.9952 - loss: 0.0162 - val_accuracy: 0.9835 - val_loss: 0.0697

Epoch 9/50
3750/3750 ————— 21s 4ms/step - accuracy: 0.9950 - loss: 0.0143 - val_accuracy: 0.9851 - val_loss: 0.0680

Epoch 10/50
3750/3750 ————— 16s 4ms/step - accuracy: 0.9956 - loss: 0.0134 - val_accuracy: 0.9819 - val_loss: 0.0822

Epoch 11/50
3750/3750 ————— 16s 4ms/step - accuracy: 0.9963 - loss: 0.0128 - val_accuracy: 0.9843 - val_loss: 0.0745

Epoch 12/50
3750/3750 ————— 16s 4ms/step - accuracy: 0.9967 - loss: 0.0097 - val_accuracy: 0.9872 - val_loss: 0.0627

Epoch 13/50
3750/3750 ————— 16s 4ms/step - accuracy: 0.9963 - loss: 0.0109 - val_accuracy: 0.9856 - val_loss: 0.0667

Epoch 14/50
3750/3750 ————— 17s 4ms/step - accuracy: 0.9971 - loss: 0.0097 - val_accuracy: 0.9885 - val_loss: 0.0574

Epoch 15/50
3750/3750 ————— 17s 4ms/step - accuracy: 0.9972 - loss: 0.0085 - val_accuracy: 0.9891 - val_loss: 0.0583

Epoch 16/50
3750/3750 ————— 16s 4ms/step - accuracy: 0.9973 - loss: 0.0083 - val_accuracy: 0.9859 - val_loss: 0.0848

Epoch 17/50
3750/3750 ————— 16s 4ms/step - accuracy: 0.9974 - loss: 0.0081 - val_accuracy: 0.9876 - val_loss: 0.0744

Epoch 18/50
3750/3750 ————— 17s 4ms/step - accuracy: 0.9980 - loss: 0.0072 - val_accuracy: 0.9874 - val_loss: 0.0749

Epoch 19/50
3750/3750 ————— 16s 4ms/step - accuracy: 0.9979 - loss: 0.0081 - val_accuracy: 0.9847 - val_loss: 0.0847

Epoch 20/50
3750/3750 ————— 17s 4ms/step - accuracy: 0.9974 - loss: 0.0082 - val_accuracy: 0.9879 - val_loss: 0.0809

Epoch 21/50
3750/3750 ————— 16s 4ms/step - accuracy: 0.9974 - loss: 0.0079 - val_accuracy: 0.9868 - val_loss: 0.0830

Epoch 22/50
3750/3750 ————— 20s 4ms/step - accuracy: 0.9976 - loss: 0.0088 - val_accuracy: 0.9868 - val_loss: 0.0702

Epoch 23/50
3750/3750 ————— 16s 4ms/step - accuracy: 0.9974 - loss: 0.0080 - val_accuracy: 0.9879 - val_loss: 0.0725

Epoch 24/50
3750/3750 ————— 16s 4ms/step - accuracy: 0.9977 - loss: 0.0084 - val_accuracy: 0.9886 - val_loss: 0.0859

Epoch 25/50
3750/3750 ————— 16s 4ms/step - accuracy: 0.9984 - loss: 0.0050 - val_accuracy: 0.9878 - val_loss: 0.0895

Epoch 26/50
3750/3750 ————— 16s 4ms/step - accuracy: 0.9980 - loss: 0.0062 - val_accuracy: 0.9852 - val_loss: 0.1030

Epoch 27/50
3750/3750 ————— 16s 4ms/step - accuracy: 0.9976 - loss: 0.0078 - val_accuracy: 0.9877 - val_loss: 0.0916

Epoch 28/50
3750/3750 ————— 15s 4ms/step - accuracy: 0.9979 - loss: 0.0074 - val_accuracy: 0.9868 - val_loss: 0.0843

Epoch 29/50
3750/3750 ————— 16s 4ms/step - accuracy: 0.9978 - loss: 0.0073 - val_accuracy: 0.9857 - val_loss: 0.1017

Epoch 30/50
3750/3750 ————— 17s 4ms/step - accuracy: 0.9977 - loss: 0.0070 - val_accuracy: 0.9860 - val_loss: 0.1050

Epoch 31/50
3750/3750 ————— 16s 4ms/step - accuracy: 0.9986 - loss: 0.0056 - val_accuracy: 0.9859 - val_loss: 0.1061

Epoch 32/50
3750/3750 ————— 17s 4ms/step - accuracy: 0.9982 - loss: 0.0061 - val_accuracy: 0.9852 - val_loss: 0.1165

Epoch 33/50
3750/3750 ————— 17s 4ms/step - accuracy: 0.9979 - loss: 0.0078 - val_accuracy: 0.9873 - val_loss: 0.1020

Epoch 34/50
3750/3750 ————— 17s 4ms/step - accuracy: 0.9988 - loss: 0.0049 - val_accuracy: 0.9860 - val_loss: 0.1180

Epoch 35/50
3750/3750 ————— 16s 4ms/step - accuracy: 0.9979 - loss: 0.0081 - val_accuracy: 0.9855 - val_loss: 0.1076

Epoch 36/50
3750/3750 ————— 16s 4ms/step - accuracy: 0.9978 - loss: 0.0093 - val_accuracy: 0.9851 - val_loss: 0.1460

Epoch 37/50
3750/3750 ————— 17s 4ms/step - accuracy: 0.9975 - loss: 0.0101 - val_accuracy: 0.9870 - val_loss: 0.1322

Epoch 38/50
3750/3750 ————— 17s 5ms/step - accuracy: 0.9980 - loss: 0.0107 - val_accuracy: 0.9868 - val_loss: 0.1117

Epoch 39/50
3750/3750 ————— 16s 4ms/step - accuracy: 0.9989 - loss: 0.0051 - val_accuracy: 0.9886 - val_loss: 0.1065

Epoch 40/50
3750/3750 ————— 16s 4ms/step - accuracy: 0.9983 - loss: 0.0070 - val_accuracy: 0.9853 - val_loss: 0.1352

```

Epoch 41/50
3750/3750 ————— 15s 4ms/step - accuracy: 0.9983 - loss: 0.0067 - v
al_accuracy: 0.9870 - val_loss: 0.1248
Epoch 42/50
3750/3750 ————— 16s 4ms/step - accuracy: 0.9987 - loss: 0.0052 - v
al_accuracy: 0.9864 - val_loss: 0.1252
Epoch 43/50
3750/3750 ————— 16s 4ms/step - accuracy: 0.9988 - loss: 0.0051 - v
al_accuracy: 0.9878 - val_loss: 0.1090
Epoch 44/50
3750/3750 ————— 16s 4ms/step - accuracy: 0.9986 - loss: 0.0043 - v
al_accuracy: 0.9852 - val_loss: 0.1398
Epoch 45/50
3750/3750 ————— 16s 4ms/step - accuracy: 0.9984 - loss: 0.0072 - v
al_accuracy: 0.9865 - val_loss: 0.1286
Epoch 46/50
3750/3750 ————— 16s 4ms/step - accuracy: 0.9988 - loss: 0.0064 - v
al_accuracy: 0.9881 - val_loss: 0.1124
Epoch 47/50
3750/3750 ————— 16s 4ms/step - accuracy: 0.9985 - loss: 0.0056 - v
al_accuracy: 0.9866 - val_loss: 0.1288
Epoch 48/50
3750/3750 ————— 16s 4ms/step - accuracy: 0.9989 - loss: 0.0055 - v
al_accuracy: 0.9885 - val_loss: 0.1159
Epoch 49/50
3750/3750 ————— 17s 4ms/step - accuracy: 0.9985 - loss: 0.0059 - v
al_accuracy: 0.9886 - val_loss: 0.1218
Epoch 50/50
3750/3750 ————— 16s 4ms/step - accuracy: 0.9987 - loss: 0.0053 - v
al_accuracy: 0.9876 - val_loss: 0.1330

```

```

In [27]: # 훈련결과 정보 확인
         history.history.keys()

```

```

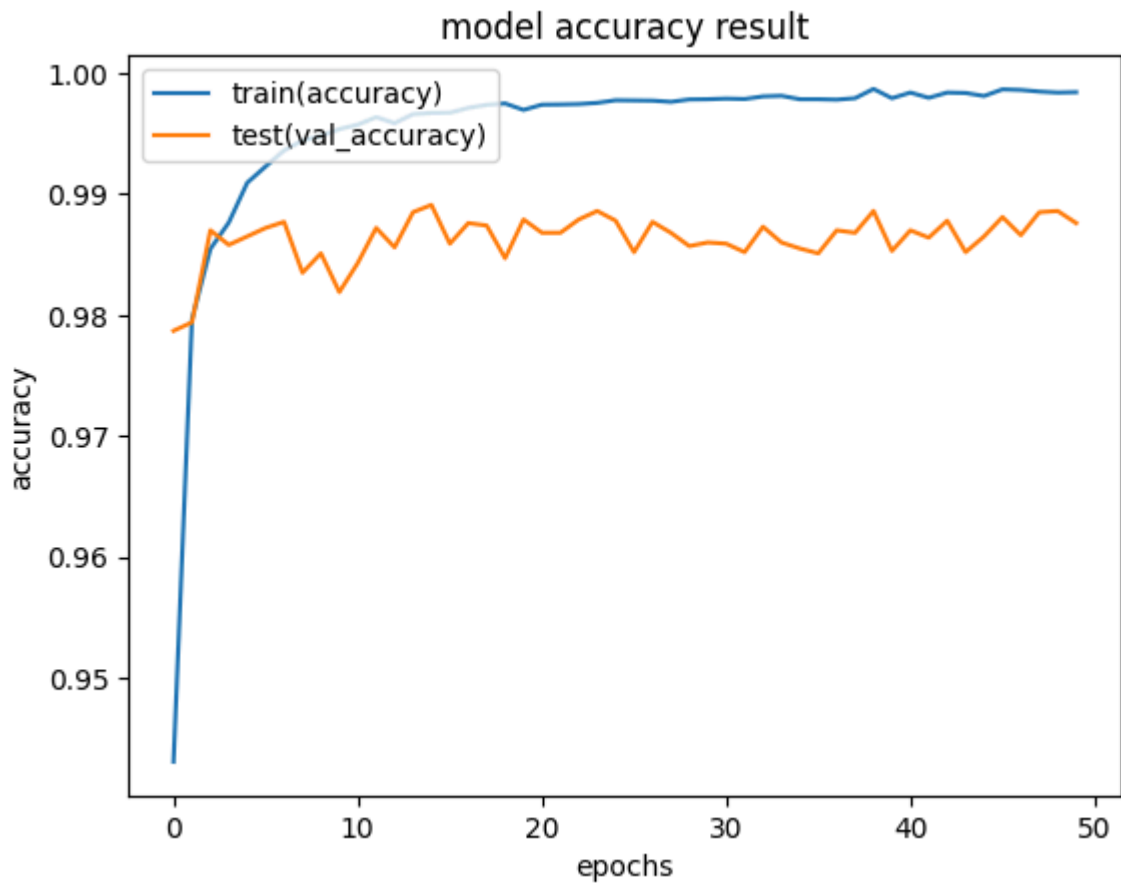
Out[27]: dict_keys(['accuracy', 'loss', 'val_accuracy', 'val_loss'])

```

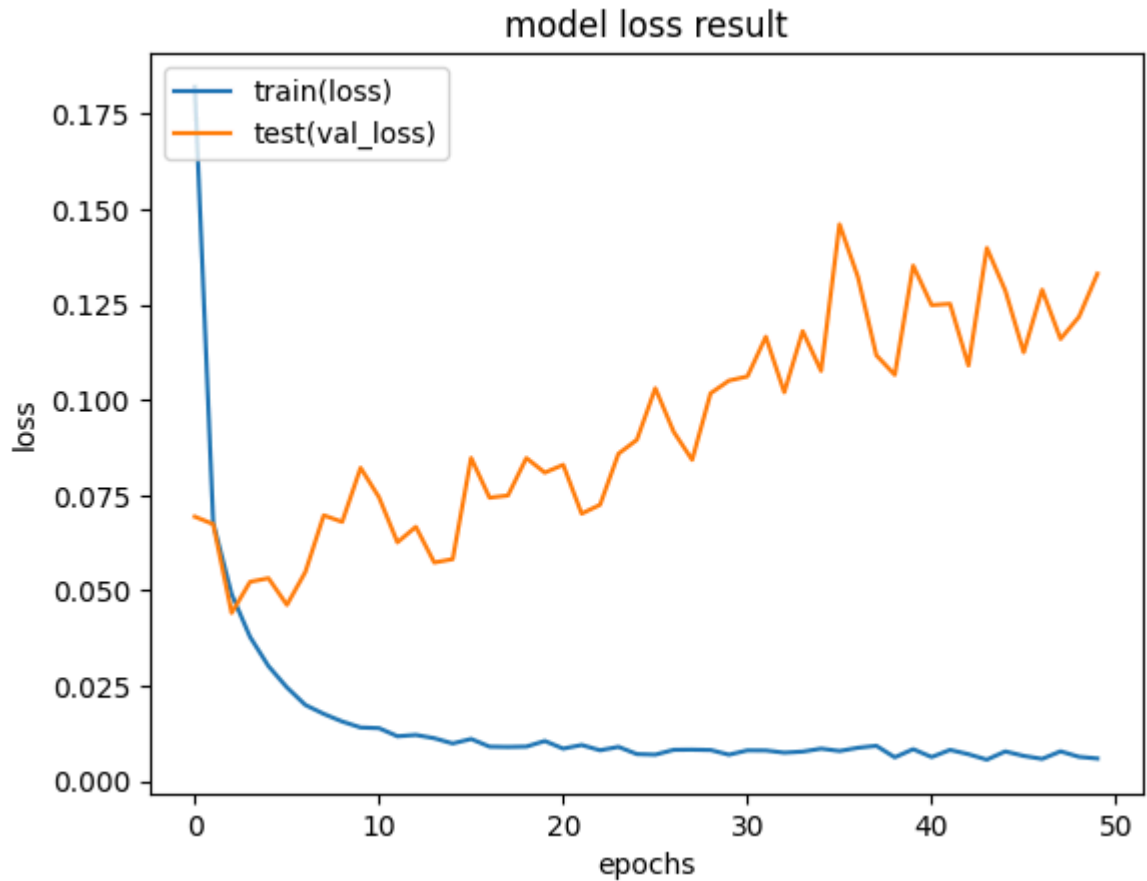
```

In [31]: # 훈련결과 그래프 표시
         plt.plot(history.history['accuracy'])
         plt.plot(history.history['val_accuracy'])
         plt.title('model accuracy result')
         plt.ylabel('accuracy') # y축
         plt.xlabel('epochs') # x축
         plt.legend(['train(accuracy)', 'test(val_accuracy)'], loc = 'upper left')
         # plt.legend(['accuracy', 'val_accuracy'], loc = 'upper left')
         plt.show()

```



```
In [32]: # 훈련결과 그래프 표시
plt.plot(history.history['loss'])
plt.plot(history.history['val_loss'])
plt.title('model loss result')
plt.ylabel('loss') # y축
plt.xlabel('epochs') # x축
plt.legend(['train(loss)', 'test(val_loss)'], loc = 'upper left')
# plt.legend(['accuracy', 'val_accuracy'], loc = 'upper left')
plt.show()
```



5. 모델 적용하기(예측 : Predict)

```
In [33]: # 모델 평가 - 문제와 답이 필요함
loss, acc = model.evaluate(test_images, test_labels, verbose=2) # verbose -> 1 :
```

313/313 - 1s - 3ms/step - accuracy: 0.9876 - loss: 0.1330

```
In [38]: acc # 98.76%
```

```
Out[38]: 0.9876000285148621
```

```
In [40]: # 모델 예측 - 문제만 있으면 됨
result = model.predict(test_images) # 10,000개
```

313/313 ————— 1s 2ms/step

```
In [43]: result[5000]
```

```
Out[43]: array([1.38281875e-29, 1.81292580e-19, 3.99213954e-22, 1.00000000e+00,
 7.19171071e-34, 1.32578095e-15, 5.60544591e-25, 1.27679835e-20,
 1.65464339e-16, 1.73610397e-30], dtype=float32)
```

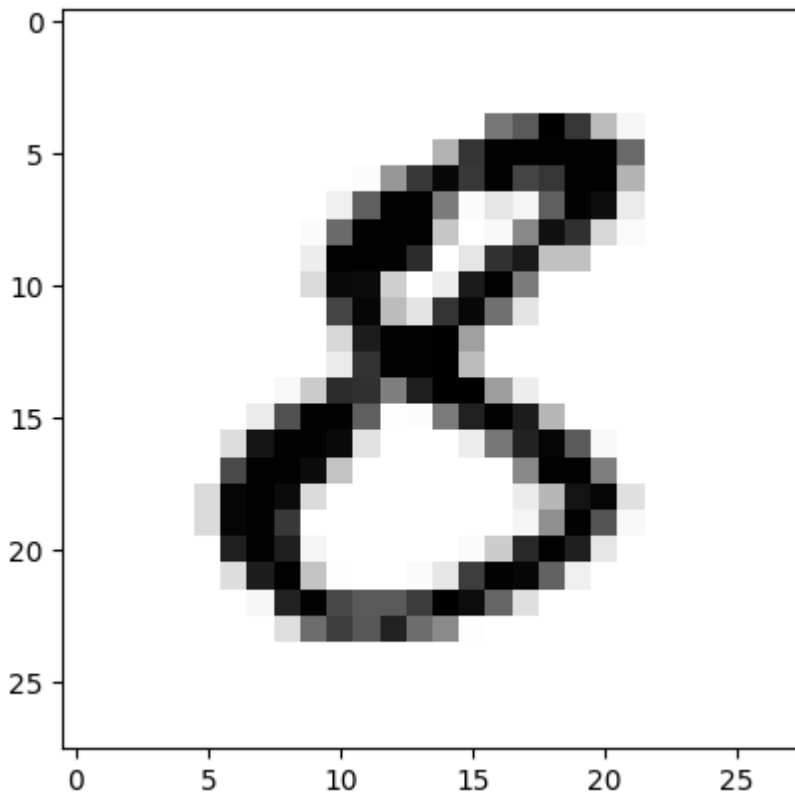
```
In [48]: result_value = np.argmax(result[1234])
```

```
In [49]: print("예측 결과값 = %d" % (result_value))
```

예측 결과값 = 8

```
In [50]: plt.imshow(test_images[1234], cmap = 'Greys')
```


Out[50]: <matplotlib.image.AxesImage at 0x1a8329d29e0>



In []: # 실제로 사용할 때 모델 저장부터 하면 됨 -> 학습은 이미 끝났음

In [51]: # 모델 저장
model.save('mnist_cnn_20240517_epochs_50_9876.h5') # FileName : dqte, epochs, ac

WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or `keras.saving.save_model(model)`. This file format is considered legacy. We recommend using instead the native Keras format, e.g. `model.save('my_model.keras')` or `keras.saving.save_model(model, 'my_model.keras')`.

6. 실제 데이터 파일 검증

In [53]: # 이미지 불러오기
image = Image.open('6.jpg')

In [55]: # 이미지 흑백으로
image = image.convert('L') # 대문자 L -> 컬러에서 흑백으로

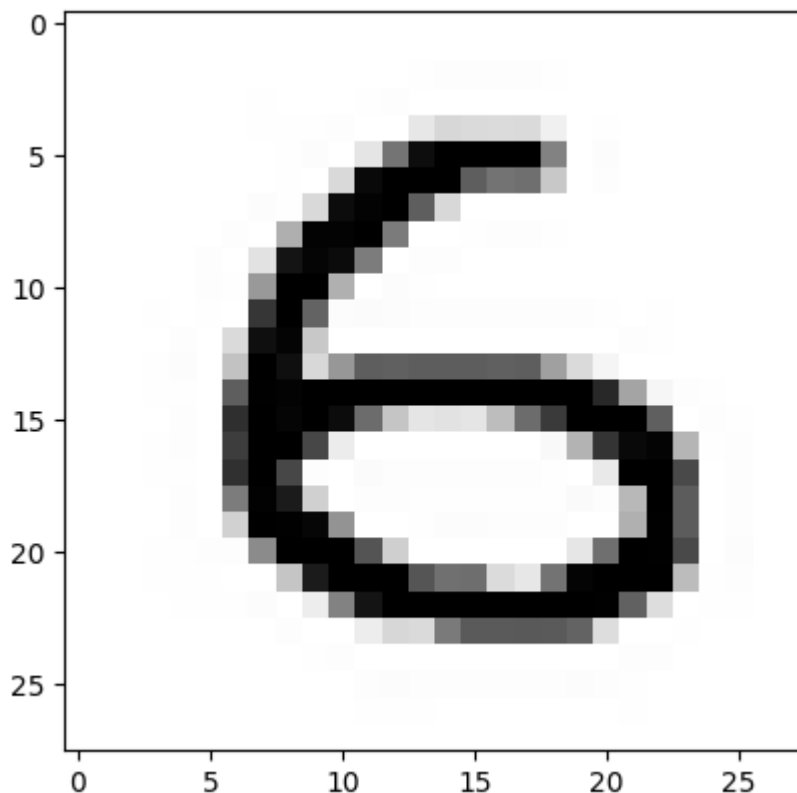
In [58]: # 인공지능에 들어갈 수 있는 형태로 변환
data = np.ndarray(shape=(1, 28, 28, 1), dtype = np.float32)
size = (28, 28)
image = ImageOps.fit(image, size, Image.LANCZOS) # 사이즈 맞춤 : 이미지, 사이즈,

image_array = np.array(image)
전처리
nor_image_array = (image_array.astype(np.float32) / 255.) # or 127.) -1
data = nor_image_array
img_data = data.reshape(1, 28, 28, 1)
img_data.shape

Out[58]: (1, 28, 28, 1)

In [59]: `plt.imshow(image, cmap = 'Greys')`

Out[59]: `<matplotlib.image.AxesImage at 0x1a832bf6d70>`



In [60]: `# 예측
result = model.predict(img_data)`

1/1 ————— 0s 17ms/step

In [61]: `result`

Out[61]: `array([[3.5319815e-17, 8.4430869e-35, 0.0000000e+00, 0.0000000e+00,
2.2788079e-26, 2.2093388e-13, 1.0000000e+00, 0.0000000e+00,
1.2352133e-21, 1.5133417e-26]], dtype=float32)`

In [62]: `rev_value = np.argmax(result)`

In [63]: `print("실제 이미지 예측 결과 = %d" % (rev_value))`

실제 이미지 예측 결과 = 6