

Chapter 43. 컨볼루션 네트워크를 다시 들여다 보자



다시 MNIST using CNN

다시 MNIST 데이터를 가지고 오고~

```
import tensorflow as tf

mnist = tf.keras.datasets.mnist

(x_train, y_train), (x_test, y_test) = mnist.load_data()
X_train, X_test = x_train/255, x_test/255

X_train = X_train.reshape((60000, 28, 28, 1))
X_test = X_test.reshape((10000, 28, 28, 1))
```

이번에는 구조를 간단히 가지고 오고~

```
from tensorflow.keras import layers, models

model = models.Sequential([
    layers.Conv2D(3, kernel_size=(3,3), strides=(1,1),
                  padding='same', activation='relu',
                  input_shape=(28,28,1)),
    layers.MaxPool2D(pool_size=(2,2), strides=(2,2)),
    layers.Dropout(0.25),
    layers.Flatten(),
    layers.Dense(1000, activation='relu'),
    layers.Dense(10, activation='softmax')
])
```

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 28, 28, 3)	30
max_pooling2d (MaxPooling2D)	(None, 14, 14, 3)	0
dropout (Dropout)	(None, 14, 14, 3)	0
flatten (Flatten)	(None, 588)	0
dense (Dense)	(None, 1000)	589000
dense_1 (Dense)	(None, 10)	10010

Total params: 599,040

Trainable params: 599,040

Non-trainable params: 0

내가 구성한 layers들은 이렇게 호출할 수도 있다

```
model.layers
```

```
[<tensorflow.python.keras.layers.convolutional.Conv2D at 0x7f865077e250>,
 <tensorflow.python.keras.layers.pooling.MaxPooling2D at 0x7f865079efd0>,
 <tensorflow.python.keras.layers.core.Dropout at 0x7f868120b510>,
 <tensorflow.python.keras.layers.core.Flatten at 0x7f868120b790>,
 <tensorflow.python.keras.layers.core.Dense at 0x7f868120ba10>,
 <tensorflow.python.keras.layers.core.Dense at 0x7f868120be10>]
```

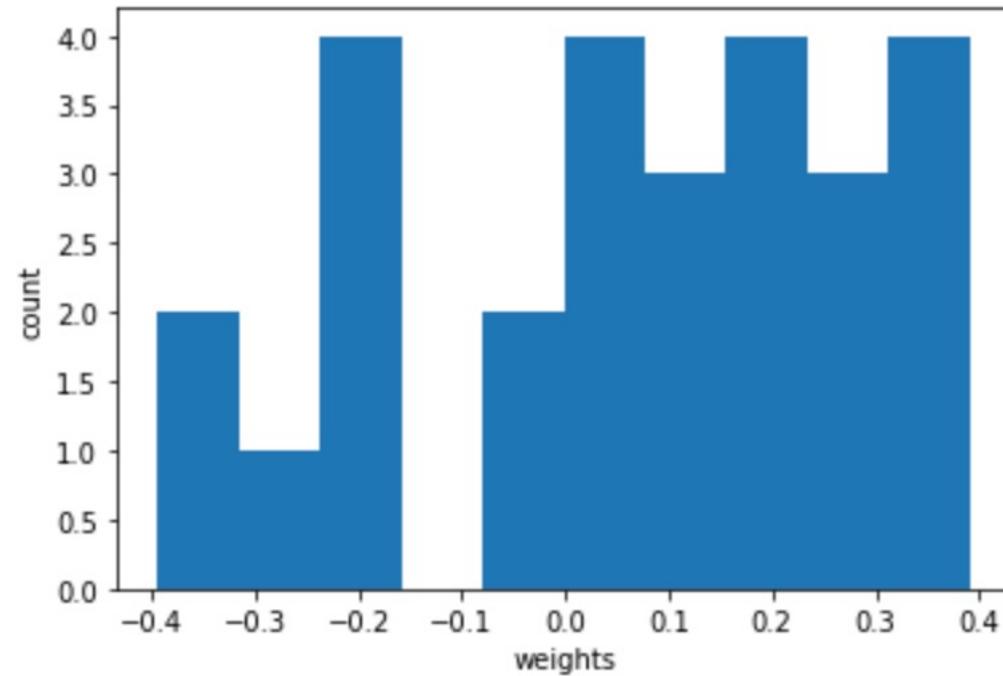
아직 학습하지 않은 conv 레이어의 웨이트의 평균

```
conv = model.layers[0]
conv_weights = conv.weights[0].numpy()
conv_weights.mean(), conv_weights.std()
```

(0.05563381, 0.22272052)

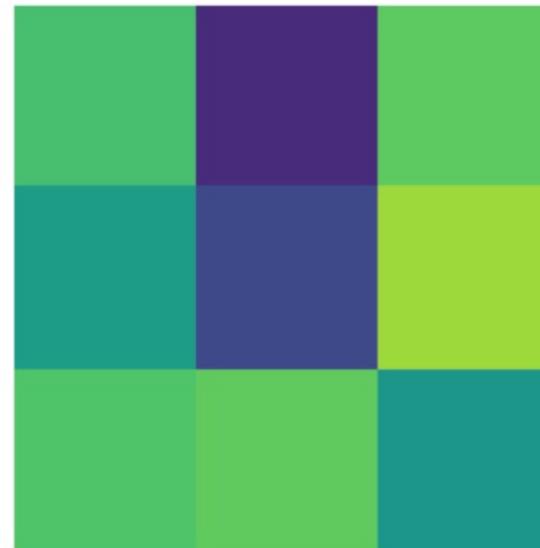
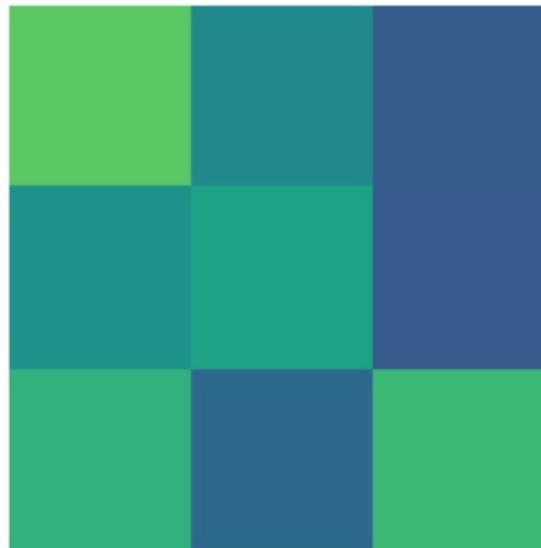
```
import matplotlib.pyplot as plt

plt.hist(conv_weights.reshape(-1,1))
plt.xlabel('weights')
plt.ylabel('count')
plt.show()
```



```
fig, ax = plt.subplots(1,3, figsize=(15,5))
for i in range(3):
    ax[i].imshow(conv_weights[:, :, 0, i], vmin=-0.5, vmax=0.5)
    ax[i].axis('off')

plt.show()
```



학습도 빨리 걸릴 것이고~

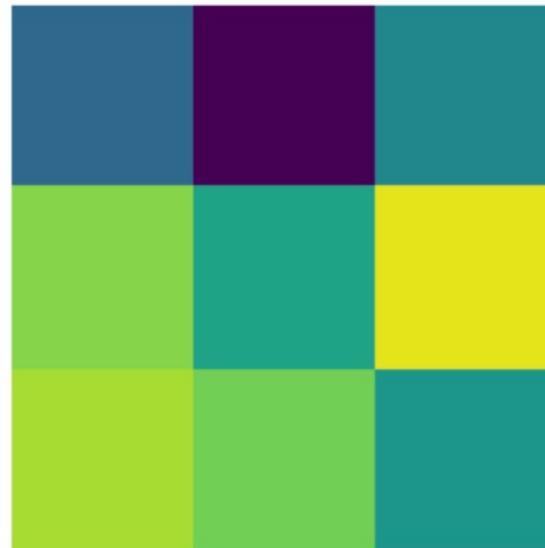
```
%%time  
  
model.compile(optimizer='adam', loss='sparse_categorical_crossentropy',  
               metrics=['accuracy'])  
  
hist = model.fit(X_train, y_train, epochs=5, verbose=1,  
                  validation_data = (X_test, y_test))
```

```
Epoch 1/5  
1875/1875 [=====] - 10s 5ms/step - loss: 0.  
4349 - accuracy: 0.8703 - val_loss: 0.0828 - val_accuracy: 0.9739
```

학습 후 conv filter의 변화

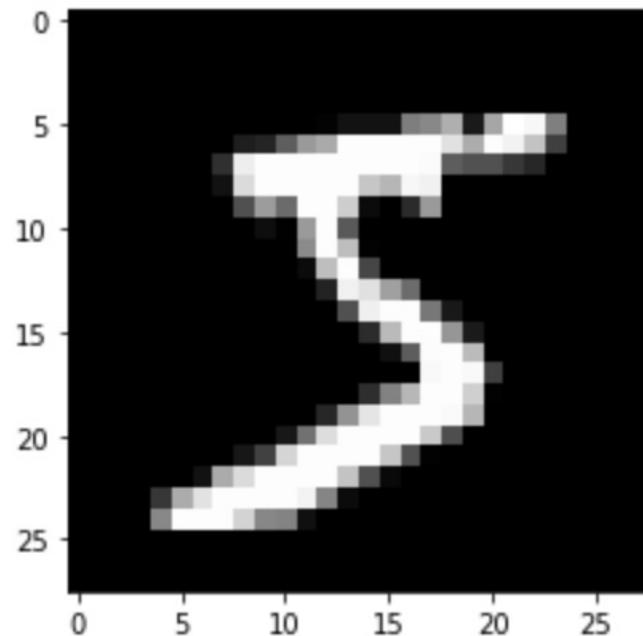
```
fig, ax = plt.subplots(1,3, figsize=(15,5))
for i in range(3):
    ax[i].imshow(conv_weights[:, :, 0, i], vmin=-0.5, vmax=0.5)
    ax[i].axis('off')

plt.show()
```



0번 데이터는 5

```
| plt.imshow(X_train[0], cmap='gray');
```



Conv 레이어에서 출력을 뽑고~

```
inputs = X_train[0].reshape(-1, 28, 28, 1)
conv_layer_output = tf.keras.Model(model.input, model.layers[0].output)
conv_layer_output.summary()
```

Model: "model"

Layer (type)	Output Shape	Param #
conv2d_input (InputLayer)	[(None, 28, 28, 1)]	0
conv2d (Conv2D)	(None, 28, 28, 3)	30

Total params: 30

Trainable params: 30

Non-trainable params: 0

입력에 대한 feature map을 뽑고~

```
| feature_maps = conv_layer_output.predict(inputs)  
feature_maps.shape
```

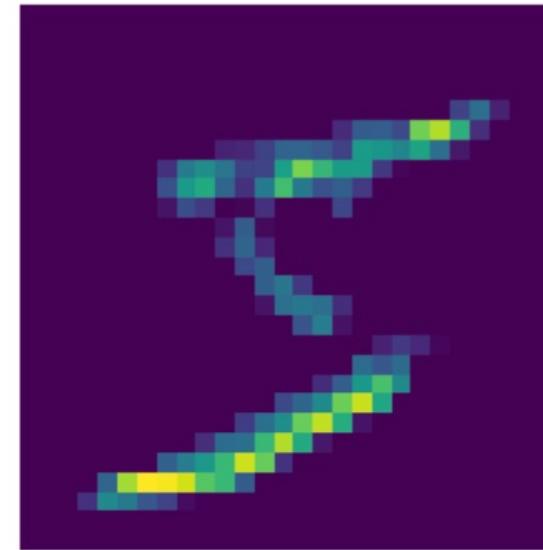
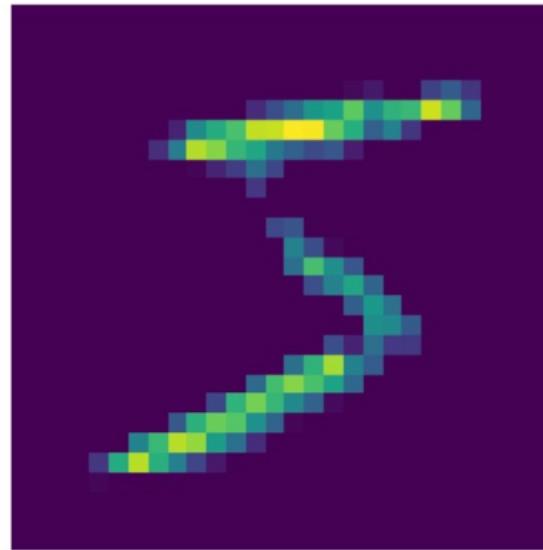
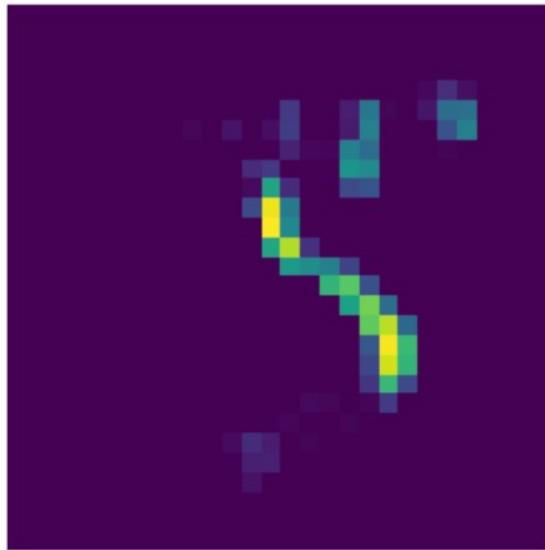
(1, 28, 28, 3)

```
| feature_maps[0, :, :, 0].shape
```

(28, 28)

Feature map이 본 숫자 5

```
| fig, ax = plt.subplots(1,3, figsize=(15,5))
| for i in range(3):
|     ax[i].imshow(feature_maps[0, :, :, i])
|     ax[i].axis('off')
|
| plt.show()
```



방금 전 과정을 함수로 만들고~

```
def draw_feature_maps(n):
    inputs = X_train[n].reshape(-1, 28, 28, 1)
    feature_maps = conv_layer_output.predict(inputs)

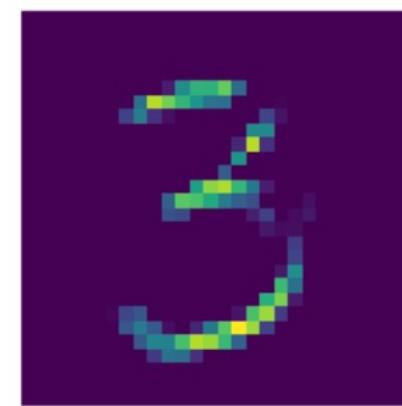
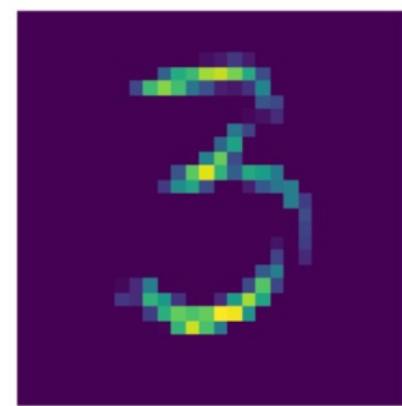
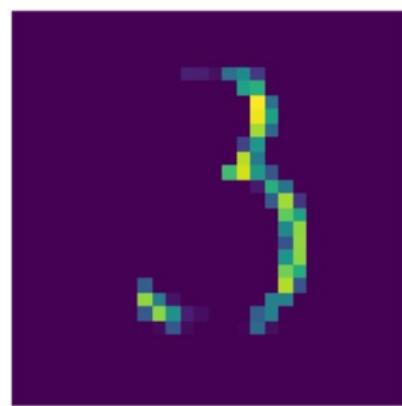
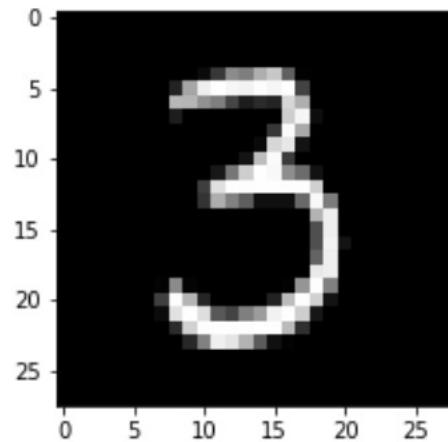
    fig, ax = plt.subplots(1,4, figsize=(15,5))

    ax[0].imshow(inputs[0, :, :, 0], cmap='gray');
    for i in range(1, 4):
        ax[i].imshow(feature_maps[0, :, :, i-1])
        ax[i].axis('off')

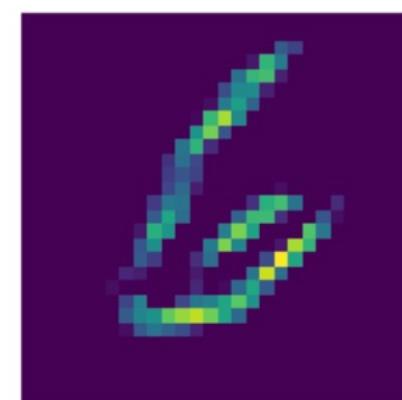
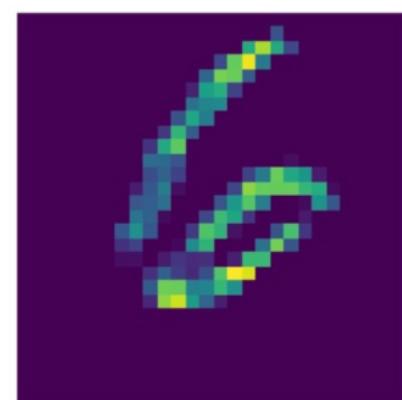
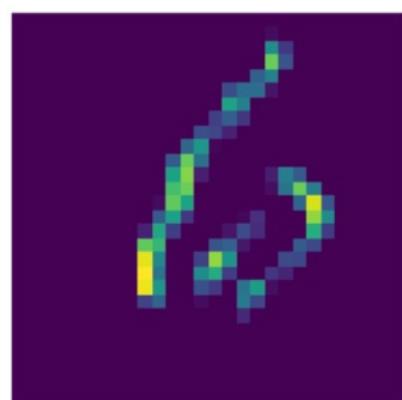
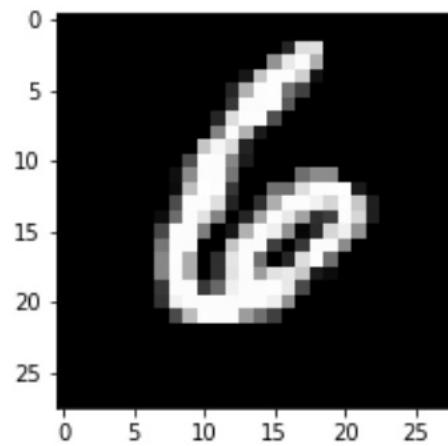
    plt.show()
```

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draw_feature_maps(50)

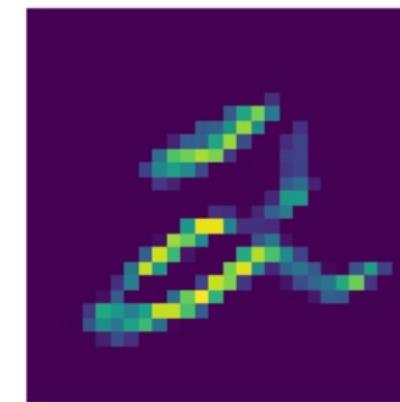
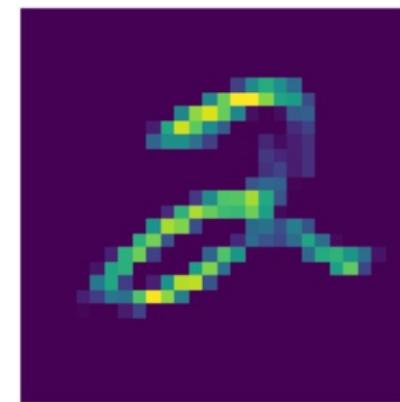
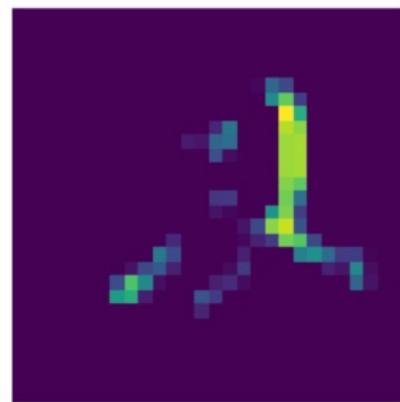
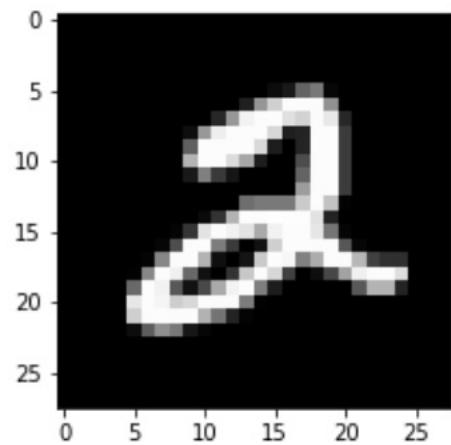


draw_feature_maps(13)

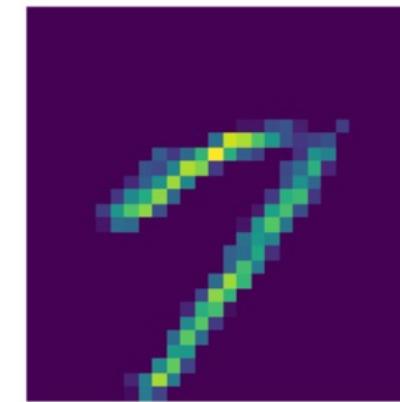
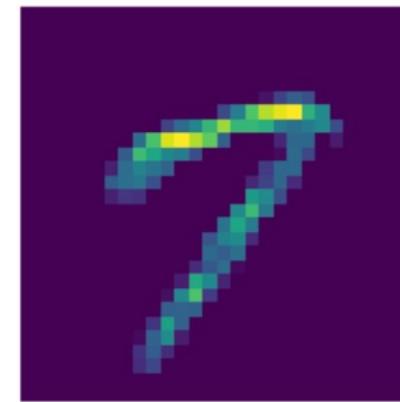
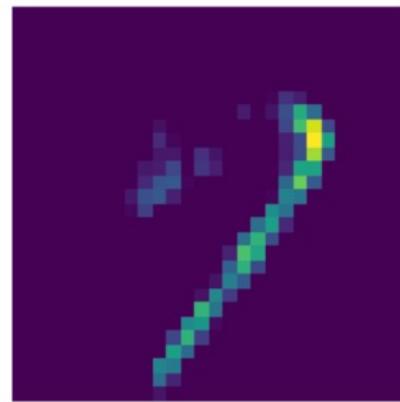
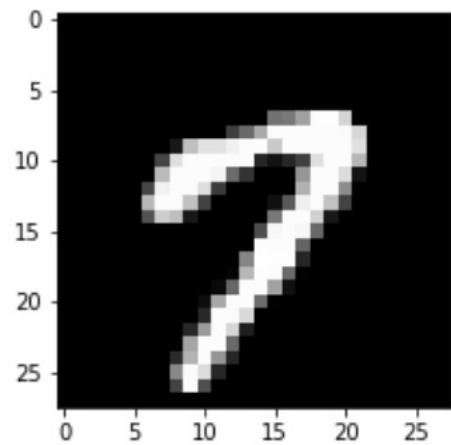


Ch

draw_feature_maps(5)



draw_feature_maps(15)



이번에는 모델 채널을 증가시키고

```
model1 = models.Sequential([
    layers.Conv2D(8, kernel_size=(3,3), strides=(1,1), padding='same',
                  activation='relu', input_shape=(28,28,1)),
    layers.MaxPool2D(pool_size=(2,2), strides=(2,2)),
    layers.Dropout(0.25),
    layers.Flatten(),
    layers.Dense(1000, activation='relu'),
    layers.Dense(10, activation='softmax')
])
```

```
%%time  
model1.compile(optimizer='adam', loss='sparse_categorical_crossentropy',  
                metrics=['accuracy'])  
hist = model1.fit(X_train, y_train, epochs=5, verbose=1,  
                   validation_data = (X_test, y_test))
```

Epoch 1/5
1875/1875 [=====] - 28s 15ms/step - loss: 0.3007
- accuracy: 0.9083 - val_loss: 0.0661 - val_accuracy: 0.9789
Epoch 2/5
1875/1875 [=====] - 28s 15ms/step - loss: 0.0697

```
conv_layer_output = tf.keras.Model(model1.input, model1.layers[0].output)

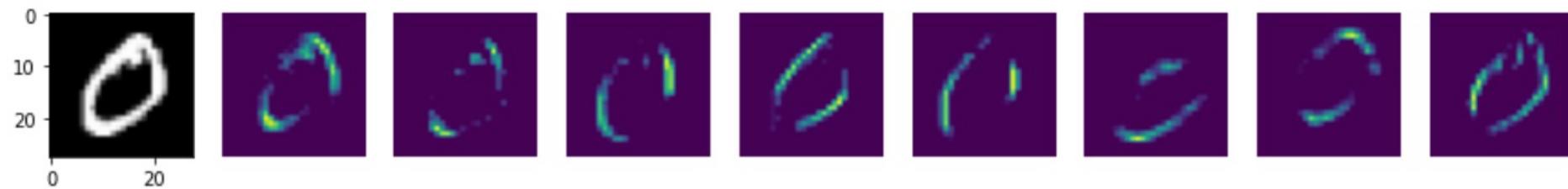
def draw_feature_maps(n):
    inputs = X_train[n].reshape(-1, 28, 28, 1)
    feature_maps = conv_layer_output.predict(inputs)

    fig, ax = plt.subplots(1, 9, figsize=(15, 5))

    ax[0].imshow(inputs[0, :, :, 0], cmap='gray');
    for i in range(1, 9):
        ax[i].imshow(feature_maps[0, :, :, i-1])
        ax[i].axis('off')

    plt.show()
```

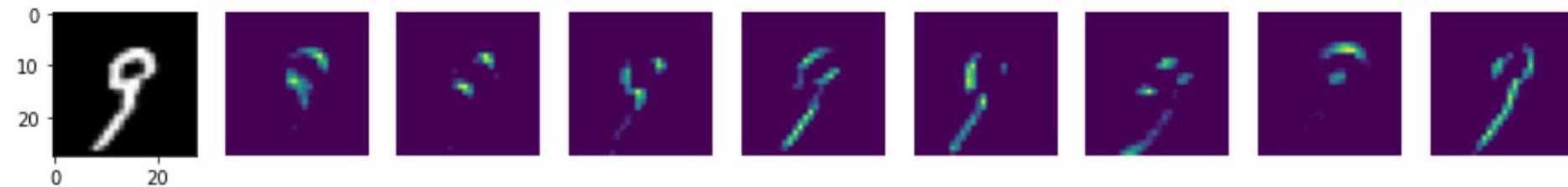
```
| draw_feature_maps(1)
```



```
| draw_feature_maps(13)
```



draw_feature_maps(19)



draw_feature_maps(25)

