

13주차(1/2)

ML Open Framework: Tensorflow & Keras

파이썬으로 배우는 기계학습

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기계학습 오픈 프레임워크: 텐서플로우 & 케라스

- 학습 목표
 - 텐서플로우 & 케라스 (**Tensorflow & Keras**)를 이해한다.
 - **Tensorflow & Keras**를 이용하여 엠니스트(**MNIST**) 데이터를 분석한다.
 - **Tensorflow & Keras**를 이용하여 합성곱 신경망(**CNN**)을 구현한다.
- 학습 내용
 - 기계학습 오픈 프레임워크 소개
 - **Tensorflow & Keras**를 살펴본다.
 - **Tensorflow & Keras**를 이용한 **MNIST** 데이터 처리
 - **Tensorflow & Keras**를 이용한 합성곱 신경망(**CNN**) 구현

1. 기계학습 오픈 프레임워크 종류: 5가지 종류



1. 기계학습 오픈 프레임워크: 텐서플로우 & 케라스

■ TensorFlow

- C++, Python 기반
- 합성곱 신경망(CNN)과 순환 신경망(RNN) 구현
- CPU, GPU 환경 모두 동작

■ Keras

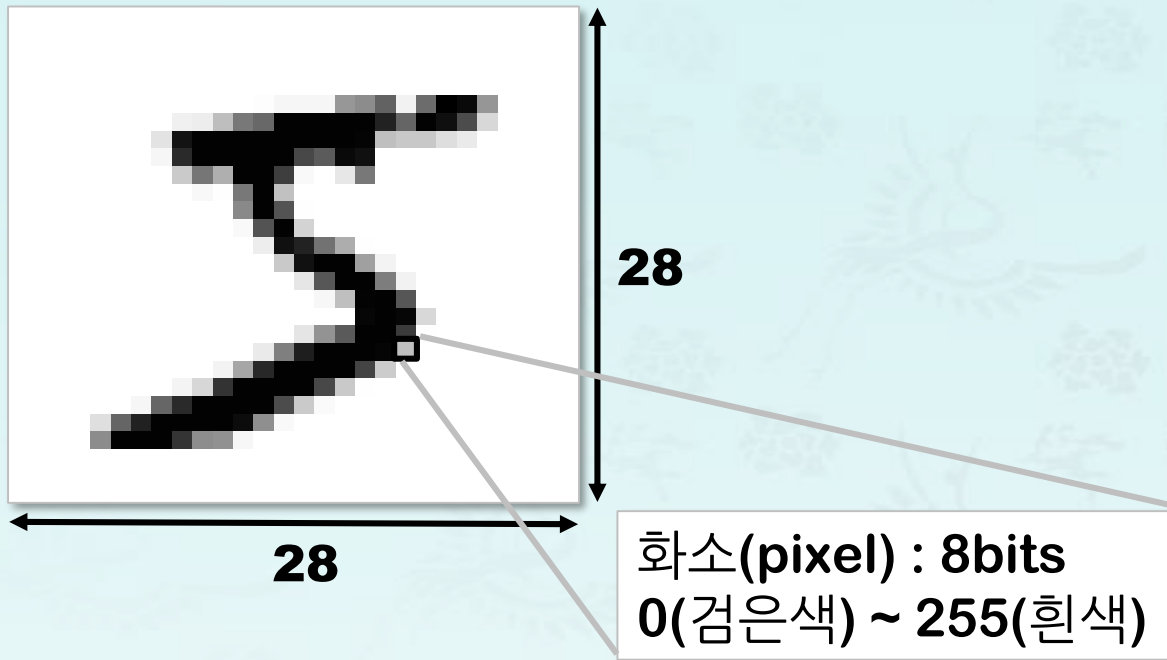
- 오픈 프레임워크 API
- Python 기반
- 문법이 간단하고 직관적



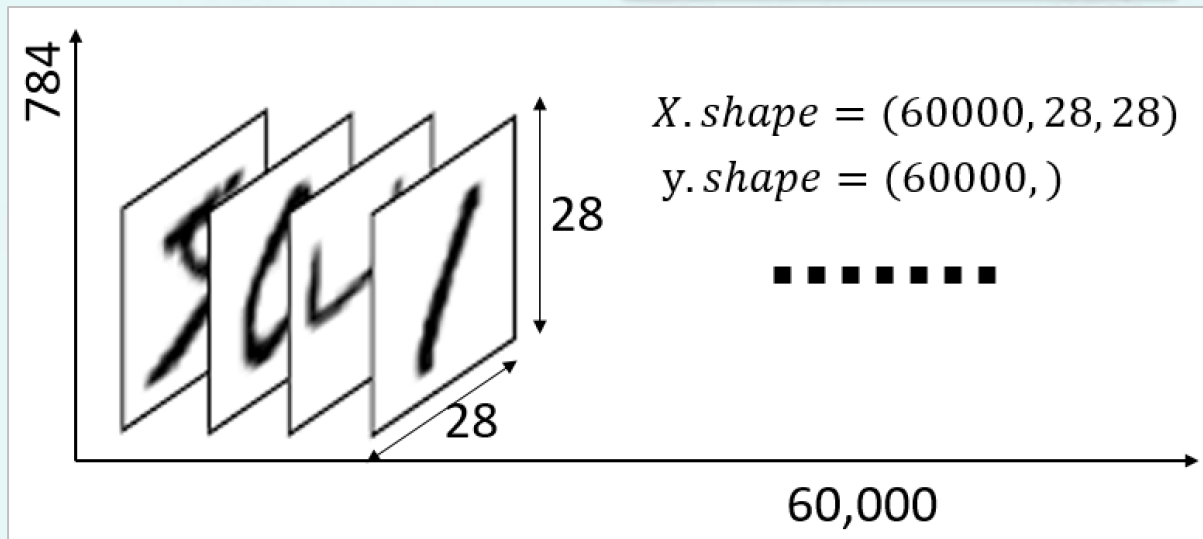
2. MNIST 데이터 분석: 데이터 읽어오기

```
1 (X_train, y_train), (X_test, y_test) =  
2   tf.keras.datasets.mnist.load_data()
```

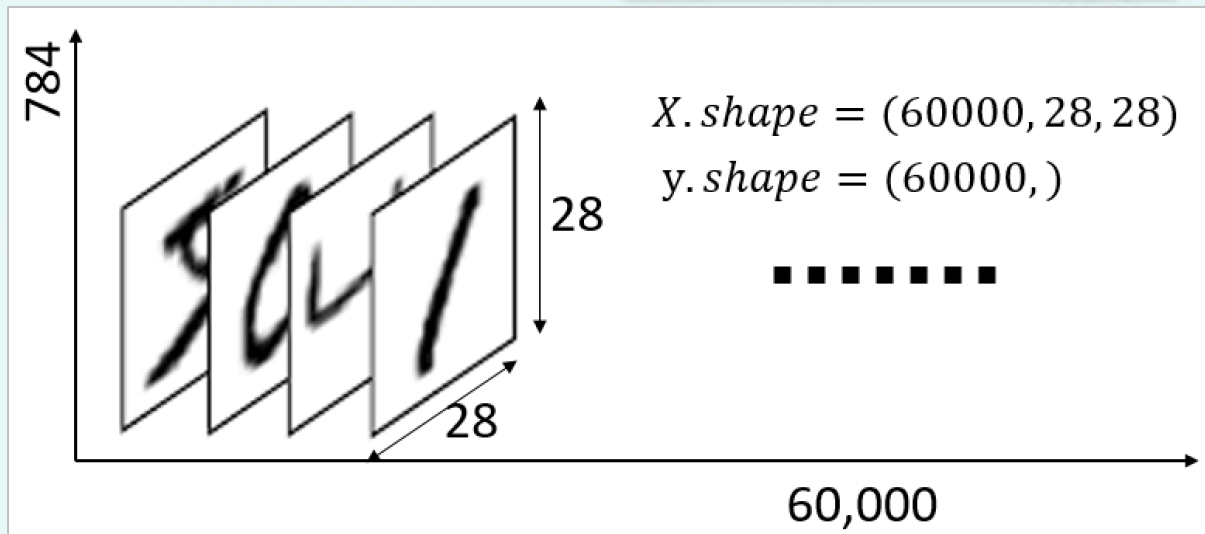
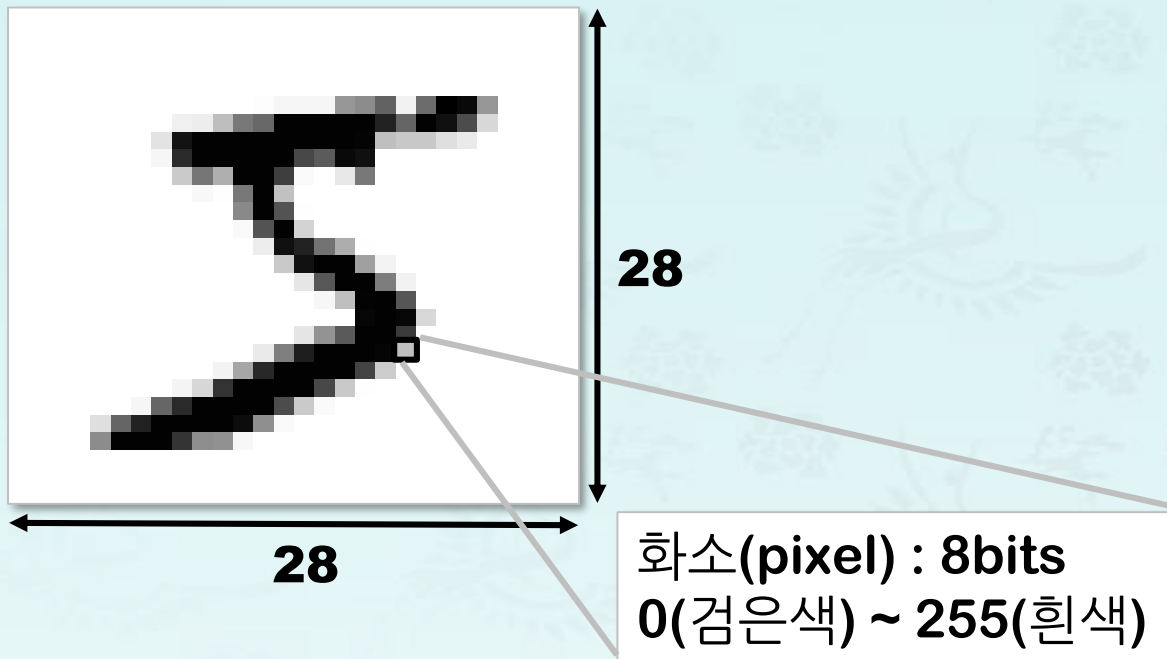
2. MNIST 데이터 분석: 데이터 읽어오기



```
1 (X_train, y_train), (X_test, y_test) =  
2 tf.keras.datasets.mnist.load_data()
```



2. MNIST 데이터 분석: 데이터 읽어오기



```
1 (X_train, y_train), (X_test, y_test) =  
2 tf.keras.datasets.mnist.load_data()
```

```
1 print(f"X_train.shape: {X_train.shape}")  
2 print(f"y_train.shape: {y_train.shape}")  
3 print(f"X_test.shape: {X_test.shape}")  
4 print(f"y_test.shape: {y_test.shape}")
```



```
X_train.shape (60000, 28, 28)  
y_train.shape (60000,)  
X_test.shape (10000, 28, 28)  
y_test.shape (10000,)
```

2. MNIST 데이터 분석: 데이터 전처리

- 정규화(Normalization)

```
1 X_train = X_train.astype('float32')/255
2 X_test = X_test.astype('float32')/255
```

- 원-핫 인코딩(One-Hot Encoding)

2. MNIST 데이터 분석: 데이터 전처리

- 정규화(Normalization)

```
1 X_train = X_train.astype('float32')/255
2 X_test = X_test.astype('float32')/255
```

- 원-핫 인코딩(One-Hot Encoding)

```
1 print(f"previous five labels in y_train: {y_train[:5]}")
2 y_train = tf.keras.utils.to_categorical(y_train, 10)
3 y_test = tf.keras.utils.to_categorical(y_test, 10)
4
5 print(f"One-hot encoded labels of y_train: \n{y_train[:5]}")
```



```
previous five labels in y_train: [5 0 4 1 9]
One-hot encoded labels of y_train:
[[0. 0. 0. 0. 0. 1. 0. 0. 0. 0.]
 [1. 0. 0. 0. 0. 0. 0. 0. 0. 0.]
 [0. 0. 0. 0. 1. 0. 0. 0. 0. 0.]
 [0. 1. 0. 0. 0. 0. 0. 0. 0. 0.]
 [0. 0. 0. 0. 0. 0. 0. 0. 0. 1.]]
```

2. MNIST 데이터 분석: 신경망 구축

- 순차모델(Sequential Model)

```
1 # define the model
2 # input_shape = (28, 28)
3 model = tf.keras.models.Sequential([
4     tf.keras.layers.Flatten(input_shape=X_train.shape[1:]),
5     tf.keras.layers.Dense(512, activation='relu'),
6     tf.keras.layers.Dropout(0.2),
7     tf.keras.layers.Dense(10, activation='softmax')
8 ])
9
10 # summarize the model
11 model.summary()
```

2. MNIST 데이터 분석: 신경망 구축

■ 입력층

- **X.shape = (60000, 28, 28)**
- **X.shape[1:] = (28, 28)**
- **노드(뉴론)의 수: $784 = 28 \times 28$**

```
1 # define the model
2 # input_shape = (28, 28)
3 model = tf.keras.models.Sequential([
4     tf.keras.layers.Flatten(input_shape=X_train.shape[1:]),
5     tf.keras.layers.Dense(512, activation='relu'),
6     tf.keras.layers.Dropout(0.2),
7     tf.keras.layers.Dense(10, activation='softmax')
8 ])
9
10 # summarize the model
11 model.summary()
```

2. MNIST 데이터 분석: 신경망 구축

- 은닉층 – Dense()
 - 노드(뉴론)의 수: 512
 - 활성화 함수: ReLU

```
1 # define the model
2 # input_shape = (28, 28)
3 model = tf.keras.models.Sequential([
4     tf.keras.layers.Flatten(input_shape=X_train.shape[1:]),
5     tf.keras.layers.Dense(512, activation='relu'),
6     tf.keras.layers.Dropout(0.2),
7     tf.keras.layers.Dense(10, activation='softmax')
8 ])
9
10 # summarize the model
11 model.summary()
```

2. MNIST 데이터 분석: 신경망 구축

- Dropout()
 - 드롭아웃 비율: 0.2 (20%)

```
1 # define the model
2 # input_shape = (28, 28)
3 model = tf.keras.models.Sequential([
4     tf.keras.layers.Flatten(input_shape=X_train.shape[1:]),
5     tf.keras.layers.Dense(512, activation='relu'),
6     tf.keras.layers.Dropout(0.2),
7     tf.keras.layers.Dense(10, activation='softmax')
8 ])
9
10 # summarize the model
11 model.summary()
```

2. MNIST 데이터 분석: 신경망 구축

- 출력층

- 10개 노드
- 활성화함수 – softmax

```
1 # define the model
2 # input_shape = (28, 28)
3 model = tf.keras.models.Sequential([
4     tf.keras.layers.Flatten(input_shape=X_train.shape[1:]),
5     tf.keras.layers.Dense(512, activation='relu'),
6     tf.keras.layers.Dropout(0.2),
7     tf.keras.layers.Dense(10, activation='softmax')
8 ])
9
10 # summarize the model
11 model.summary()
```

2. MNIST 데이터 분석: 신경망 구축

- **model.summary()**

```
1 # define the model
2 # input_shape = (28, 28)
3 model = tf.keras.models.Sequential([
4     tf.keras.layers.Flatten(input_shape=X_train.shape[1:]),
5     tf.keras.layers.Dense(512, activation='relu'),
6     tf.keras.layers.Dropout(0.2),
7     tf.keras.layers.Dense(10, activation='softmax')
8 ])
9
10 # summarize the model
11 model.summary()
```



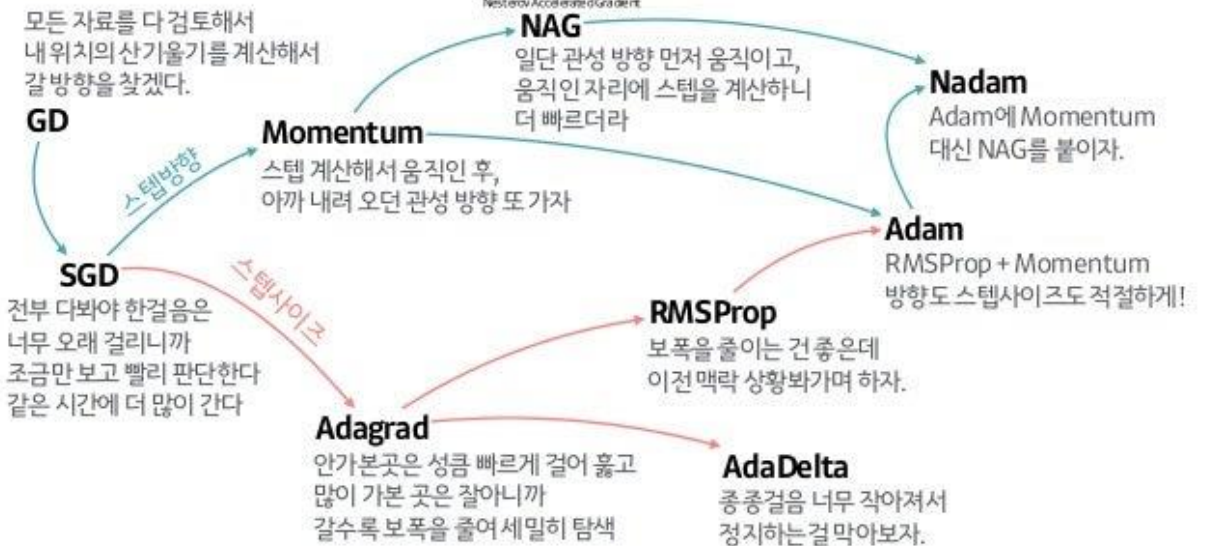
Layer (type)	Output Shape	Param #
flatten_1 (Flatten)	(None, 784)	0
dense_1 (Dense)	(None, 512)	401920
dropout_1 (Dropout)	(None, 512)	0
dense_2 (Dense)	(None, 10)	5130
Total params: 407,050		
Trainable params: 407,050		
Non-trainable params: 0		

2. MNIST 데이터 분석: 컴파일

- 손실함수(Loss function)
- 옵티마이저(Optimizer)
- 정확도(Accuracy)

```
1 model.compile(  
2     loss='categorical_crossentropy',  
3     optimizer='rmsprop',  
4     metrics=['accuracy']  
5 )
```

산 내려오는 작은 오솔길 찾기(Optimizer)의 발달 계보



출처: 하용호, 자습해도 모르겠던 딥러닝

2. MNIST 데이터 분석: 모델 학습

- **ModelCheckpoint()**
 - 가중치 저장 파일 (mnist_best.h5)

```
1  # train the model
2  → checkpointer = tf.keras.callbacks.ModelCheckpoint(
3                        filepath='mnist_best.h5',
4                        verbose=1,
5                        save_best_only=True)
6  model.fit(X_train, y_train, batch_size=128, epochs=10,
7            validation_split=0.2,
8            callbacks=[checker],
9            verbose=1, shuffle=True)
```

2. MNIST 데이터 분석: 모델 학습

- **fit()** – 학습 파라미터 정하기

```
1 # train the model
2 checkpointer = tf.keras.callbacks.ModelCheckpoint(
3     filepath='mnist_best.h5',
4     verbose=1,
5     save_best_only=True)
6 model.fit(X_train, y_train, batch_size=128, epochs=10,
7     validation_split=0.2,
8     callbacks=[checker],
9     verbose=1, shuffle=True)
```

2. MNIST 데이터 분석: 분류 정확도 측정

- 학습된 모델 가중치
(mnist_best.h5)
사용하기

```
1 model.load_weights('mnist_best.h5')
2
3 loss_and_metrics = model.evaluate(X_test, y_test)
4 accuracy = 100 * loss_and_metrics[1]
5
6 print("Test accuracy: {}".format(accuracy))
```

2. MNIST 데이터 분석: 분류 정확도 측정

- 평가 데이터 분류하기

```
1 model.load_weights('mnist_best.h5')
2
3 loss_and_metrics = model.evaluate(X_test, y_test)
4 accuracy = 100 * loss_and_metrics[1]
5
6 print("Test accuracy: {}".format(accuracy))
```

2. MNIST 데이터 분석: 분류 정확도 측정

```
1 model.load_weights('mnist_best.h5')
2
3 loss_and_metrics = model.evaluate(X_test, y_test)
4 accuracy = 100 * loss_and_metrics[1]
5
6 print("Test accuracy: {}".format(accuracy))
```

Test accuracy: 98.00999999999999%

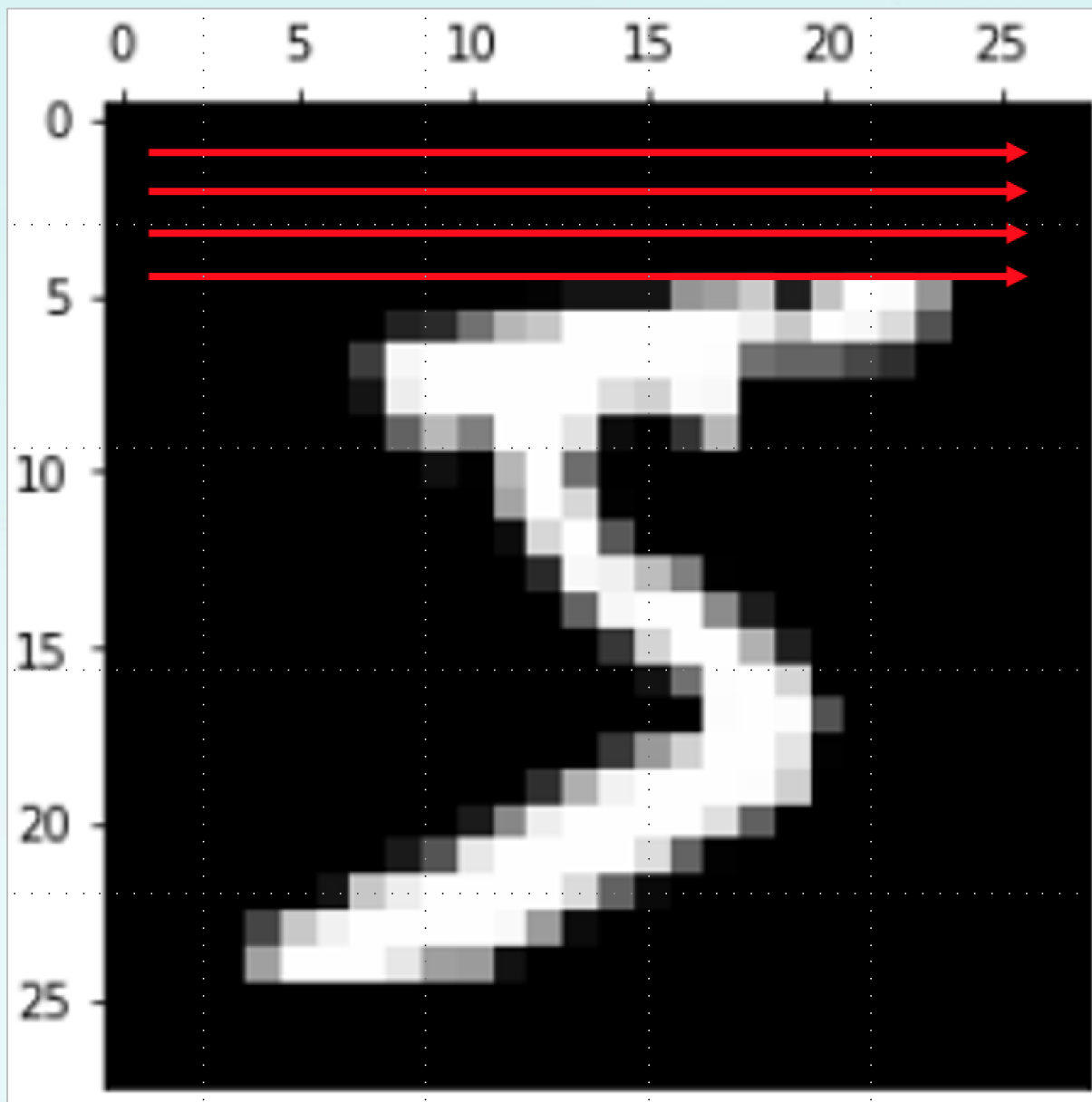
3. CNN 구현: 인공 신경망(ANN)의 한계

■ 입력층

- **X.shape = (60000, 28, 28)**
- **X.shape[1:] = (28, 28)**
- **노드(뉴론)의 수: $784 = 28 \times 28$**

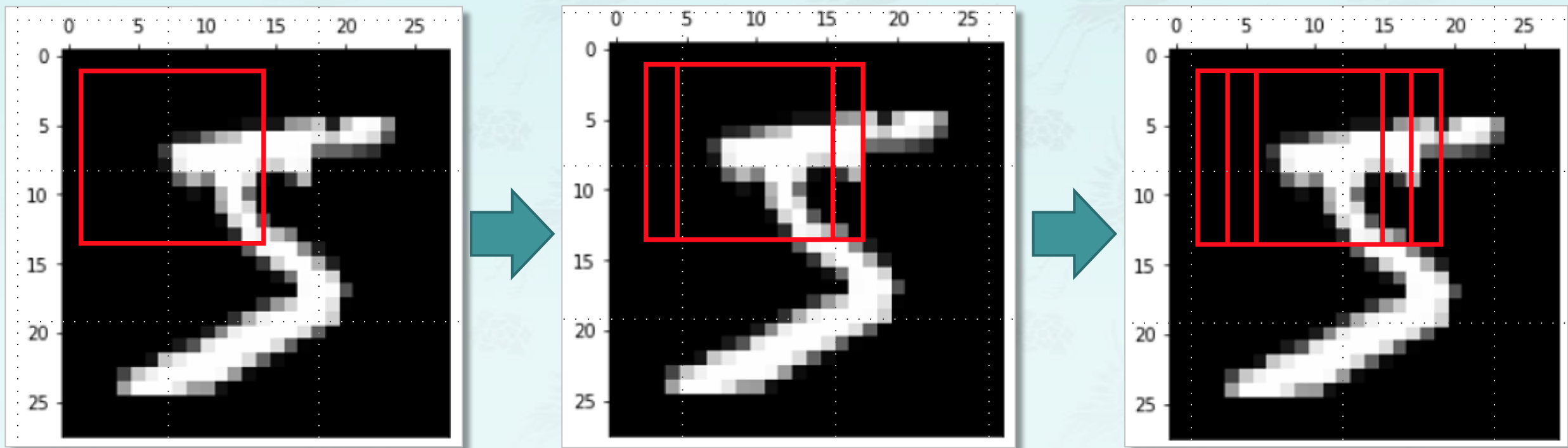
```
1 # define the model
2 # input_shape = (28, 28)
3 model = tf.keras.models.Sequential([
4     tf.keras.layers.Flatten(input_shape=X_train.shape[1:]),
5     tf.keras.layers.Dense(512, activation='relu'),
6     tf.keras.layers.Dropout(0.2),
7     tf.keras.layers.Dense(10, activation='softmax')
8 ])
9
10 # summarize the model
11 model.summary()
```

3. CNN 구현: 인공 신경망(ANN)의 한계



```
1 # define the model
2 # input_shape = (28, 28)
3 model = tf.keras.models.Sequential([
4     tf.keras.layers.Flatten(input_shape=X_train.shape[1:]),
5     tf.keras.layers.Dense(512, activation='relu'),
6     tf.keras.layers.Dropout(0.2),
7     tf.keras.layers.Dense(10, activation='softmax')
8 ])
9
10 # summarize the model
11 model.summary()
```

3. CNN 구현: 합성곱층(Convolutional Layer)



3. CNN 구현: Pooling Layer

- Max Pooling Layer

12	20	30	0
8	12	2	0
37	4	34	70
25	12	100	112

- Global Average Pooling Layer

3. CNN 구현: Pooling Layer

12	20	30	0
8	12	2	0
37	4	34	70
25	12	100	112



- Max Pooling

20	30
37	112

- Average Pooling

3. CNN 구현: Pooling Layer

12	20	30	0
8	12	2	0
37	4	34	70
25	12	100	112



■ Max Pooling

20	30
37	112



■ Average Pooling

13	8
20	79

3. CNN 구현: MNIST 데이터 분석

- kernel_size = 2
- filters = 16

```
1  # define the model
2  model = Sequential([
3      tf.keras.layers.Conv2D(
4          filters=16, kernel_size=2,
5          padding='valid', activation='relu',
6          input_shape=(28, 28, 1)),
7      tf.keras.layers.Dropout(0.2),
8      tf.keras.layers.MaxPooling2D(pool_size=2),
9      tf.keras.layers.Conv2D(
10         filters=32, kernel_size=2,
11         padding='valid', activation='relu'),
12      tf.keras.layers.Dropout(0.2),
13      tf.keras.layers.MaxPooling2D(pool_size=2),
14      tf.keras.layers.Conv2D(
15         filters=64, kernel_size=2,
16         padding='valid', activation='relu'),
17      tf.keras.layers.Dropout(0.2),
18      tf.keras.layers.MaxPooling2D(pool_size=2),
19      tf.keras.layers.Flatten(),
20      tf.keras.layers.Dense(10, activation='softmax')
21  ])
22
23  # summarize the model
24  model.summary()
```

3. CNN 구현: MNIST 데이터 분석

- MaxPooling2D()

```
1  # define the model
2  model = Sequential([
3      tf.keras.layers.Conv2D(
4          filters=16, kernel_size=2,
5          padding='valid', activation='relu',
6          input_shape=(28, 28, 1)),
7      tf.keras.layers.Dropout(0.2),
8      tf.keras.layers.MaxPooling2D(pool_size=2),
9      tf.keras.layers.Conv2D(
10         filters=32, kernel_size=2,
11         padding='valid', activation='relu'),
12     tf.keras.layers.Dropout(0.2),
13     tf.keras.layers.MaxPooling2D(pool_size=2),
14     tf.keras.layers.Conv2D(
15         filters=64, kernel_size=2,
16         padding='valid', activation='relu'),
17     tf.keras.layers.Dropout(0.2),
18     tf.keras.layers.MaxPooling2D(pool_size=2),
19     tf.keras.layers.Flatten(),
20     tf.keras.layers.Dense(10, activation='softmax')
21 ])
22
23 # summarize the model
24 model.summary()
```

3. CNN 구현: MNIST 데이터 분석

```
1 model.compile(  
2     loss='categorical_crossentropy',  
3     optimizer='rmsprop',  
4     metrics=['accuracy']  
5 )
```

```
1 # define the model  
2 model = Sequential([  
3     tf.keras.layers.Conv2D(  
4         filters=16, kernel_size=2,  
5         padding='valid', activation='relu',  
6         input_shape=(28, 28, 1)),  
7     tf.keras.layers.Dropout(0.2),  
8     tf.keras.layers.MaxPooling2D(pool_size=2),  
9     tf.keras.layers.Conv2D(  
10        filters=32, kernel_size=2,  
11        padding='valid', activation='relu'),  
12    tf.keras.layers.Dropout(0.2),  
13    tf.keras.layers.MaxPooling2D(pool_size=2),  
14    tf.keras.layers.Conv2D(  
15        filters=64, kernel_size=2,  
16        padding='valid', activation='relu'),  
17    tf.keras.layers.Dropout(0.2),  
18    tf.keras.layers.MaxPooling2D(pool_size=2),  
19    tf.keras.layers.Flatten(),  
20    tf.keras.layers.Dense(10, activation='softmax')  
21 ])  
22  
23 # summarize the model  
24 model.summary()
```

3. CNN 구현: MNIST 데이터 분석

```
1 model.compile(  
2  
3     from keras.callbacks import ModelCheckpoint  
4     checkpointer = ModelCheckpoint(  
5         filepath='mnist_best_cnn.h5',  
6         verbose=1,  
7         save_best_only=True)  
8     model.fit(X_train, y_train,  
9         batch_size=128, epochs=10,  
10        validation_split=0.2,  
11        callbacks=[checker],  
12        verbose=1, shuffle=True)
```

```
1 # define the model  
2 model = Sequential([  
3     tf.keras.layers.Conv2D(  
4         filters=16, kernel_size=2,  
5         padding='valid', activation='relu',  
6         input_shape=(28, 28, 1)),  
7     tf.keras.layers.Dropout(0.2),  
8     tf.keras.layers.MaxPooling2D(pool_size=2),  
9     tf.keras.layers.Conv2D(  
10        filters=32, kernel_size=2,  
11        padding='valid', activation='relu'),  
12    tf.keras.layers.Dropout(0.2),  
13    tf.keras.layers.MaxPooling2D(pool_size=2),  
14    tf.keras.layers.Conv2D(  
15        filters=64, kernel_size=2,  
16        padding='valid', activation='relu'),  
17    tf.keras.layers.Dropout(0.2),  
18    tf.keras.layers.MaxPooling2D(pool_size=2),  
19    tf.keras.layers.Flatten(),  
20    tf.keras.layers.Dense(10, activation='softmax')  
21 ])  
22  
23 # summarize the model  
24 model.summary()
```

3. CNN 구현: MNIST 데이터 분석

```
1 model.compile(
2     loss=loss_categorical_crossentropy,
3     metrics=['accuracy'],
4     optimizer=optimizer_adam)
5
6 from keras.callbacks import ModelCheckpoint
7
8 checkpointer = ModelCheckpoint(
9     filepath='mnist_best_cnn.h5',
10    verbose=1,
11    save_best_only=True)
12
13 model.fit(X_train, y_train,
14         batch_size=128, epochs=10,
15         callbacks=[checkpointer])
16
17 model.load_weights('mnist_best_cnn.h5')
18
19 loss_and_metrics = model.evaluate(X_test, y_test)
20 accuracy = 100 * loss_and_metrics[1]
21
22 print("Test accuracy: {}".format(accuracy))
```

```
1 # define the model
2 model = Sequential([
3     tf.keras.layers.Conv2D(
4         filters=16, kernel_size=2,
5         padding='valid', activation='relu',
6         input_shape=(28, 28, 1)),
7     tf.keras.layers.Dropout(0.2),
8     tf.keras.layers.MaxPooling2D(pool_size=2),
9     tf.keras.layers.Conv2D(
10        filters=32, kernel_size=2,
11        padding='valid', activation='relu'),
12    tf.keras.layers.Dropout(0.2),
13    tf.keras.layers.MaxPooling2D(pool_size=2),
14    tf.keras.layers.Conv2D(
15        filters=64, kernel_size=2,
16        padding='valid', activation='relu'),
17    tf.keras.layers.Dropout(0.2),
18    tf.keras.layers.MaxPooling2D(pool_size=2),
19    tf.keras.layers.Flatten(),
20    tf.keras.layers.Dense(10, activation='softmax')
21 ])
22
23 # summarize the model
24 model.summary()
```


3. CNN 구현: MNIST 데이터 분석

```
1 model.compile(
2     loss=loss_function,
3     optimizer=optimizer,
4     metrics=['accuracy'])
5
6 from keras.callbacks import ModelCheckpoint
7
8 checkpointer = ModelCheckpoint(
9     filepath='mnist_best_cnn.h5',
10    verbose=1,
11    save_best_only=True)
12
13 model.fit(X_train, y_train,
14         batch_size=128, epochs=10,
15         validation_data=(X_test, y_test),
16         callbacks=[checker])
17
18 model.load_weights('mnist_best_cnn.h5')
19
20 loss_and_metrics = model.evaluate(X_test, y_test)
21 accuracy = 100 * loss_and_metrics[1]
22
23 print("Test accuracy: {}".format(accuracy))
```

Test accuracy: 98.79%

```
1 # define the model
2 model = Sequential([
3     tf.keras.layers.Conv2D(
4         filters=16, kernel_size=2,
5         padding='valid', activation='relu',
6         input_shape=(28, 28, 1)),
7     tf.keras.layers.Dropout(0.2),
8     tf.keras.layers.MaxPooling2D(pool_size=2),
9     tf.keras.layers.Conv2D(
10        filters=32, kernel_size=2,
11        padding='valid', activation='relu'),
12    tf.keras.layers.Dropout(0.2),
13    tf.keras.layers.MaxPooling2D(pool_size=2),
14    tf.keras.layers.Conv2D(
15        filters=64, kernel_size=2,
16        padding='valid', activation='relu'),
17    tf.keras.layers.Dropout(0.2),
18    tf.keras.layers.MaxPooling2D(pool_size=2),
19    tf.keras.layers.Flatten(),
20    tf.keras.layers.Dense(10, activation='softmax')
21 ])
22
23 # summarize the model
24 model.summary()
```

오픈 프레임워크 – 텐서플로우 & 케라스

- 학습 내용
 - 신경망 구현을 위한 오픈 프레임워크는 무엇이 있는지 알아보기.
 - **Tensorflow & Keras**는 무엇인지 이해하기.
 - **Tensorflow & Keras** 를 이용하여 **MNIST** 데이터를 분석하기.
 - **Tensorflow & Keras** 를 이용하여 **CNN**을 구현하기
- 차시 예고
 - 오픈 프레임워크 – **PyTorch**
 - 기계학습 모델 **YOLO**
 - 기계학습 모델 **GAN**

ML Open Framework: Tensorflow & Keras

파이썬으로 배우는 기계학습

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