머신러닝 기반 이미지 분류

붓꽃 분류하기

머신러닝 기반 이미지 분류

[1] 머신러닝을 이용한 이미지 분류 실습

- 아이리스 꽃 학습 데이터 세트 준비
- 데이터 전처리 및 피쳐 분석
- 머신러닝 모델 적용 및 실행
- 머신러닝 모델 별 성능 분석

[2] 딥러닝을 이용한 이미지 분류 실습

- MNIST 학습 데이터 세트 준비
- MNIST 데이터 피쳐 분석
- 딥러닝 모델(CNN) 적용 및 실행
- CNN 모델 구조 별 성능 분석

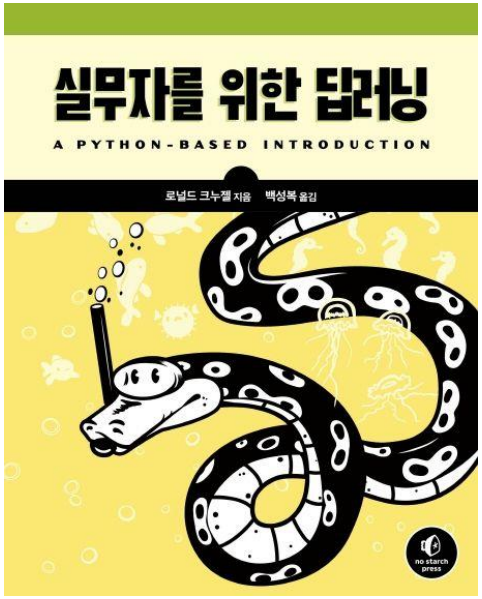
참고 자료

[1] 참고 교재: "실무자를 위한 딥러닝"

- 5장: 데이터 세트 만들기
- 7장: 고전 모델 실습
- 13장: 케라스와 MNIST를 활용한 CNN 분석

<Original Text>

- Practical Deep Learning: A Python-Based Introduction



[2] 소스 코드 및 데이터 세트

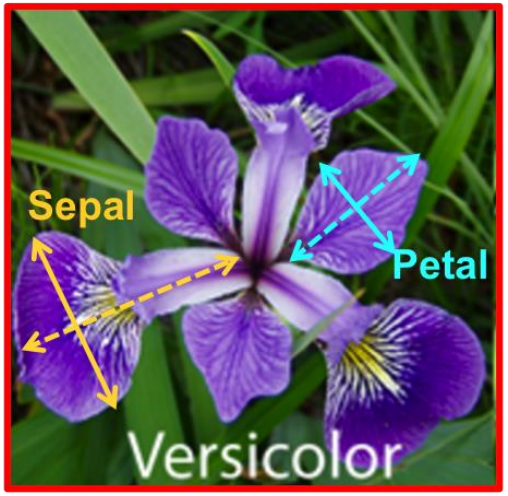
- 소스

https://github.com/rkneusel9/PracticalDeepLearningPython/

- 데이터

https://archive.ics.uci.edu/dataset/53/iris

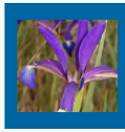
Iris 꽃 종류



- Setosa: 부채 붓꽃
- Versicolour: 베르시 붓꽃
- Virginica: 버지니카 붓꽃

Iris 꽃 데이터
UC Irvine ML Repository

https://archive.ics.uci.edu/dataset/53/iris



Iris
Donated on 6/30/1988

A small classic dataset from Fisher, 1936. One of the earliest known datasets used for evaluating classification methods.

Dataset Characteristics
Tabular

Subject Area
Biology

Associated Tasks
Classification

Feature Type
Real

 # Instances
150

 # Features
4

DOWNLOAD

IMPORT IN PYTHON

CITE

352 citations

540551 views

Keywords

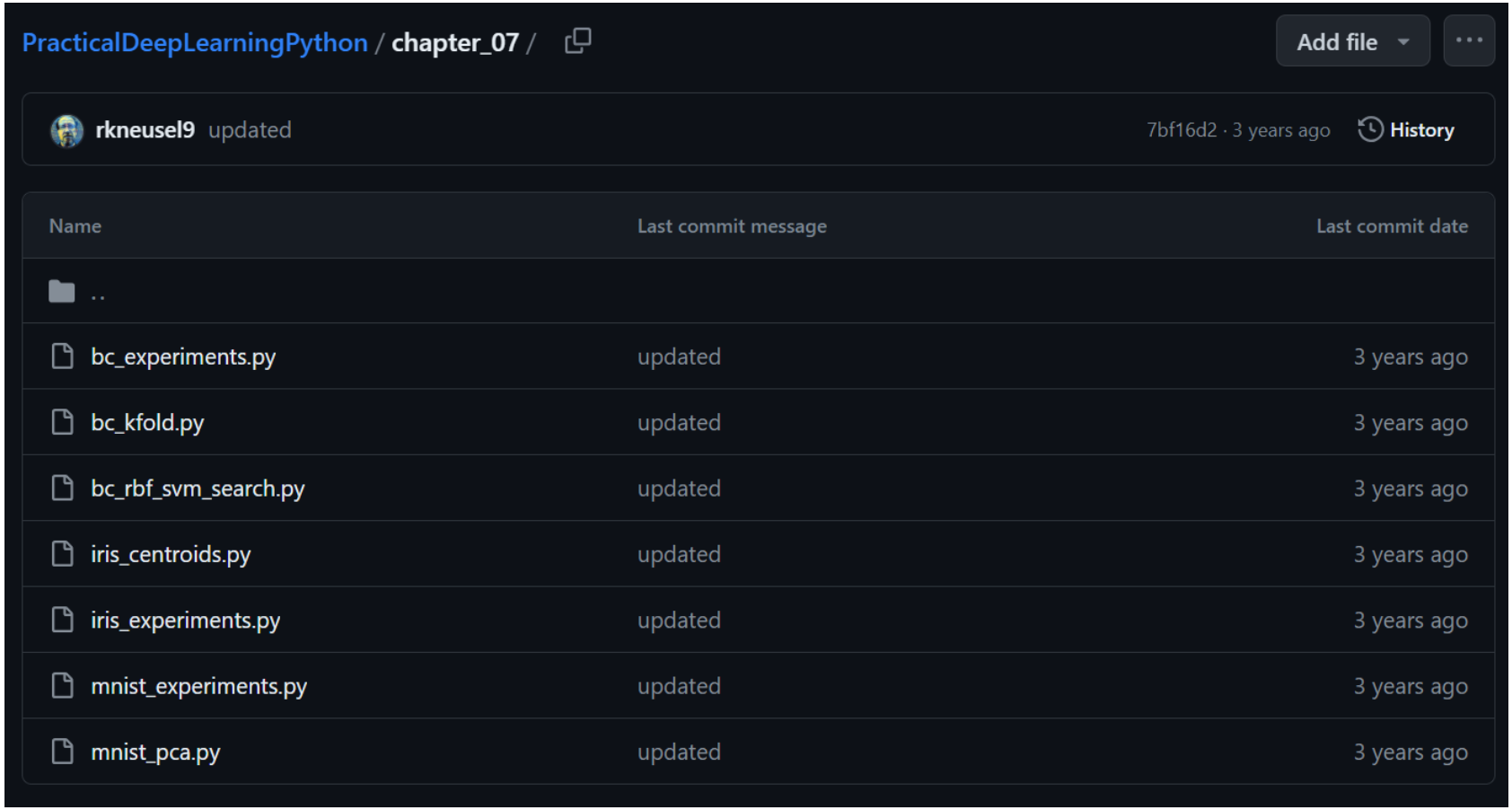
ecology

Variables Table

Variable Name	Role	Type	Demographic	Description	Units	Missing Values
sepal length	Feature	Continuous			cm	no
sepal width	Feature	Continuous			cm	no
petal length	Feature	Continuous			cm	no
petal width	Feature	Continuous			cm	no
class	Target	Categorical		class of iris plant: Iris Setosa, Iris Versicolour, or Iris Virginica		no

Iris 꽃 분류 소스 코드

https://github.com/rkneusel9/PracticalDeepLearningPython/tree/main/chapter_07

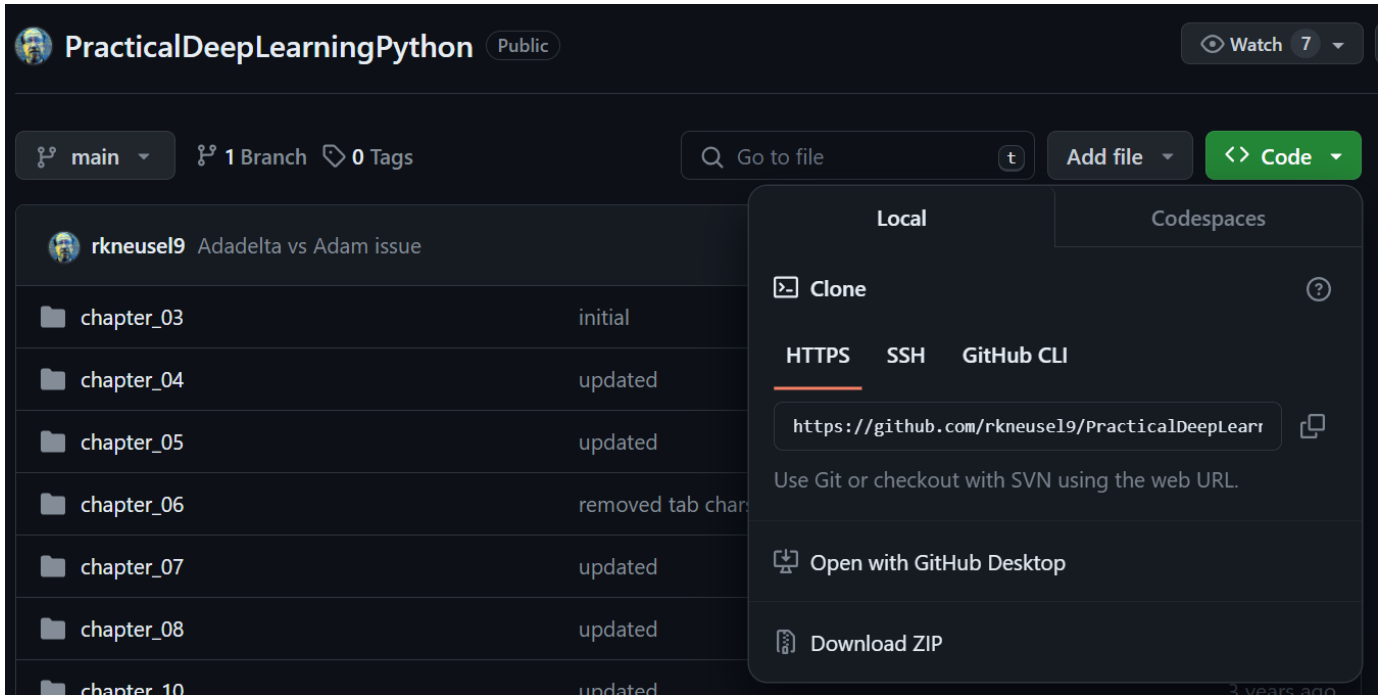


소스 코드 준비

% git clone https://github.com/rkneusel9/PracticalDeepLearningPython

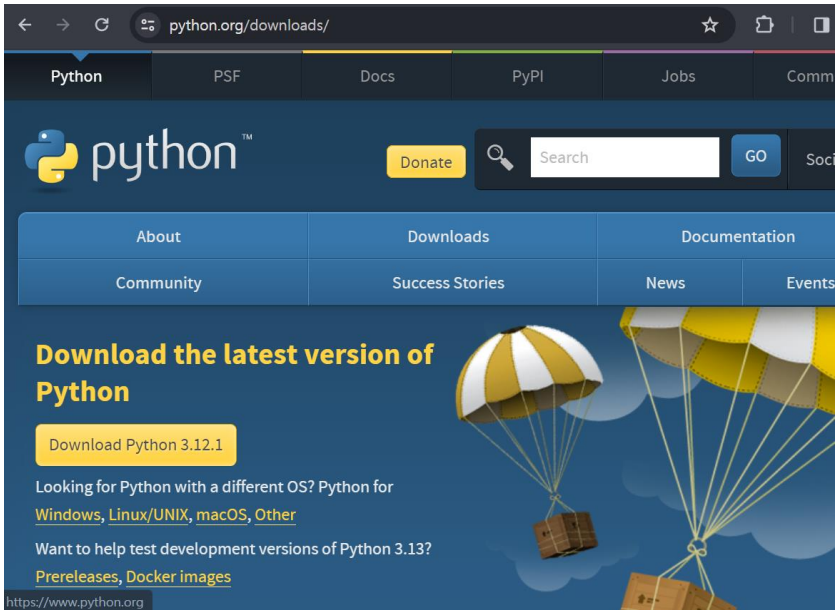
```
sbbaik@SEONGBOK-THINKPAD MINGW64 /d/ttt
$ git clone https://github.com/rkneusel9/PracticalDeepLearningPython
Cloning into 'PracticalDeepLearningPython'...
remote: Enumerating objects: 310, done.
remote: Counting objects: 100% (43/43), done.
remote: Compressing objects: 100% (17/17), done.
remote: Total 310 (delta 31), reused 35 (delta 26), pack-reused 267
Receiving objects: 100% (310/310), 416.71 KiB | 10.16 MiB/s, done.
Resolving deltas: 100% (171/171), done.
```

또는, zip 파일 다운로드

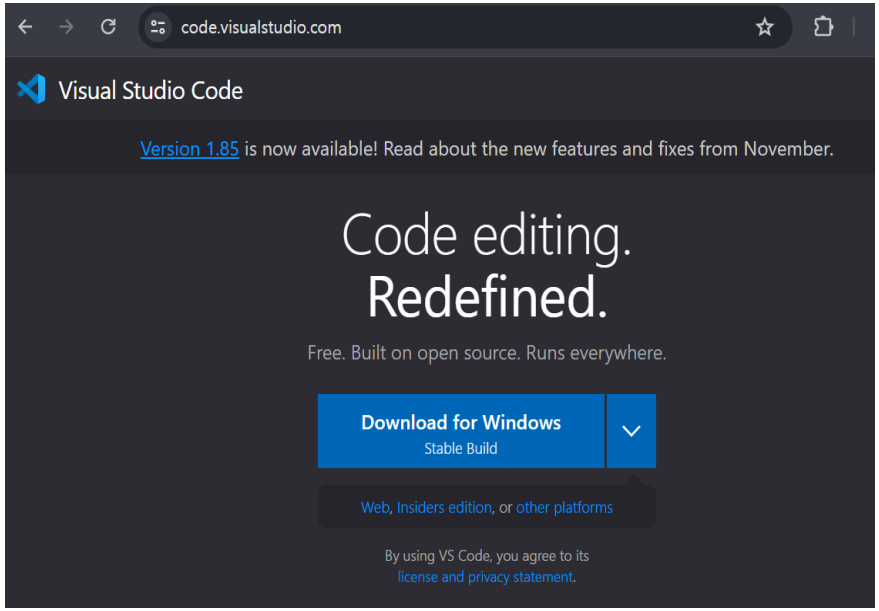


실행 환경 구성

- 파이썬 설치 ==> python.org/downloads



- 개발 환경 설치 ==> vs code



- 파이썬 패키지 설치

- pip install numpy # 넘파이
- pip install scikit-learn # 싸이킷-런 패키지

- 또는, 다른 개발 환경 ==> 파이참(PyCharm)

Iris 꽃 데이터 로딩
load_iris.py

```
with open("data/iris/iris.data") as f:
    lines = [i[:-1] for i in f.readlines()]

n = ["Iris-setosa","Iris-versicolor","Iris-virginica"]
x = [n.index(i.split(",")[-1]) for i in lines if i != ""]
x = np.array(x, dtype="uint8")

y = [[float(j) for j in i.split(",")[:-1]] for i in lines if i != ""]
y = np.array(y)

i = np.argsort(np.random.random(x.shape[0]))
x = x[i]
y = y[i]

np.save("data/iris/iris_features.npy", y)
np.save("data/iris/iris_labels.npy", x)
```

'₩n' 문자 지우기

붓꽃 종류별 레이블 0,1,2로 달아주기

붓꽃 데이터 값 2차원 배열로 저장 (레코드, 4가지 특징값)

무작위로 섞어 주기

Iris 꽃 분류 학습
iris_experiments.py

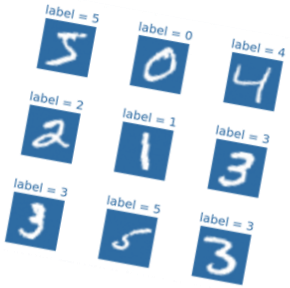
```
...
from sklearn.neighbors import NearestCentroid
...

def run(x_train, y_train, x_test, y_test, clf):
    clf.fit(x_train, y_train)
    print("     predictions  :", clf.predict(x_test))
    print("     actual labels:", y_test)
    print("     score = %0.4f" % clf.score(x_test, y_test))

def main():
    x = np.load("../data/iris/iris_features.npy")
    y = np.load("../data/iris/iris_labels.npy")
    N = 120
    x_train = x[:N]; x_test = x[N:]
    y_train = y[:N]; y_test = y[N:]
    xa_train=np.load
("../data/iris/iris_train_features_augmented.npy")
    ya_train=np.load
("../data/iris/iris_train_labels_augmented.npy")
    xa_test =np.load
("../data/iris/iris_test_features_augmented.npy")
    ya_test =np.load
("../data/iris/iris_test_labels_augmented.npy")
    print("Nearest centroid:")
    run(x_train, y_train, x_test, y_test, NearestCentroid())
    print("k-NN classifier (k=3):")
    run(x_train, y_train, x_test, y_test,
KNeighborsClassifier(n_neighbors=3))
    print("Naive Bayes classifier (Gaussian):")
    run(x_train, y_train, x_test, y_test, GaussianNB())
    print("Naive Bayes classifier (Multinomial):")
    run(x_train, y_train, x_test, y_test, MultinomialNB())
    print("Decision Tree classifier:")
    run(x_train, y_train, x_test, y_test, DecisionTreeClassifier())
    print("Random Forest classifier (estimators=5):")
    run(xa_train, ya_train, xa_test, ya_test,
RandomForestClassifier(n_estimators=5))

    print("SVM (linear, C=1.0):")
    run(xa_train, ya_train, xa_test, ya_test, SVC(kernel="linear",
C=1.0))
    print("SVM (RBF, C=1.0, gamma=0.25):")
    run(xa_train, ya_train, xa_test, ya_test, SVC(kernel="rbf",
C=1.0, gamma=0.25))
    print("SVM (RBF, C=1.0, gamma=0.001, augmented)")
    run(xa_train, ya_train, xa_test, ya_test, SVC(kernel="rbf",
C=1.0, gamma=0.001))
    print("SVM (RBF, C=1.0, gamma=0.001, original)")
    run(x_train, y_train, x_test, y_test, SVC(kernel="rbf", C=1.0,
gamma=0.001))
```

딥러닝 기반 이미지 분류



손글씨(MNIST) 분류하기

머신러닝 기반 이미지 분류

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[2] 딥러닝을 이용한 이미지 분류 실습

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- CNN 모델 구조 별 성능 분석

MNIST 데이터 세트

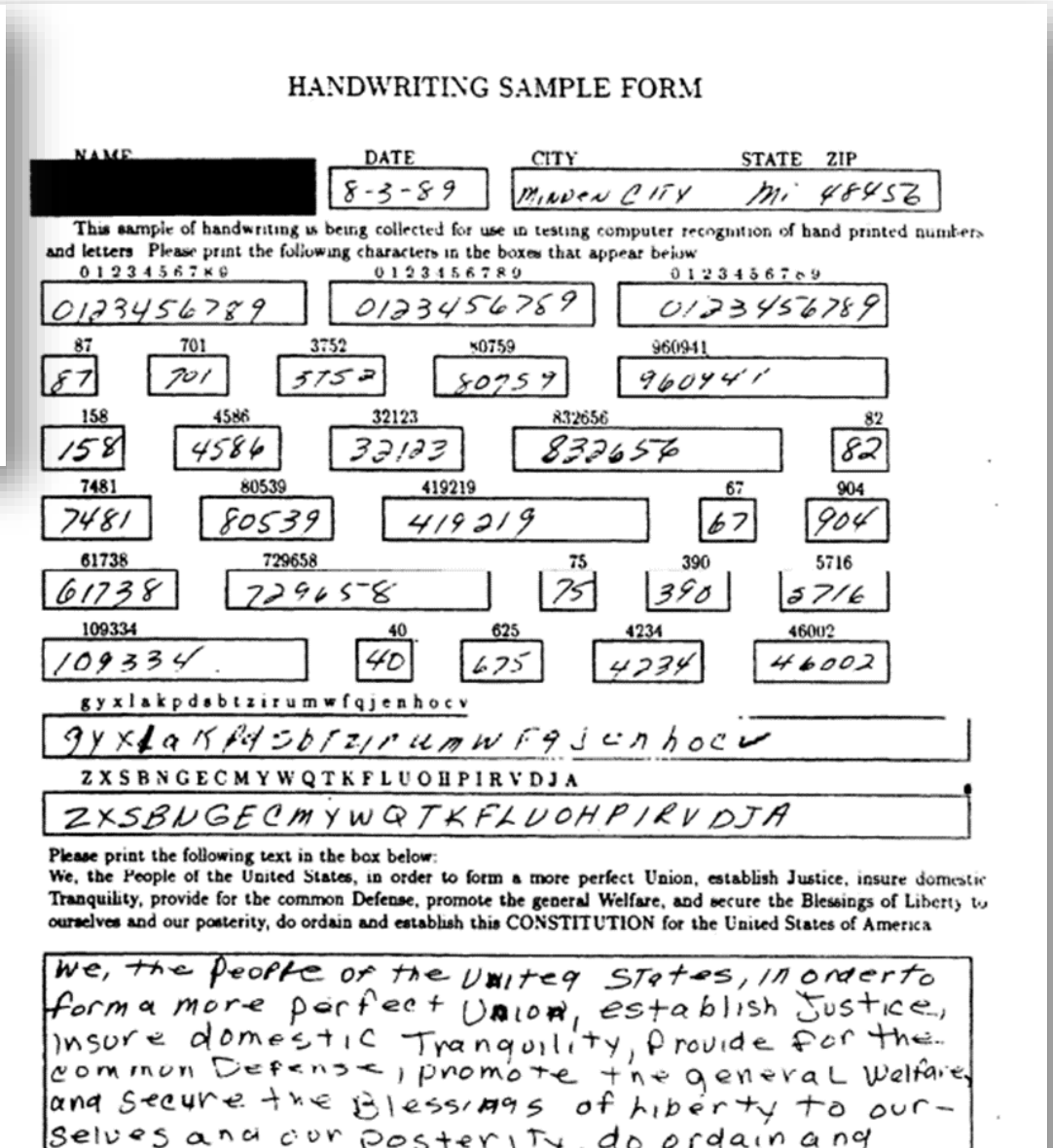
NIST Special Database 19
Handprinted Forms and Characters Database

Patrick J Grother
Visual Image Processing Group
Advanced Systems Division
National Institute of Standards and Technology

patrick@magi.ncsl.nist.gov

March 16, 1995

Modified NIST
데이터베이스



HANDWRITING SAMPLE FORM

NAME [redacted] DATE 8-3-89 CITY Minden City STATE MI ZIP 48456

This sample of handwriting is being collected for use in testing computer recognition of hand printed numbers and letters. Please print the following characters in the boxes that appear below.

0 1 2 3 4 5 6 7 8 9 | 0 1 2 3 4 5 6 7 8 9 | 0 1 2 3 4 5 6 7 8 9

0123456789 | 0123456789 | 0123456789

87 | 701 | 3752 | 80759 | 960941

158 | 4586 | 32123 | 832656 | 82

7481 | 80539 | 419219 | 67 | 904

61738 | 729658 | 75 | 390 | 5716

109334 | 40 | 625 | 4234 | 46002

g y x l a k p d s b t z i r u m w f q j e n h o c v

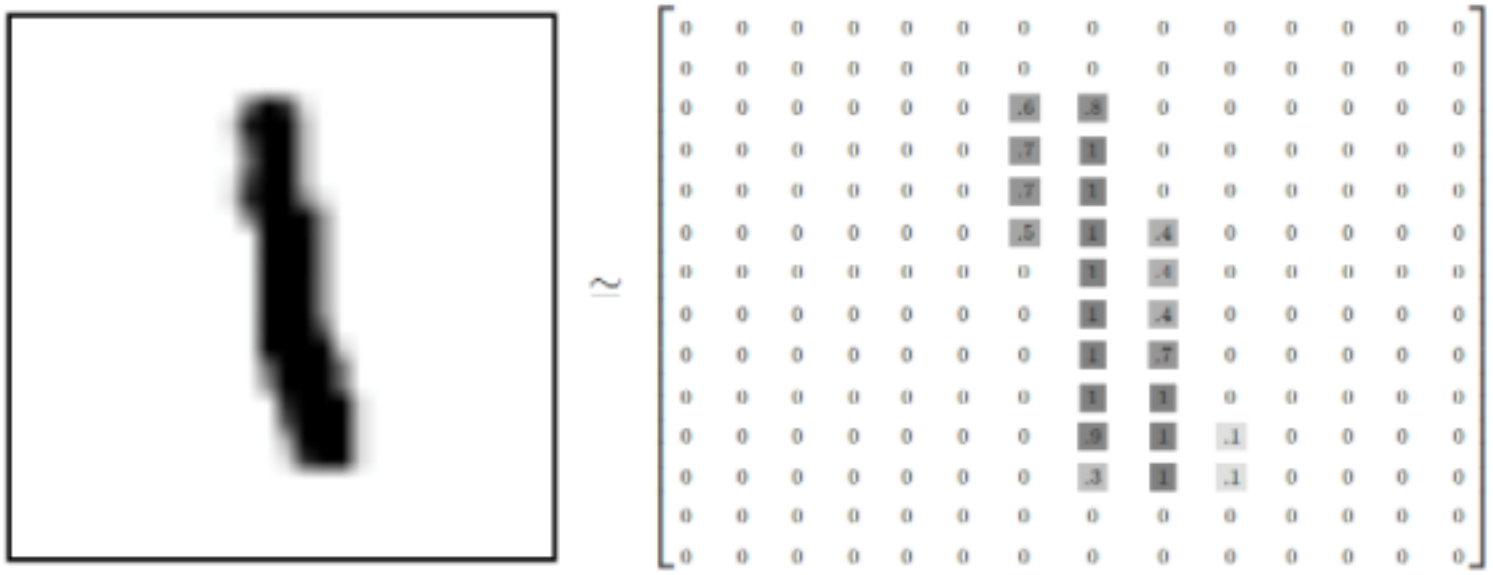
ZXSBNGECMYWQTKFLUOHPIRVDJA

Please print the following text in the box below:
We, the People of the United States, in order to form a more perfect Union, establish Justice, insure domestic Tranquility, provide for the common Defense, promote the general Welfare, and secure the Blessings of Liberty to ourselves and our posterity, do ordain and establish this CONSTITUTION for the United States of America

We, the People of the United States, in order to form a more perfect Union, establish Justice, insure domestic Tranquility, provide for the common Defense, promote the general Welfare, and secure the Blessings of liberty to ourselves and our Posterity, do ordain and

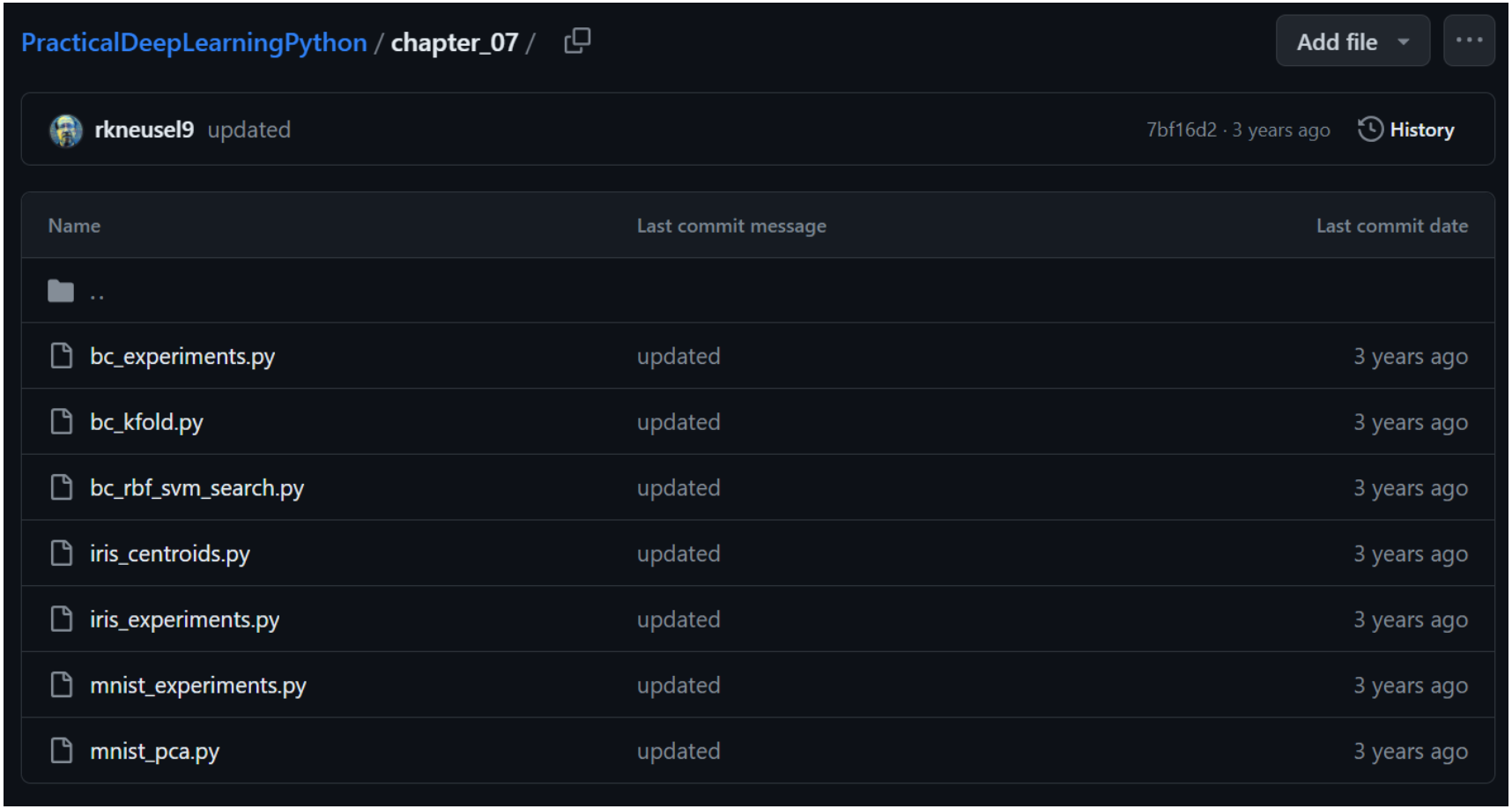
MNIST 데이터 세트

- 딥러닝 분야의 Hello, World
- 60,000개의 training set / 10,000개의 test set
- 28x28 흑백 이미지



ML기반 MNIST 소스 코드

https://github.com/rkneusel9/PracticalDeepLearningPython/tree/main/chapter_07

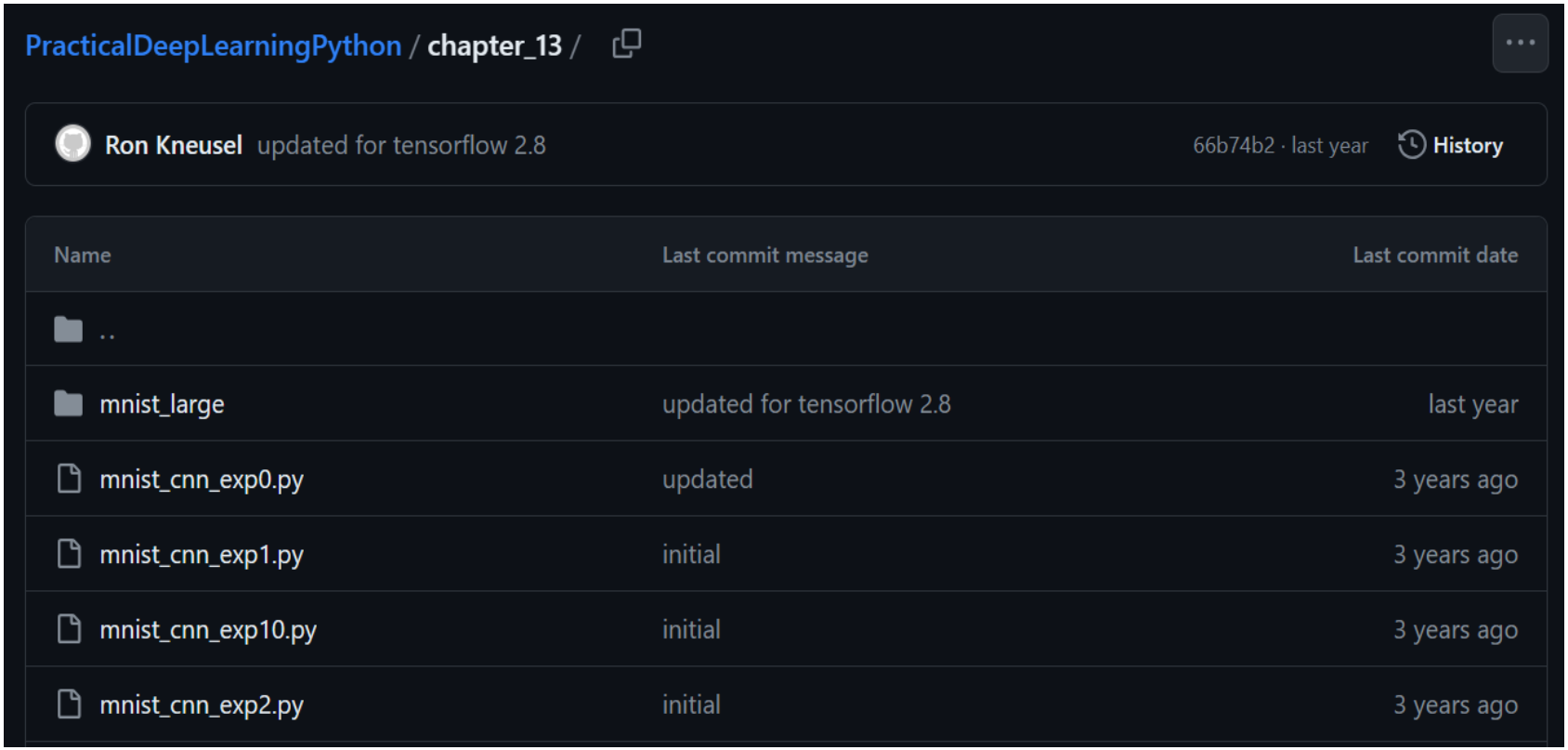


ML기반 MNIST 분류 성능

- LinearSVM (C=10.0) : score = 0.8784 (time, train= 880.605, test= 0.035)
- Nearest centroid : score = 0.7523 (time, train= 0.024, test= 0.005)
- k-NN classifier (k=3) : score = 0.9360 (time, train= 0.294, test= 4.541)
- k-NN classifier (k=7) : score = 0.9372 (time, train= 0.170, test= 5.206)
- Naive Bayes (Gaussian) : score = 0.7999 (time, train= 0.032, test= 0.025)
- Decision Tree : score = 0.8422 (time, train= 5.981, test= 0.006)
- Random Forest (trees= 5) : score = 0.8816 (time, train= 4.082, test= 0.016)
- Random Forest (trees= 50) : score = 0.9252 (time, train= 41.157, test= 0.138)
- Random Forest (trees=500) : score = 0.9270 (time, train= 472.070, test= 1.372)
- Random Forest (trees=1000): score = 0.9269 (time, train= 820.068, test= 2.689)

DL기반 MNIST 소스 코드

https://github.com/rkneusel9/PracticalDeepLearningPython/tree/main/chapter_13



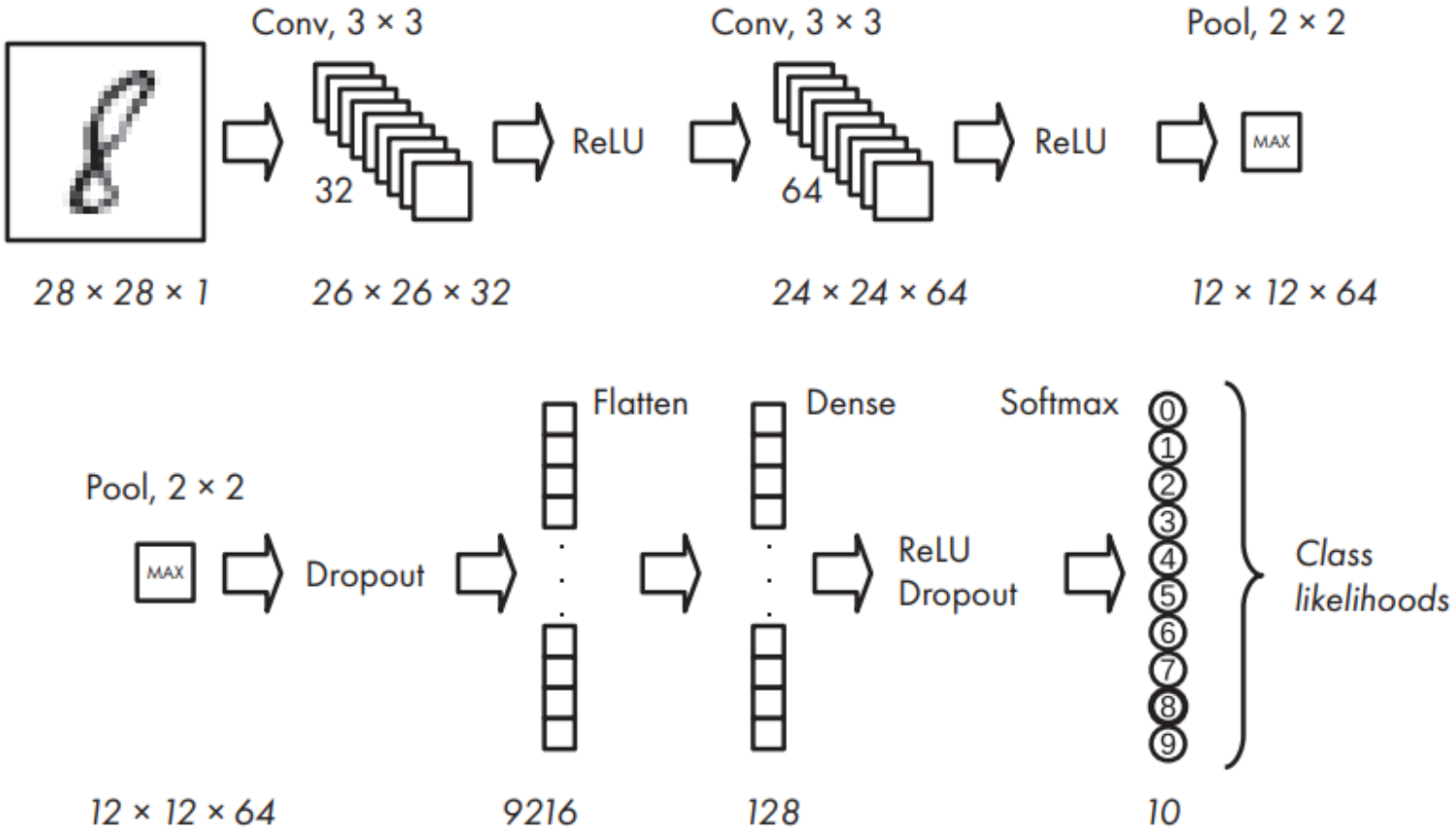
mnist_cnn_exp0.py

```
import keras
from keras.datasets import mnist
from keras.models import Sequential
from keras.layers import Dense, Dropout, Flatten
from keras.layers import Conv2D, MaxPooling2D
from keras import backend as K

(x_train, y_train), (x_test, y_test) = mnist.load_data()

x_train = x_train.astype('float32')
x_test = x_test.astype('float32')
x_train /= 255
x_test /= 255

y_train = keras.utils.to_categorical(y_train, num_classes)
y_test = keras.utils.to_categorical(y_test, num_classes)
```

```
model = Sequential()
model.add(Conv2D(32, kernel_size=(3, 3),
                 activation='relu',
                 input_shape=input_shape))
model.add(Conv2D(64, (3, 3), activation='relu'))
model.add(MaxPooling2D(pool_size=(2, 2)))
model.add(Dropout(0.25))
model.add(Flatten())
model.add(Dense(128, activation='relu'))
model.add(Dropout(0.5))
model.add(Dense(num_classes, activation='softmax'))

model.compile(loss=keras.losses.categorical_crossentropy,
              optimizer=keras.optimizers.Adadelta(),
              metrics=['accuracy'])

history = model.fit(x_train, y_train,
          batch_size=batch_size,
          epochs=epochs,
          verbose=1,
          validation_data=(x_test, y_test))
score = model.evaluate(x_test, y_test, verbose=0)
```

mnist_cnn_exp0.py

CNN기반 모델 구조

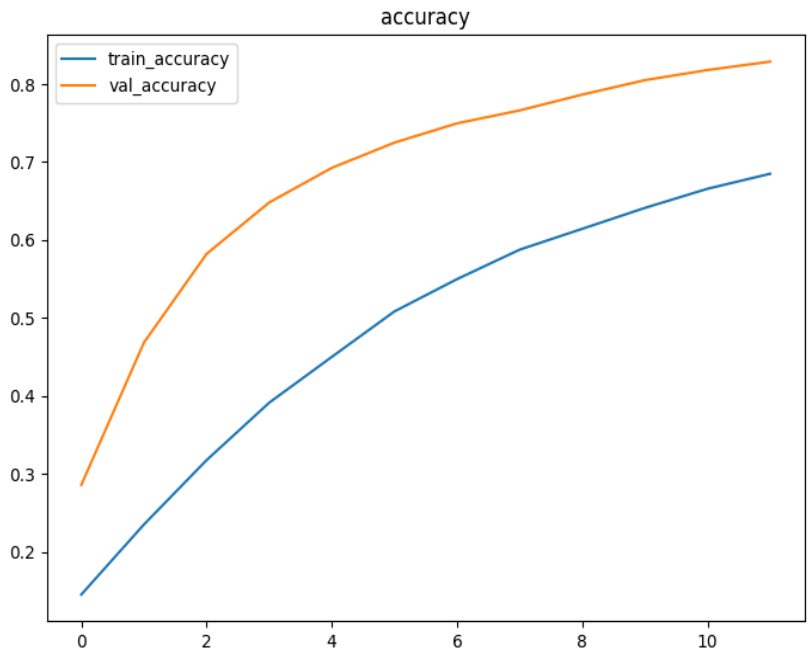


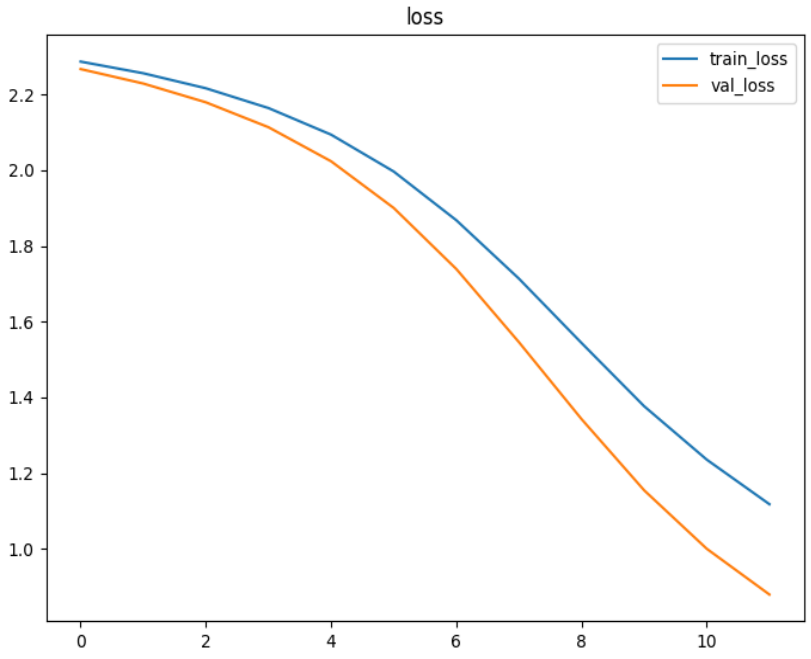
mnist_cnn_exp0.py
model summary

```
_________________________________________________________________
 Layer (type)                Output Shape              Param #
=================================================================
 conv2d (Conv2D)             (None, 26, 26, 32)        320

 conv2d_1 (Conv2D)           (None, 24, 24, 64)        18496

 max_pooling2d(MaxPooling2D) (None, 12, 12, 64)        0

 dropout (Dropout)           (None, 12, 12, 64)        0

 flatten (Flatten)           (None, 9216)              0

 dense (Dense)               (None, 128)               1179776

 dropout_1 (Dropout)         (None, 128)               0

 dense_1 (Dense)             (None, 10)                1290

=================================================================
Total params: 1199882 (4.58 MB)
Trainable params: 1199882 (4.58 MB)
Non-trainable params: 0 (0.00 Byte)
_________________________________________________________________
```

mnist_cnn_exp0.py
분류 성능

- Epoch 12/12 469/469 [==============================]
  - 4s 9ms/step - loss: 1.1179 - accuracy: 0.6851 - val_loss: 0.8793 - val_accuracy: 0.8289





전처리 추가

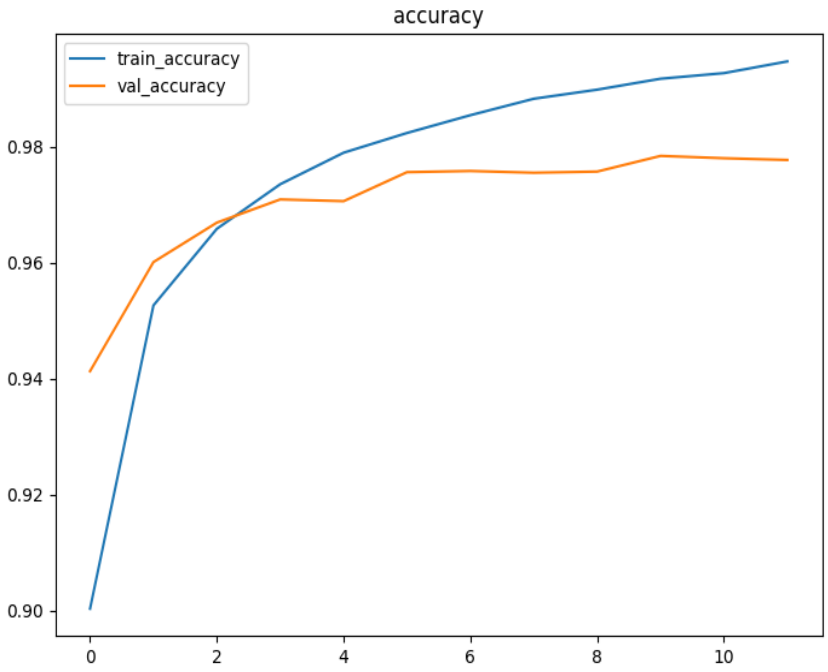
https://colab.research.google.com/github/tensorflow/datasets/blob/master/docs/keras_example.ipynb?hl=ko

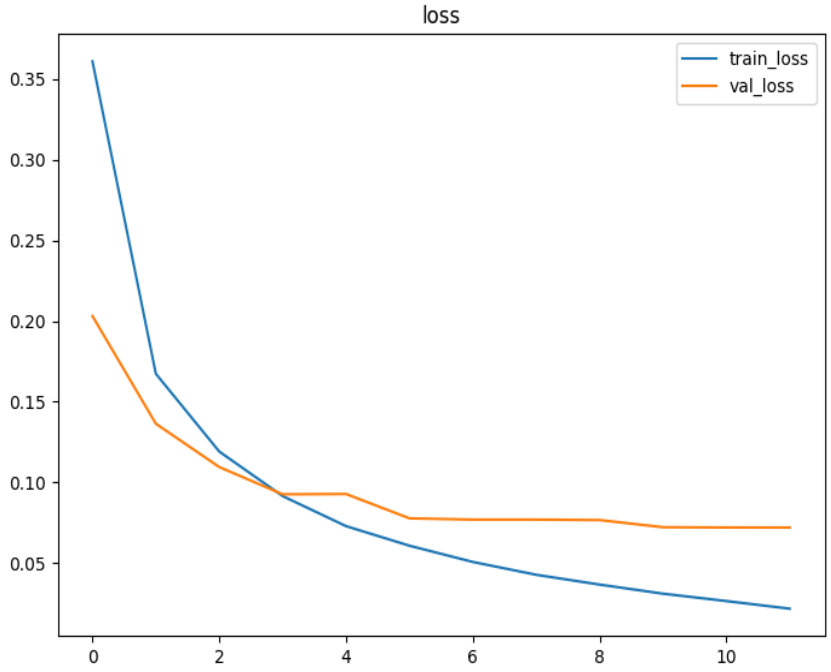
```
import tensorflow as tf
import tensorflow_datasets as tfds

(ds_train, ds_test), ds_info = tfds.load(
    'mnist',
    split=['train', 'test'],
    shuffle_files=True,
    as_supervised=True,
    with_info=True,
)

def normalize_img(image, label):
  """Normalizes images: `uint8` -> `float32`."""
  return tf.cast(image, tf.float32) / 255., label

ds_train = ds_train.map(
    normalize_img, num_parallel_calls=tf.data.AUTOTUNE)
ds_train = ds_train.cache()
ds_train =
ds_train.shuffle(ds_info.splits['train'].num_examples)
ds_train = ds_train.batch(128)
ds_train = ds_train.prefetch(tf.data.AUTOTUNE)
```

```
ds_test = ds_test.map(
    normalize_img, num_parallel_calls=tf.data.AUTOTUNE)
ds_test = ds_test.batch(128)
ds_test = ds_test.cache()
ds_test = ds_test.prefetch(tf.data.AUTOTUNE)

model = tf.keras.models.Sequential([
  tf.keras.layers.Flatten(input_shape=(28, 28)),
  tf.keras.layers.Dense(128, activation='relu'),
  tf.keras.layers.Dense(10)
])
model.compile(
    optimizer=tf.keras.optimizers.Adam(0.001),

loss=tf.keras.losses.SparseCategoricalCrossentropy(from_l
ogits=True),
    metrics=[tf.keras.metrics.SparseCategoricalAccuracy()],
)

history = model.fit(
    ds_train,
    epochs=6+6,
    validation_data=ds_test,
)
```

전처리 추가 모델 분류 성능

- Epoch 12/12 469/469 [==============================]
  - 2s 3ms/step - loss: 0.0218 - sparse_categorical_accuracy: 0.9947
  - val_loss: 0.0721 - val_sparse_categorical_accuracy: 0.9777





CNN 기반 전처리 추가 모델

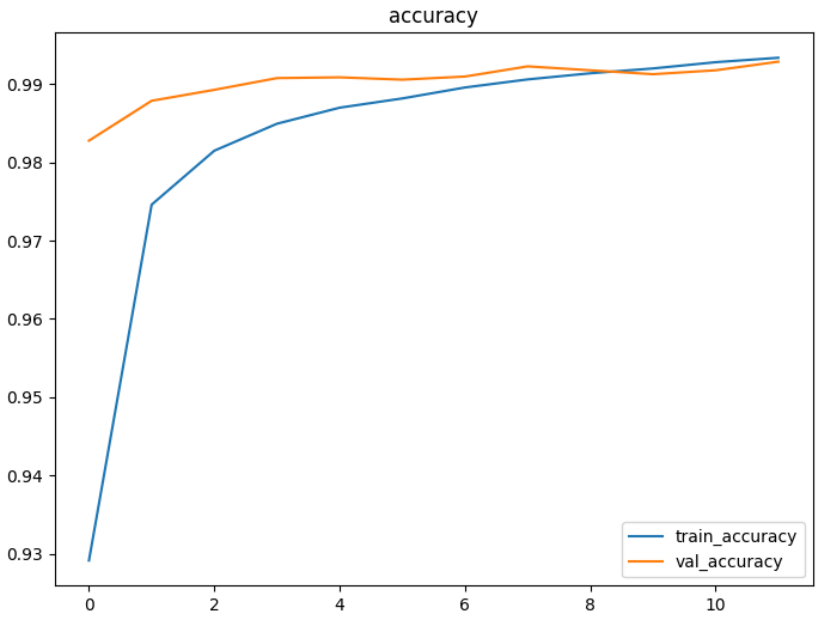
https://colab.research.google.com/github/tensorflow/datasets/blob/master/docs/keras_example.ipynb?hl=ko

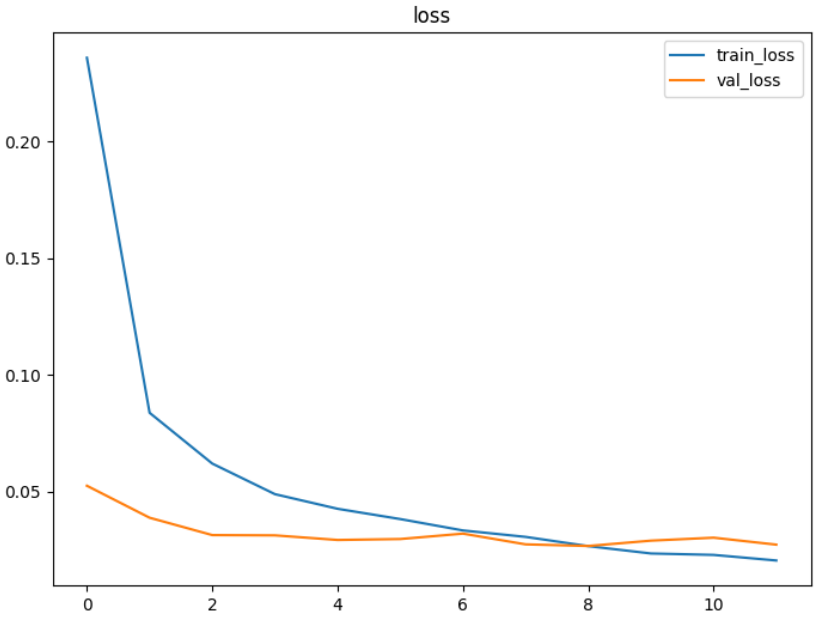
```
(ds_train, ds_test), ds_info = tfds.load(
    'mnist',
    split=['train', 'test'],
    shuffle_files=True,
    as_supervised=True,
    with_info=True,
)
def normalize_img(image, label):
    """Normalizes images: `uint8` -> `float32`."""
    return tf.cast(image, tf.float32) / 255., label
ds_train = ds_train.map(normalize_img,
num_parallel_calls=tf.data.AUTOTUNE)
ds_train = ds_train.cache()
ds_train = ds_train.shuffle(ds_info.splits['train'].num_examples)
ds_train = ds_train.batch(128)
ds_train = ds_train.prefetch(tf.data.AUTOTUNE)

ds_test = ds_test.map(normalize_img,
num_parallel_calls=tf.data.AUTOTUNE)
ds_test = ds_test.batch(128)
ds_test = ds_test.cache()
ds_test = ds_test.prefetch(tf.data.AUTOTUNE)
```

```
model = Sequential()
model.add(Conv2D(32, kernel_size=(3, 3), activation='relu',
input_shape=(28, 28, 1)))
model.add(Conv2D(64, (3, 3), activation='relu'))
model.add(MaxPooling2D(pool_size=(2, 2)))
model.add(Dropout(0.25))
model.add(Flatten())
model.add(Dense(128, activation='relu'))
model.add(Dropout(0.5))
model.add(Dense(num_classes, activation='softmax'))

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

print("Model parameters = %d" % model.count_params())
print(model.summary())

history = model.fit(
    ds_train,
    epochs=12,
    validation_data=ds_test,
)
```

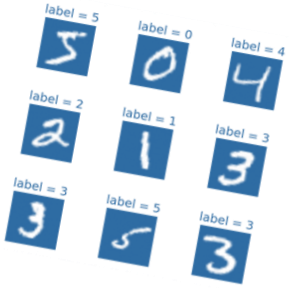
CNN 기반 전처리 추가 모델 성능

- Epoch 12/12 469/469 [==============================]
  - 5s 10ms/step - loss: 0.0204 - accuracy: 0.9934 - val_loss: 0.0272 - val_accuracy: 0.9929





Q&A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