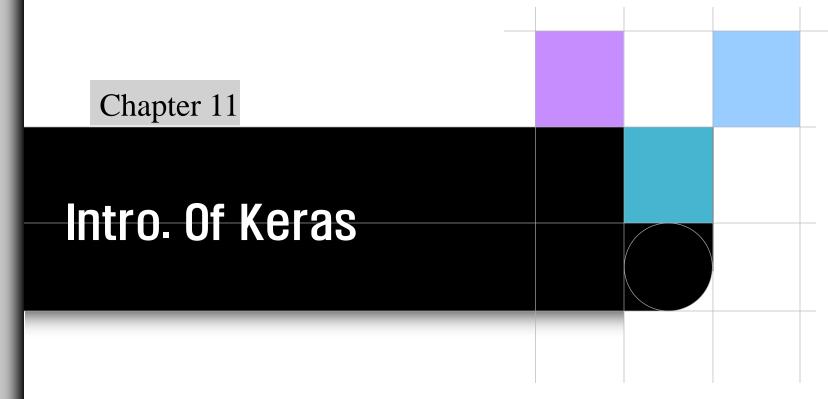
딥러닝/클라우드



오세종 MIT DANKOOK UNIVERSITY

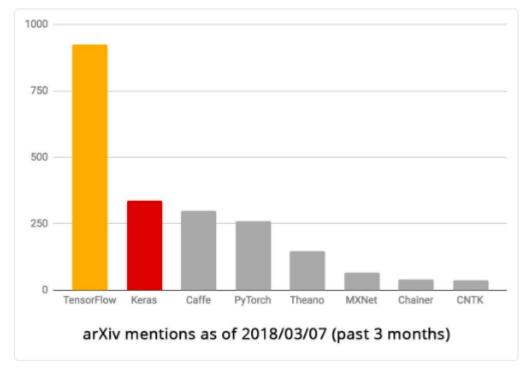
Contents

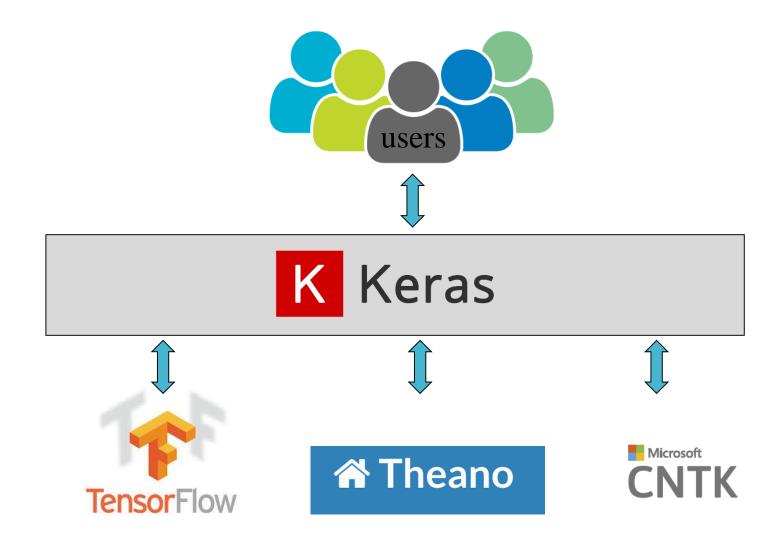
- 1. Summary of Keras
- 2. Install Keras (window)
- 3. Example of multi-layer NN
- 4. Model object
- 5. Keras functions

- 파이썬으로 구현된 간결한 딥러닝 라이브러리
- 비전문가도 쉽게 딥러닝 모델을 개발하고 활용 가능
- 직관적 API 제공
- 내부적으로는 Tensorflow, Theano, CNTK 등의 딥러닝 엔진 사용 (사용자 에게는 감추어져 있음)

• 구글 엔지니어인 프랑소와 쏠레(Francois Chollet)에 의해 개발, 유지보수

됨





- 주요 특징
 - 모듈화
 - Keras 에서 제공하는 모든 모듈은 독립적, 설정가능, 서로 연결 가능
 - 신경망층, 비용함수, 최적화, 활성함수, 정규화 기법 등이 모두 독립적 모듈로 제공
 - 이들을 조합하여 새로운 모델 구성
 - 최소주의
 - 각 모듈은 짧고 간결, 쉽게 이해 가능
 - 쉬운 확장성
 - 새로운 클래스나 함수로 모듈을 아주 쉽게 추가 가능
 - 파이썬 기반
 - 파이썬 내에서 모델의 구현 가능



https://keras.io/



https://keras.io/ko



Keras API reference

Models API

Layers API

Callbacks API

Data preprocessing

Optimizers

Metrics

Losses

Built-in small datasets

Keras Applications

Utilities

2. Keras 설치

• (2) window console 에서 다음 명령어 실행

conda install keras

C:\Windows\system32\cmd.exe

Tensorflow 를 기본 엔진으로 설치함

2. Keras 설치

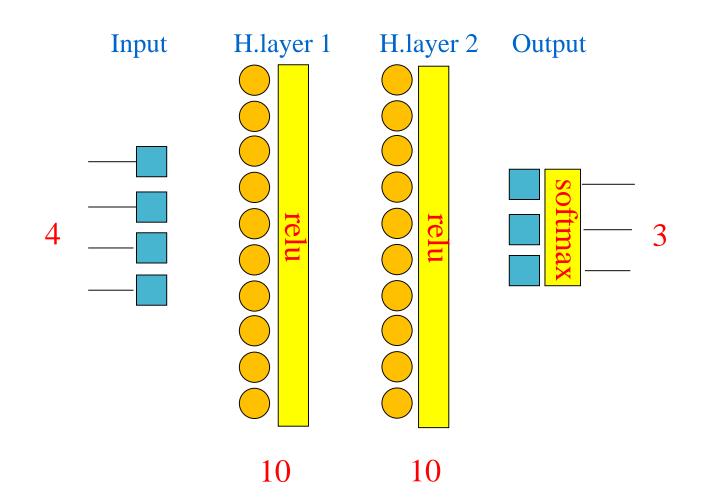
(3) Spyder 에서 keras 테스트

```
import tensorflow as tf
import keras

print('tensorflow ' + tf.__version__)
print('keras ' + keras.__version__)
```

```
Spyder (Python 3.7)
                                                                                                                          X
File Edit Search Source Run Debug Consoles Projects Tools View Help
[가 🔁 🖺 🛼 🌉 @ 🕨 🚰 🗗 🚱 🕻 세 📫 🧲 🖟 🍱 🗗 🐼 🖟 🏕 🌪 D:WOneDrive - 단국대학교₩안과데이터분석₩안지사진공부₩keras_cnn
D:\OneDrive - 단국대학교\lab세미나\xgboost\python_ml\untitled0.py
                                                                  B B C × * C O O Q Q 114%
test_pytorch,py untitled0,py* cnn_fundus_vgg16,py
      # -*- coding: utf-8 -*-
       Created on Fri Jul 3 19:23:36 2020
       @author: DKU-MANGO
       import tensorflow as tf
      import keras
      print('tensorflow ' + tf.__version__)
       print('keras ' + keras.__version__)
                                                                                        Variable explorer Plots Files
 13
                                                                  Console 1/A 🛛
                                                                                                                            In [44]: import tensorflow as tf
                                                                      ...: import keras
                                                                      ...: print('tensorflow ' + tf.__version__)
                                                                      ...: print('keras ' + keras.__version )
                                                                  tensorflow 1.14.0
                                                                  keras 2.3.1
                                                                  In [45]:
                                                                                           IPvthon console
                                                  ♠ Kite: ready (no index) ⊕ conda: base (Python 3,7,6) Line 13, Col 1 UTF-8 CRLF RW Mem 56%
```

- iris dataset 이용 품종 예측
 - Training 60%, test(validation) 40%
 - Two layer neural network



iris dataset

iris.csv

	А	В	С	D	Е
1	Sepal.Lenç	Sepal.Widt	Petal.Leng	Petal.Widtl	Species
2	5.1	3.5	1.4	0.2	setosa
3	4.9	3	1.4	0.2	setosa
4	4.7	3.2	1.3	0.2	setosa
5	4.6	3.1	1.5	0.2	setosa
6	5	3.6	1.4	0.2	setosa
7	5.4	3.9	1.7	0.4	setosa
8	4.6	3.4	1.4	0.3	setosa
9	5	3.4	1.5	0.2	setosa
10	4.4	2.9	1.4	0.2	setosa
11	4.9	3.1	1.5	0.1	setosa
12	5.4	3.7	1.5	0.2	setosa
13	4.8	3.4	1.6	0.2	setosa
14	4.8	3	1.4	0.1	setosa
15	4.3	3	1.1	0.1	setosa
16	5.8	4	1.2	0.2	setosa
17	5.7	4.4	1.5	0.4	setosa

```
# load required modules
from keras.models import Sequential
from keras.layers import Dense
from keras.utils import np_utils
from sklearn.preprocessing import LabelEncoder
from sklearn.model selection import train test split
import pandas
import matplotlib.pyplot as plt
import numpy as np
```

```
# load dataset
dataframe = pandas.read csv("D:/Rworks/iris.csv")
dataset = dataframe.values
X = dataset[:,0:4].astype(float)
Y = dataset[:,4]
# encode class values as integers
encoder = LabelEncoder()
encoder.fit(Y)
encoded Y = encoder.transform(Y)
# convert integers to dummy variables (i.e. one hot
encoded)
dummy_y = np_utils.to_categorical(encoded_Y)
# Divide train, test
train_X, test_X, train_y, test_y = train_test_split(X,
dummy_y, test_size=0.4, random state=321)
```

```
In [91]: Y
 Out[91]:
 array(['setosa', 'setosa', 'setosa', 'setosa', 'setosa',
      'setosa', 'setosa', 'versicolor', 'versicolor', 'versicolor',
      'versicolor', 'versicolor', 'versicolor', 'versicolor',
      'versicolor'. 'versicolor'. 'versicolor'. 'versicolor'.
In [92]: encoded Y
Out[92]:
1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2,
    In [93]: dummy y
 Out[93]:
 array([[1., 0., 0.],
     [1., 0., 0.],
     [1., 0., 0.],
     [1., 0., 0.],
     [1 0 0]
```

```
# define model (DNN structure)
epochs = 50
batch size = 10
model = Sequential()
model.add(Dense(10, input dim=4, activation='relu'))
model.add(Dense(10, activation='relu'))
model.add(Dense(3, activation='softmax'))
model.summary() # show model structure
# Compile model
model.compile(loss='categorical_crossentropy',
              optimizer='adam',
              metrics=['accuracy'])
```

H.layer 2 Output

H.layer 1

In [94]: model.summary()
Model: "sequential_16"

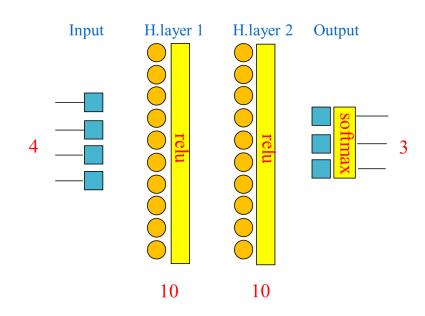
weight values (4*10) + nodes(10)

Layer (type)	Output Shape	Param #
dense_32 (Dense)	(None, 10)	50
dense_33 (Dense)	(None, 10)	110
dense_34 (Dense)	(None, 3)	33

H.Layer 1

H.Layer 2
Output Layer

Total params: 193 Trainable params: 193 Non-trainable params: 0



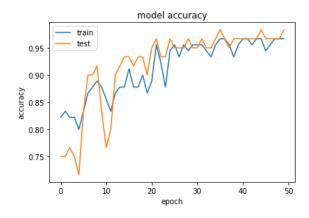


```
# Test model
pred = model.predict(test X)
print(pred)
y_classes = [np.argmax(y, axis=None, out=None) for y in pred]
print(y_classes) # result of prediction
# model performance
score = model.evaluate(test_X, test_y, verbose=0)
print('Test loss:', score[0])
print('Test accuracy:', score[1])
```

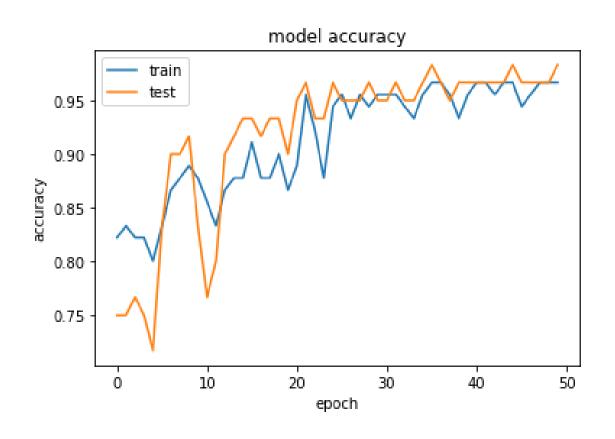
```
In [98]: print(pred)
                                            In [99]: print(y classes)
[[0.8795931 0.09356935 0.02683761]
                                            [0, 2, 2, 1, 0, 1, 2, 0, 2, 0, 0, 0, 1]
 [0.01129904 0.30443922 0.68426174]
                                            2, 2, 2, 0, 2, 2, 1, 0, 2, 2, 0, 2, 2
 [0.01580369 0.33628127 0.64791507]
                                            2, 2, 2, 0, 2, 2, 2, 2, 2, 0]
 [0.11909769 0.45477384 0.42612845]
 [0.94082993 0.0478655 0.01130463]
                                             In [100]: print('Test loss:', score[0])
 [0.09734467 0.45580214 0.44685316]
                                                  ...: print('Test accuracy:', score[1])
 [0.01631222 0.31903324 0.66465455]
                                             Test loss: 0.485007252295812
                                             Test accuracy: 0.7333333492279053
```



```
# summarize history for accuracy
plt.plot(disp.history['accuracy'])
plt.plot(disp.history['val_accuracy'])
plt.title('model accuracy')
plt.ylabel('accuracy')
plt.xlabel('epoch')
plt.legend(['train', 'test'], loc='upper left')
plt.show()
```







Note. 이 코드는 실행할 때 마다 weight value 의 초기값을 random 하게 할당하기 때문에 학습할 때 마다 결과가 다르게 나온다.

```
# model weights
for lay in model.layers:
      print(lay.name)
      print(lay.get weights())
                                                                      H.layer 1 H.layer 2 Output
  In [104]: for lay in model.layers:
       ...: print(lay.name)
       ...: print(lay.get_weights())
  dense 32
  [array([[ 0.5289092 , -0.15780598, 0.01167274, -0.02926999, 0.05397177,
           0.43013152, 0.17118984, -0.38582698, 0.5064401, 0.3571461],
         [ 0.3323511 , -0.33000568, 0.04344529, -0.31324032, -0.21702161,
                                                                           4 x 10
          0.5186461 , 0.8281973 , 0.4229083 , 0.10245298, -0.12464952],
         [ 0.56385374, 0.08948827, -0.29152125, -0.49173585, -0.1845164 ,
         -0.31305426, -0.64084774, 1.0362344, -0.448405, -0.4769767],
         [-0.11043015, -0.15350175, -0.45630628, -0.45117027, -0.21350479,
         -0.0442345 , -0.07903751, 0.9940308 , -0.5409244 , 0.00975268]],
        dtype=float32), array([ 0.12229796,  0. ,  0.
                                                                           , 0.
         -0.02556496, 0.35797217, -0.20785096, 0.23644954, 0.12502621],
        dtype=float32)]
  dense 33
  [array([[ 0.35183033, -0.34312344, -0.31698382, -0.17515983, 0.28551614,
           0.47306767, 0.14369607, 0.26379722, -0.4349746, 0.54658985],
 20
```

[Note]

Save model & reload

```
# save model
model.save('path/to/location')

# load model
from tensorflow import keras
model = keras.models.load_model('path/to/location')
```

Reference:

https://www.tensorflow.org/guide/keras/save_and_serialize

7

https://keras.io/ko/initializers/

- Initializers
 - Zeros
 - Ones
 - Constant
 - RandomNormal
 - RandomUniform
 - he_normal
 - He_uniform
 - lecun_normal

Initialize weight, bias values

https://keras.io/activations/

- Activation 함수
 - softmax
 - relu
 - tanh
 - sigmoid
 - elu
 - selu
 - softplus
 - softsign
 - hard_sigmoid
 - linear

```
model.add(Dense(64, activation='tanh'))
```

Loss function

https://keras.io/losses/

- mean_squared_error
- categorical_crossentropy
- mean_absolute_error
- mean_absolute_percentage_error
- mean_squared_logarithmic_error
- squared_hinge
- hinge
- categorical_hinge
- logcosh
- sparse_categorical_crossentropy
- binary_crossentropy
- kullback_leibler_divergence
- poisson
- cosine_proximity

```
model.compile(loss='mean_squared_error', optimizer='sgd')
```

https://keras.io/optimizers/

- Optimizer
 - sgd
 - RMSprop
 - Adagrad
 - Adadelta
 - Adam
 - Adamax
 - Nadam

Optimizers are algorithms or methods used to change the attributes of your neural network such as weights and learning rate in order to reduce the losses.

```
from keras import optimizers

model = Sequential()
model.add(Dense(64, kernel_initializer='uniform', input_shape=(10,)))
model.add(Activation('softmax'))

sgd = optimizers.SGD(lr=0.01, decay=1e-6, momentum=0.9, nesterov=True)
model.compile(loss='mean_squared_error', optimizer=sgd)
```

Metric

https://keras.io/metrics/

- acc
- binary_accuracy
- categorical_accuracy
- sparse_categorical_accuracy
- top_k_categorical_accuracy
- sparse_top_k_categorical_accuracy

A metric is a function that is used to judge the performance of your model

