Insights on Developing Neural Network With Keras using python 3.7

April 8, 2019

This is an insight on Jason Brownlee's Develop Your First Neural Network in Python With Keras Step-By-Step https://machinelearningmastery.com/tutorial-first-neural-network-python-keras/. Note that I do not own the original code. Some coding parts are modified.

Python 3.7 Jupyter Notebook IreneToo 7/4/2019

1 initialize the random number generator with any seed preferred.

The seed value is important in terms of computer security to pseudorandomly generate a strong secret encryption key.

```
In [12]: from keras.models import Sequential
    from keras.layers import Dense
    import numpy
    # fix random seed for reproducibility
    numpy.random.seed(7)
```

1.1 import data from website link into jupyter notebook (csv format)

```
In [13]: import requests # This library is used to make requests to internet

# url // We are storing url of dataset
url = 'https://raw.githubusercontent.com/jbrownlee/Datasets/master/pima-indians-diabe
# requests // We are creating a requests variable with the above url
r = requests.get(url, allow_redirects=True)
# request dataset.url' //
# We are writing the content of above request to 'dataset.url' file
open('pima-indians-diabetes.data.csv', 'wb').write(r.content)

# Now, we have our files downloaded
ds = open('pima-indians-diabetes.data.csv','r')
In [14]: # LOAD DATASET
dataset = numpy.loadtxt("pima-indians-diabetes.data.csv", delimiter=",")
```

```
# split into input (X) and output (Y) variables
X = dataset[:,0:8]
Y = dataset[:,8]
```

2 Keras Sequential model

The Sequential model is a linear stack of layers. You can add layers via the model add() method

3 First layer

The first layer in the network here is technically a hidden layer, hence it has an activation function -----Jason Brownlee

- input_dim=8: input layer have 8 inputs.
- Dense(12, activation='relu'): a hidden layer with 12 neurons, connected to the input layer that use relu activation function. It initializes all weights using a sample of uniform random numbers.

after that add in 2nd layer and then 3rd layer. activate sigmoid function

4 compile model

5 why use binary_crossentropy

in this case we are executing multi-labels classification task, each label corresponds to a binary classification problem, so one comment could have multiple labels ----bangda, Kaggle

• metrics=['accuracy'] A metric function is similar to a loss function, except that the results from evaluating a metric are not used when training the model. You may use any of the loss functions as a metric function.

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768/768 [========]	-	0s	168us/step	-	loss:	0.4754	- acc:	0.7682
Epoch 136/150								
768/768 [====================================	-	0s	230us/step	-	loss:	0.4724	- acc:	0.7721
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