

Winter FDP on Deep Learning & Applications 09-13 Dec. 2019

Introduction to Keras

Keras

- Keras is a high-level neural networks API, written in Python and capable of running on top of TensorFlow, CNTK, or Theano.
- Here we are going to use Keras on the top of TensorFlow.
- Keras Model creation- The core data structure of Keras is a model, a way to organize layers. The simplest type of model is the Sequential model, a linear stack of layers.

from keras.models import Sequential

model = Sequential()

Keras cont...

Adding layers to model - Stacking layers is as easy as .add()

```
from keras.layers import Dense

model.add(Dense(units=64, activation='relu', input_dim=100))

model.add(Dense(units=10, activation='softmax'))
```

Model Compilation - Configure model learning process with .compile()
 model.compile(loss='categorical_crossentropy', optimizer='sgd', metrics=['accuracy'])

Keras cont...

- Model Training To iterate on training data in batches.
 model.fit(x_train, y_train, epochs=5, batch_size=32)
- Evaluate model performance model.evaluate(x_test, y_test, batch_size=128)
- Predict output on new data classes = model.predict(x_test, batch_size=128)

We are going to use House value prediction dataset https://www.kaggle.com/c/zillow-prize-1/data and our objective is to determine whether the house price is above or below the median.

Reading the data file using panda library -

```
import pandas as pd

df = pd.read_csv('housepricedata.csv')

df.head()
```

Data Separation into input and target –

```
dataset= df.values
X = dataset[:,0:10]
Y = dataset[:,10]
```

Data Normalization in the range of 0 to 1 -

```
from sklearn import preprocessing
min_max_scaler = preprocessing.MinMaxScaler()
X_scale = min_max_scaler.fit_transform(X)
X_scale
```

• Splitting the data into training, testing and validation - Here we are splitting the data 80% for training, 10 % for testing and 10% for validation.

```
from sklearn.model_selection import train_test_split
```

```
X_{train}, X_{val} and test, Y_{train}, Y_{val} and test = train_{test} split(X_{scale}, Y_{test} size=0.2)
```

```
X_{val}, X_{test}, Y_{val}, Y_{test} = train_{test_{split}}(X_{val_{and_{test}}}, Y_{val_{and_{test}}}, Y_{test_{size}})
```

Model Building with the help of Dense Layers -

```
from keras.models import Sequential
from keras.layers import Dense
model = Sequential()
model.add(Dense(16, activation='relu', input_shape=(10,)))
model.add(Dense(8, activation='relu'))
model.add(Dense(1, activation='sigmoid'))
model.summary()
```

 Model Compilation with SGD optimizer to evaluate the performance with loss and accuracy metrics.

```
model.compile(optimizer='sgd',

loss='binary_crossentropy',

metrics=['accuracy'])
```

Model Training in batch size of 32 for 30 iterations -

```
hist = model.fit(X_train, Y_train,

batch_size=32, epochs=30,

validation_data=(X_val, Y_val))
```

Model Testing to check the performance of trained model -

```
test_evaluate =model.evaluate(X_test, Y_test)
print("Testing Accuracy:"+ str(test_evaluate[1]*100))
print("Testing Loss:"+ str(test_evaluate[0]))
```

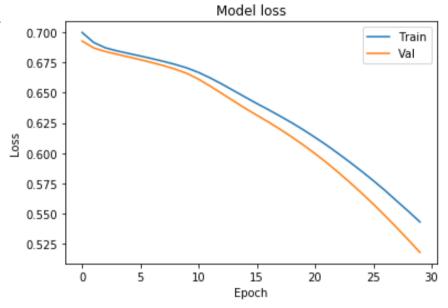
Saving the weights and model for future use -

```
from keras.models import load_model
model.save('my_model.h5')
model.save_weights('my_model_weights.h5')
```

plt.show()

 Plot the training and validation of Loss -

```
import matplotlib.pyplot as plt
plt.plot(hist.history['loss'])
plt.plot(hist.history['val_loss'])
plt.title('Model loss')
plt.ylabel('Loss')
plt.xlabel('Epoch')
plt.legend(['Train', 'Val'], loc='upper right')
```



Plotting the accuracy graph *plt.plot(hist.history['acc'])*

plt.plot(hist.history['val_acc'])

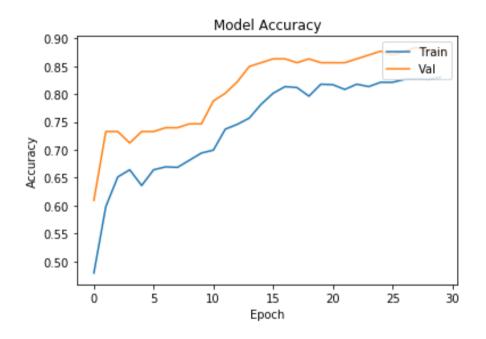
plt.title('Model Accuracy')

plt.ylabel('Accuracy')

plt.xlabel('Epoch')

plt.legend(['Train', 'Val'], loc='upper right')

plt.show()



Exercise

You have to load the weights of the saved model and then add one more layer Dense layer with 8 neuron then check the performance of the model.

References

• https://keras.io/

Thank You