Python For Data Science Cheat Sheet

Keras

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Keras

Keras is a powerful and easy-to-use deep learning library for Theano and TensorFlow that provides a high-level neural networks API to develop and evaluate deep learning models.

A Basic Example

```
>>> import numpy as np
>>> from keras.models import Sequential
>>> from keras.layers import Dense
>>> data = np.random.random((1000,100))
>>> labels = np.random.randint(2, size=(1000,1))
>>> model = Sequential()
>>> model.add(Dense(32,
                    activation='relu',
                    input dim=100))
>>> model.add(Dense(1, activation='sigmoid'))
>>> model.compile(optimizer='rmsprop',
                  loss='binary crossentropy',
                  metrics=['accuracy'])
>>> model.fit(data, labels, epochs=10, batch size=32)
>>> predictions = model.predict(data)
```

Data

Also see NumPy, Pandas & Scikit-Learn

Your data needs to be stored as NumPy arrays or as a list of NumPy arrays. Ideally, you split the data in training and test sets, for which you can also resort to the train test split module of sklearn.cross validation.

Keras Data Sets

```
>>> from keras.datasets import boston housing,
                                   cifar10,
                                   imdb
>>> (x train,y train),(x test,y test) = mnist.load data()
>>> (x_train2,y_train2),(x_test2,y_test2) = boston_housing.load_data()
>>> (x_train3,y_train3),(x_test3,y_test3) = cifar10.load_data()
>>> (x_train4,y_train4), (x_test4,y_test4) = imdb.load data(num words=20000)
>>> num classes = 10
```

Other

```
>>> from urllib.request import urlopen
>>> data = np.loadtxt(urlopen("http://archive.ics.uci.edu/
ml/machine-learning-databases/pima-indians-diabetes/
pima-indians-diabetes.data"),delimiter=",")
>>> X = data[:,0:8]
>>> y = data [:,8]
```

Model Architecture

Sequential Model

```
>>> from keras.models import Sequential
>>> model = Sequential()
>>> model2 = Sequential()
>>> model3 = Sequential()
```

Multilayer Perceptron (MLP)

Binary Classification

```
>>> from keras.lavers import Dense
>>> model.add(Dense(12.
                    input dim=8,
                    kernel initializer='uniform',
                    activation='relu'))
>>> model.add(Dense(8,kernel initializer='uniform',activation='relu'))
>>> model.add(Dense(1, kernel initializer='uniform', activation='sigmoid'))
```

Multi-Class Classification

```
>>> from keras.layers import Dropout
>>> model.add(Dense(512,activation='relu',input shape=(784,)))
>>> model.add(Dropout(0.2))
>>> model.add(Dense(512,activation='relu'))
>>> model.add(Dropout(0.2))
>>> model.add(Dense(10,activation='softmax'))
```

>>> model.add(Dense(64,activation='relu',input_dim=train_data.shape[1])) >>> model.add(Dense(1))

>>> from keras.layers import Activation, Conv2D, MaxPooling2D, Flatten

Convolutional Neural Network (CNN)

```
>>> model2.add(Conv2D(32,(3,3),padding='same',input shape=x train.shape[1:]))
>>> model2.add(Activation('relu'))
>>> model2.add(Conv2D(32,(3,3)))
>>> model2.add(Activation('relu'))
>>> model2.add(MaxPooling2D(pool size=(2,2)))
>>> model2.add(Dropout(0.25))
>>> model2.add(Conv2D(64,(3,3), padding='same'))
>>> model2.add(Activation('relu'))
>>> model2.add(Conv2D(64,(3, 3)))
>>> model2.add(Activation('relu'))
>>> model2.add(MaxPooling2D(pool size=(2,2)))
>>> model2.add(Dropout(0.25))
>>> model2.add(Flatten())
>>> model2.add(Dense(512))
>>> model2.add(Activation('relu'))
>>> model2.add(Dropout(0.5))
>>> model2.add(Dense(num classes))
>>> model2.add(Activation('softmax'))
```

Recurrent Neural Network (RNN)

```
>>> from keras.klayers import Embedding,LSTM
>>> model3.add(Embedding(20000,128))
>>> model3.add(LSTM(128,dropout=0.2,recurrent dropout=0.2))
>>> model3.add(Dense(1,activation='sigmoid'))
```

Preprocessing

Sequence Padding

```
>>> from keras.preprocessing import sequence
>>> x train4 = sequence.pad sequences(x train4, maxlen=80)
>>> x test4 = sequence.pad sequences(x test4, maxlen=80)
```

One-Hot Encoding

```
>>> from keras.utils import to categorical
>>> Y train = to categorical(y train, num_classes)
>>> Y_test = to_categorical(y_test, num_classes)
>>> Y_train3 = to_categorical(y_train3, num_classes)
>>> Y_test3 = to_categorical(y_test3, num_classes)
```

Train and Test Sets

```
>>> from sklearn.model selection import train test split
>>> X train5,X test5,y train5,y test5 = train test split(X,
                                                           test size=0.33,
random state=42)
```

Also see NumPy & Scikit-Learn

Standardization/Normalization

```
>>> from sklearn.preprocessing import StandardScaler
>>> scaler = StandardScaler().fit(x train2)
>>> standardized X = scaler.transform(x train2)
>>> standardized X test = scaler.transform(x test2)
```

Inspect Model

```
Model output shape
>>> model.output shape
>>> model.summary()
                                     Model summary representation
>>> model.get config()
                                     Model configuration
>>> model.get weights()
                                     List all weight tensors in the model
```

Compile Model

```
MLP: Binary Classification
>>> model.compile(optimizer='adam',
```

```
loss='binary crossentropy',
                  metrics=['accuracy'])
MLP: Multi-Class Classification
>>> model.compile(optimizer='rmsprop',
                  loss='categorical crossentropy',
```

MLP: Regression

```
>>> model.compile(optimizer='rmsprop',
                  loss='mse',
                  metrics=['mae'])
```

Recurrent Neural Network

```
>>> model3.compile(loss='binary crossentropy',
                  optimizer='adam',
                  metrics=['accuracv'])
```

metrics=['accuracy'])

Model Training

```
>>> model3.fit(x train4.
              y train4,
              batch size=32,
             epochs=15,
             verbose=1,
             validation data=(x test4, y test4))
```

Evaluate Your Model's Performance

```
>>> score = model3.evaluate(x test.
                             batch size=32)
```

Prediction

```
>>> model3.predict(x test4, batch size=32)
>>> model3.predict classes(x test4,batch size=32)
```

Save/Reload Models

```
>>> from keras.models import load model
>>> model3.save('model file.h5')
>>> my model = load model('my model.h5')
```

Model Fine-tuning

Optimization Parameters

```
>>> from keras.optimizers import RMSprop
>>> opt = RMSprop(lr=0.0001, decay=1e-6)
>>> model2.compile(loss='categorical crossentropy',
                   optimizer=opt,
                   metrics=['accuracy'])
```

Early Stopping

```
>>> from keras.callbacks import EarlyStopping
>>> early stopping monitor = EarlyStopping(patience=2)
>>> model3.fit(x train4,
             y train4,
             batch size=32,
             epochs=15,
             validation data=(x test4, y test4),
             callbacks=[early stopping monitor])
```

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