Keras Cheat Sheet

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Keras is a powerfuland 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 no

>>> 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

input_dim=100))

>>> model.add(Dense(1, activation='sigmoid'))

>>> model.compile(optimizer='rmsprop',

loss='binary crossentropy'.

metrics=['accuracy'])

Data

Also see NumPv. 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,

>>> (x train.v train).(x test.v 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

>>> model.fit(data,labels,epochs=10,batch_size=32)

>>> predictions = model.predict(data)

Other

>>> from urllib.request import urlopen

>>> data = np.loadtxt(urlopen("http://archive.ics.uci.edu/ ml/machine-learning-databases/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()

Multilaver Perceptron (MLP)

Binary Classification

>>> from keras.layers 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))

Convolutional Neural Network (CNN)

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

>>> 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.klavers 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'))

Inspect Model

>>> model.output shape

>>> model.summarv()

>>> model.get_config()

>>> model.get weights()

Model output shape Model summary representation Model configuration

List all weight tensors in the model

Prediction

>>> model3.predict(x_test4, batch_size=32)

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

Model Training

>>> model3.fit(x train4,

Model Fine-tuning

Optimization Parameters

>>> from keras.optimizers import RMSprop

>>> opt = RMSprop(lr=0.0001, decay=1e-6)

>>> from keras.callbacks import EarlyStopping

y_train4, batch_size=32,

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

>>> model.compile(optimizer='rmspro

>>> model.compile(optimizer='rmsprop',

Recurrent Neural Network

Save/ Reload Models

>>> from keras.models import load model

>>> my_model = load_model('my_model.h5')

>>> model3 save('model_file.h5')

MLP: Regression

>>> model3.compile(loss='binar

Early Stopping

>>> model3.fit(x_train4,

Compile Model

>>> model2.compile(loss='categorical crossentropy',

optimizer=opt,

>>> early_stopping_monitor = EarlyStopping(patience=2)

metrics=['accuracy'])

epochs=15, validation_data=(x_test4,y_test4),

metrics=['accuracy'])

loss='categorical_cro metrics=['accuracy'])

metrics=['mae'])

optimizer='adam', metrics=['accuracy'])

MLP: Multi-Class Classification

callbacks=[early stopping monitor])

y_train4, batch size=32. verbose=1

validation_data=(x_test4,y_test4))

Evaluate Your Model's Performance

>>> score = model3.evaluate(x test

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)

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)