Inspect Model

DataCamp

Model output shape Model summary representation Model configuration List all weight tensors in the model Sequential Model model.output_shape model.summary() model.get_config() model.get_weights() Keras from keras.models import Sequential model = Sequential() model2 = Sequential() model3 = Sequential() Learn Python for data science Interactively at www.DataCamp.com **9** Compile Model MLP:Binary Classification >>> model.compile(optimizer='adam', loss='binary_crossentropy', mlp:Multi-Class Classification >>> model.compile(optimizer='msprop', loss='categorical_crossentr metrics=['accuracy']) MLP:Regression Multilayer Perceptron (MLP) Binary Classification >>> from keras.layers imp >>> model.add(Dense(12, 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 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.random((2,size=(1000,1)) >>> model = Sequential() >>> model.add(Dense(32 articles = 100)) >>> model.add(Dense(1, activation='relu', input_dim=100)) >>> model.compile(optimizer='rmsprop' logar binary_consentropy', model-grade = 10 factores = 10 f >>> model.compile(optimizer='rmsprop', loss='mse', metrics=['mae']) >>> model.add(Dense(l,kernel_initializer='uniform',activation='sign Multi-Class (Cassification >>> 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(Dense(10.2)) >>> model.add(Dense(10.2)) >>> model.add(Dense(10,activation='softmax')) Percersing Regression >>> model.add(Dense(64,activation='relu',input_dim=train_data.shape[1])) >>> model.add(Dense(1)) **Model Training** Convolutional Neural Network (CNN) epochs=15, verbose=1, validation_data=(x_test4,y_test4)) from keras.layers import Activation, Conv2D, MaxPooling2D, Flatten model2.add(Conv2D(32, (3, 3, a, padding='same', input_shape=x_train.shape[1:])) model2.add(Conv2D(32, (3, 3))) model2.add(Conv2D(32, (3, 3))) model2.add(MaxPooling2D(pool_size=(2,2))) model2.add(MaxPooling2D(pool_size=(2,2))) model2.add(Conv2D(64, (3, 3), padding='same')) Data **Evaluate Your Model's Performance** 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. y_test, batch_size=32) model2.add(Copput(vi.25)) model2.add(Conv2D(64,(3,3), padding='same')) model2.add(Conv2D(64,(3,3))) model2.add(Conv2D(64,(3,3))) model2.add(Activation('relu')) model2.add(MaxPooling2D(pool_size=(2,2))) model2.add(Dropout(0.25)) Keras Data Sets Prediction >> model3.predict(x_test4, batch_size=32) >> model3.predict_classes(x_test4,batch_size=32) model2.add(Propout(0.25)) model2.add(Platten()) model2.add(Dense(512)) model2.add(Activation('relu')) model2.add(Dropout(0.5)) model2.add(Dense(num_classes)) model2.add(Activation('softmax Save/ Reload Models from keras.models import load_model model3.save('model_file.h5') my model = load model('my model.h5') Model Fine-tuning from urllib.request import urlopen data = np.loadtxt(urlopen("http://archive.ics.uci.edu/ machine-learning-databases/pima-indians-diabetes/ a-indians-dlabetes.data"),delimiter=",") Recurrent Neural Network (RNN) Optimization Parameters from keras.klayers import Embedding,LSTM model3.add(Embedding(20000,128)) model3.add(LSTM(128,tropout=0.2,recurrent_dropout=0.2)) model3.add(Dense(1,activation='sigmoid')) Preprocessing Train and Test Sets >> from keras.callbacks import EarlyStopping >> early stopping_monitor = EarlyStopping(patience=2) >> model3.fit(x_train4, >>> from keras.preprocessing import sequence >>> x_train4 = sequence.pad_sequences(x_train4,maxlen=80) >>> x_test4 = sequence.pad_sequences(x_test4,maxlen=80) ty_train4, batch size=32, epochs=15, validation_data=(x_test4,y_test4), callbacks=[early_stopping_monitor]) Standardization/Normalization om keras.utils import to categorical :rain = to_categorical(y_train, num_classes) :est = to_categorical(y_test, num_classes) :rain3 = to_categorical(y_train3, num_classes) :est3 = to_categorical(y_test3, num_classes)

>> scaler = StandardScaler().fft(x train2) >> standardized X = scaler.transform(x train2) >> standardized X test = scaler.transform(x test2)

Python For Data Science Cheat Sheet Model Architecture