

Introduction

KerasTuner is a general-purpose hyperparameter tuning library. It has strong integration with Keras workflows, but it isn't limited to them: you could use it to tune scikit-learn models, or anything else. In this tutorial, you will see how to tune model architecture, training process, and data preprocessing steps with KerasTuner. Let's start from a simple example.

https://keras.io/guides/keras_tuner/getting_started/

Import

```
import tensorflow as tf
from tensorflow import keras
from keras import layers
import keras_tuner as kt
import numpy as np

print(tf.__version__)

2.8.0
```

Functions

```
In [4]:
         def build_model_units_activation_dropout_lr(hp):
             model = keras.Sequential()
             model.add(layers.Flatten())
             model.add(
                 layers.Dense(
                     # Tune number of units.
                     units=hp.Int("units", min_value=32, max_value=512, step=32),
                     # Tune the activation function to use.
                     activation=hp.Choice("activation", ["relu", "tanh"]),
                 )
             # Tune whether to use dropout.
             if hp.Boolean("dropout"):
                 model.add(layers.Dropout(rate=0.25))
             model.add(layers.Dense(10, activation="softmax"))
             # Define the optimizer learning rate as a hyperparameter.
             learning_rate = hp.Float("lr", min_value=1e-4, max_value=1e-2, sampling="log")
             model.compile(
                 optimizer=keras.optimizers.Adam(learning_rate=learning_rate),
                 {\tt loss="categorical\_crossentropy",}
```

```
metrics=["accuracy"],
              return model
In [5]:
          hp = kt.HyperParameters() # Hyperparameter class.
          print(hp.Int("units", min_value=32, max_value=512, step=32))
In [6]:
          # Template for build and compile.
          def call_existing_code(units, activation, dropout, lr):
              model = keras.Sequential()
              model.add(layers.Flatten())
              model.add(layers.Dense(units=units, activation=activation))
              if dropout:
                  model.add(layers.Dropout(rate=0.25))
              model.add(layers.Dense(10, activation="softmax"))
              model.compile(
                  optimizer=keras.optimizers.Adam(learning_rate=lr),
                  loss="categorical_crossentropy",
                  metrics=["accuracy"],
              return model
In [7]:
          def build model(hp):
            units = hp.Int("units", min_value=32, max_value=512, step=32)
            activation = hp.Choice("activation", ['relu', 'tanh'])
            dropout = hp.Boolean('dropout')
            lr = hp.Float('lr', min_value=1e-4, max_value=1e-2, sampling='log')
            model = call_existing_code(units=units, activation=activation, dropout=dropout, lr=lr)
            return model
In [8]:
          # check if the model builds.
          hp = kt.HyperParameters()
          build model(hp)
         <keras.engine.sequential.Sequential at 0x7f961326f350>
Out[8]:
In [9]:
          def build_compile_model(hp):
              model = keras.Sequential()
              model.add(layers.Flatten())
              # Tune the number of layers.
              for i in range(hp.Int("num_layers", 1, 3)):
                print('number of layers')
                model.add(
                layers.Dense(
                    # Tune number of units separately.
                    units=hp.Int(f"units_{i}", min_value=32, max_value=512, step=32),
                    activation=hp.Choice("activation", ["relu", "tanh"]),
                    )
                )
              print()
              print('Dropout is {}'.format(hp.Boolean('dropout')))
              if hp.Boolean("dropout"):
                  model.add(layers.Dropout(rate=0.25))
              model.add(layers.Dense(10, activation="softmax"))
              learning_rate = hp.Float("lr", min_value=1e-4, max_value=1e-2, sampling="log")
              print('learning rate is {}'.format(learning_rate))
              model.compile(
                  optimizer=keras.optimizers.Adam(learning_rate=learning_rate),
                  loss="categorical_crossentropy",
                  metrics=["accuracy"],
              return model
          build_model(kt.HyperParameters())
         <keras.engine.sequential.Sequential at 0x7f9589392f10>
Out[9]:
In [10]:
          tuner = kt.RandomSearch(
```

```
hypermodel=build_compile_model,
             objective="val_accuracy",
             max_trials=3,
             executions_per_trial=2,
             overwrite=True,
             directory="my_dir",
             project_name="helloworld",
        number of layers
        Dropout is False
        learning rate is 0.0001
In [11]:
        tuner.search_space_summary()
         Search space summary
        Default search space size: 5
        num_layers (Int)
         {'default': None, 'conditions': [], 'min_value': 1, 'max_value': 3, 'step': 1, 'sampling': None}
        units 0 (Int)
         {'default': None, 'conditions': [], 'min_value': 32, 'max_value': 512, 'step': 32, 'sampling': None}
         activation (Choice)
         {'default': 'relu', 'conditions': [], 'values': ['relu', 'tanh'], 'ordered': False}
         dropout (Boolean)
         {'default': False, 'conditions': []}
         lr (Float)
         {'default': 0.0001, 'conditions': [], 'min_value': 0.0001, 'max_value': 0.01, 'step': None, 'sampling':
         'log'}
         Load the dataset
In [12]:
         (x, y), (x_test, y_test) = keras.datasets.mnist.load_data()
         # In axis0, until last 10000 elements. And, all elements in all other axes.
         x_{train} = x[:-10000]
         y_{train} = y[:-10000]
         # In axis0, last 10000 elements. And, all elements in all other axes.
         x_val = x[-10000:]
         y_val = y[-10000:]
         Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist.npz
        In [13]:
         print('shape of x dataset {}'.format(x.shape))
         print('shape of y dataset {}'.format(y.shape))
         print()
         print('shape of xtrain dataset {}'.format(x_train.shape))
         print('shape of ytrain dataset {}'.format(y_train.shape))
         shape of x dataset (60000, 28, 28)
         shape of y dataset (60000,)
         shape of xtrain dataset (50000, 28, 28)
         shape of ytrain dataset (50000,)
In [14]:
         # Add a new dimension at the end. Need?
         de_x_train = np.expand_dims(x_train, axis=-1)
         de_x_val = np.expand_dims(x_val, axis=-1)
         de_x_test = np.expand_dims(x_test, axis=-1)
         print('shape of dimension extended x_train {}'.format(de_x_train.shape))
         print('shape of dimension extended x_val {}'.format(de_x_val.shape))
         print('shape of dimension extended x_test {}'.format(de_x_test.shape))
         shape of dimension extended x_train (50000, 28, 28, 1)
         shape of dimension extended x_val (10000, 28, 28, 1)
         shape of dimension extended x_test (10000, 28, 28, 1)
In [15]:
         # Scale the value between 0 and 1.
         de_x_{train} = de_x_{train.astype('float32')} / 255.0
         de_x_val = de_x_val.astype('float32') / 255.0
         de_x_test = de_x_test.astype('float32') / 255.0
```

```
in Libl: | num_classes = 10
          ca_y_train = keras.utils.to_categorical(y_train, num_classes)
          ca_y_val = keras.utils.to_categorical(y_val, num_classes)
          ca_y_test = keras.utils.to_categorical(y_test, num_classes)
In [17]:
          print('shape of y_train is {}'.format(y_train.shape))
          print('shape of ca_y_train is {}'.format(ca_y_train.shape))
          print()
          print('example of {}: {}'.format(y_train[5], ca_y_train[5]))
          print('example of {}: {}'.format(y_train[15]), ca_y_train[15]))
print('example of {}: {}'.format(y_train[25]), ca_y_train[25]))
          print('example of {}: {}'.format(y_train[35], ca_y_train[35]))
         shape of y_{train} is (50000,)
         shape of ca_y_train is (50000, 10)
         example of 2: [0. 0. 1. 0. 0. 0. 0. 0. 0. 0.]
         example of 7: [0. 0. 0. 0. 0. 0. 0. 1. 0. 0.]
         example of 2: [0. 0. 1. 0. 0. 0. 0. 0. 0. 0.]
         example of 5: [0. 0. 0. 0. 0. 1. 0. 0. 0. 0.]
In [18]:
          a = np.array([[1,2,3], [11,12,13], [21,22,23], [31,32,33]])
          print(a[:2]) # print from [beginning, 2]
          print()
          print(a[2:]) # print from (2, everything]
          [[1 2 3]
           [11 12 13]]
          [[21 22 23]
          [31 32 33]]
In [19]:
          tuner.search(de_x_train, ca_y_train, epochs=2, validation_data=(de_x_val, ca_y_val))
         Trial 3 Complete [00h 00m 42s]
         val_accuracy: 0.9628500044345856
         Best val_accuracy So Far: 0.9671000242233276
         Total elapsed time: 00h 01m 30s
         INFO:tensorflow:Oracle triggered exit
         Get the best models
In [20]:
          models = tuner.get_best_models(num_models=2)
          best_model = models[0]
          # Build the model
```

```
best_model.build(input_shape=(None, 28, 28))
best_model.summary()
number of layers
Dropout is False
learning rate is 0.003449485159787074
number of layers
number of layers
Dropout is False
learning rate is 0.0003271739205768162
Model: "sequential"
                             Output Shape
Layer (type)
                                                       Param #
flatten (Flatten)
                             (None, 784)
dense (Dense)
                             (None, 192)
                                                       150720
dense_1 (Dense)
                                                       1930
                             (None, 10)
Total params: 152,650
```

Trainable params: 152,650 Non-trainable params: 0

```
curier . courco_oummary()
         Results summary
        Results in my\_dir/helloworld
        Showing 10 best trials
         <keras_tuner.engine.objective.Objective object at 0x7f958d582a90>
        Trial summary
        Hyperparameters:
        num_layers: 1
        units_0: 192
         activation: tanh
         dropout: False
        lr: 0.003449485159787074
        units_1: 480
        units_2: 64
        Score: 0.9671000242233276
        Trial summary
        Hyperparameters:
        num_layers: 2
        units_0: 384
        activation: tanh
         dropout: False
        lr: 0.0003271739205768162
        units_1: 288
        units_2: 160
        Score: 0.9628500044345856
        Trial summary
        Hyperparameters:
        num_layers: 3
        units_0: 224
        activation: tanh
         dropout: False
        lr: 0.005256055651516419
        units_1: 32
        units_2: 32
        Score: 0.943450003862381
        Choose best hyperparameters
In [22]:
         best_hps = tuner.get_best_hyperparameters(5)
         model = build_compile_model(best_hps[1])
        number of layers
        number of layers
        Dropout is False
        learning rate is 0.0003271739205768162
In [23]:
         model.build(input_shape=(None, 28, 28))
In [24]:
         model.summary()
        Model: "sequential_1"
                                    Output Shape
         Layer (type)
                                                             Param #
         flatten_1 (Flatten)
                                    (None, 784)
         dense_3 (Dense)
                                    (None, 384)
                                                             301440
         dense 4 (Dense)
                                    (None, 288)
                                                             110880
         dense_5 (Dense)
                                    (None, 10)
         _____
         Total params: 415,210
         Trainable params: 415,210
```

```
In [36]: best_hps[0].get('units_1')
Out[36]: 480
In []:
```

Non-trainable params: 0