Train-Test Split

Introduction

In this exercise we will work with the concept of splitting datasets into training and test sets ad dig deeper into the implementations of said concepts, to get a better understading of what is happening in the background of machine learning.

Setup the Housing data form §2 [HOML]

We use the housing data from the book, this cell will set everything up for you...

```
# To support both python 2 and python 3
from __future__ import division, print_function, unicode_literals
# Common imports
import numpy as np
import os
# to make this notebook's output stable across runs
np.random.seed(42)
# To plot pretty figures
%matplotlib inline
import matplotlib
import matplotlib.pyplot as plt
plt.rcParams['axes.labelsize'] = 14
plt.rcParams['xtick.labelsize'] = 12
plt.rcParams['ytick.labelsize'] = 12
# Where to save the figures
PROJECT ROOT DIR = ".."
CHAPTER ID = "end to end project"
IMAGES_PATH = os.path.join(PROJECT_ROOT_DIR, "images", CHAPTER_ID)
def save_fig(fig_id, tight_layout=True, fig_extension="png", resolution=300):
    #path = os.path.join(IMAGES_PATH, fig_id + "." + fig extension)
    print("IGNORING: Saving figure", fig_id)
    #if tight layout:
         plt.tight layout()
    #plt.savefig(path, format=fig_extension, dpi=resolution)
# Ignore useless warnings (see SciPy issue #5998)
import warnings
warnings.filterwarnings(action="ignore", message="^internal gelsd")
import os
import tarfile
from six.moves import urllib
DOWNLOAD_ROOT = "https://raw.githubusercontent.com/ageron/handson-ml/master/"
HOUSING_PATH = os.path.join("../datasets", "housing")
HOUSING_URL = DOWNLOAD_ROOT + "../datasets/housing/housing.tgz"
def fetch_housing_data(housing_url=HOUSING_URL, housing_path=HOUSING_PATH):
    if not os.path.isdir(housing_path):
        os.makedirs(housing_path)
    tgz_path = os.path.join(housing_path, "housing.tgz")
    urllib.request.urlretrieve(housing_url, tgz_path)
    housing_tgz = tarfile.open(tgz_path)
    housing tgz.extractall(path=housing path)
    housing tgz.close()
fetch housing data()
import pandas as pd
def load_housing_data(housing_path=HOUSING_PATH):
    csv_path = os.path.join(housing_path, "housing.csv")
    return pd.read csv(csv path)
```

```
housing = load_housing_data()

#housing.head()
print("housing.shape=",housing.shape,"\n")
housing.info()

%matplotlib inline
import matplotlib.pyplot as plt
#housing.hist(bins=50, figsize=(20,15))
#save_fig("attribute_histogram_plots")
#plt.show()

# NOTE: ITMAL, convert Pandas dataframe to numpy array, i.e. matrix
# and use H Later instead of housing
H = housing.values
print('H.shape=',H.shape,", type(H)=",type(H))
print('OK')

housing.shape= (20640, 10)
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 20640 entries, 0 to 20639
Data columns (total 10 columns):
longitude
                                20640 non-null float64
latitude
                                20640 non-null float64
housing_median_age 20640 non-null float64 total_rooms 20640 non-null float64 total_bedrooms 20433 non-null float64 population 20640 non-null float64
population
                                20640 non-null float64
households 20640 non-null float64
median_income 20640 non-null float64
median_house_value 20640 non-null float64
ocean proximity 20640 non-null object
                                20640 non-null float64
ocean proximity
                                20640 non-null object
dtypes: float64(9), object(1)
memory usage: 1.6+ MB
H.shape= (20640, 10) , type(H)= <class 'numpy.ndarray'>
OK
```

Create our own train-test split function

Qa Create Your Own Split Function

create your own split function, that can do the data shuffling (as it is now) or do a simpler split without shuffling.

Notice that it would be better to name the function <code>my_split_train_test</code> to avoid clashing problems later with the Scikit-learn function of the same name. The <code>test_ratio</code> parameter has also been renamed to <code>test_size</code>.

Also note that the split function in [HOML] operates on Pandas data frames, and this will give us a mixup problem later, when we pass the function numpy arrays (matrices).

Test that your new split function returns the same number of train and test data no matter if shuffleling is on or off, using the test stub below.

Qa Implementation

In [3]:

```
# TODO: Qa...define your my split train test here
def my_split_train_test(data, test_size, shuffle=False):
    if shuffle == True:
        indices = np.random.permutation(len(data))
    else:
        indices = np.arange(len(data))
    test set size = int(len(data) * test_size)
    test indices = indices[len(data)-test set size:len(data)]
    train indices = indices[0:len(data)-test set size]
    return data.iloc[train indices], data.iloc[test indices]
# TEST VECTORS: use the housing panda dataframe or the H numpy object, your choice
dat=housing
#dat=H
def TestSize(train set, test set):
    # works only for 0.2 split
    expected_n_train=16512
    expected n test=4128
    assert len(train_set) == expected_n_train, 'Oh, mismatch in expected train n'
    assert len(test_set) ==expected_n_test, 'Oh, mismatch in expected test n'
    print(len(train_set), "train +", len(test_set), "test","..OK")
train_set, test_set = my_split_train_test(dat, 0.2, shuffle=False)
TestSize(train set, test set)
train_set, test_set = my_split_train_test(dat, 0.2, shuffle=True)
TestSize(train_set, test_set)
from sklearn.model_selection import train_test_split
train set, test set = train test split(dat, test size=0.2, shuffle=True, random state=4
2)
TestSize(train_set, test_set)
train_set, test_set = train_test_split(dat, test_size=0.2, shuffle=False)
TestSize(train_set, test_set)
16512 train + 4128 test ..OK
16512 train + 4128 test ..OK
16512 train + 4128 test ..OK
```

Qa Result

16512 train + 4128 test ..OK

In the above printouts it's clear that our implementation of the Split_train_test function splits the set in two parts of the same size as the inbuilt version

Qb Why Shuffling

Explain why disabling shuffling is a bad idea?

Disabling shuffling is a bad idea since you may not have acquired your data randomly. Maybe you've gone from sector to sector which means that taking the first 20% of the dataset might only contain data from one sector, instead of a mix of all the sectors.

Qc Test and Compare

Compare your split function with the one from Scikit-learn, first using the simple X-y data set generated below and then using the housing data via the H numpy array variable.

Splitting the dataset via your split function and the built-in split does not yield a logical true for the comparison

Why is it so? Find the exact values in H[i,j] that are not equal and explain the problem.

Qc Implementation

In [9]:

```
# Simple data for Qc
import numpy as np
X1, y = np.arange(10).reshape((5, 2)), np.array([list(range(5))])
print("X=",X1)
print("y=",y)
# TODO: Qc...
dataset = pd.DataFrame({'Column1':X1[:,0],'Column2':X1[:,1]})
H = dat
def test_tts(X):
    # TEST VECTORS: notice that H is not splitted into X-v parts
    train, test = train test split(X, test size=0.2, shuffle=False)
    print("build-in split: len(train)=",len(train),", len(test)=", len(test))
    train_my, test_my = my_split_train_test(X, test_size=0.2, shuffle=False)
    print("my split:
                           len(train)=",len(train_my),", len(test)=", len(test_my))
    # Test for equality here...
    assert train.shape==train my.shape
    equal train=(train.values == train my.values).all().all()
    equal_test =(test.values == test_my.values).all().all()
    # TODO: why not equal?
    print("equal_train=", equal_train, ", equal_test=",equal_test)
test tts(H)
test_tts(dataset)
X = [[0 \ 1]]
 [2 3]
 [4 5]
```

```
X= [[0 1]
  [2 3]
  [4 5]
  [6 7]
  [8 9]]
y= [[0 1 2 3 4]]
build-in split: len(train)= 16512 , len(test)= 4128
my split: len(train)= 16512 , len(test)= 4128
equal_train= False , equal_test= False
build-in split: len(train)= 4 , len(test)= 1
my split: len(train)= 4 , len(test)= 1
equal_train= True , equal_test= True
```

Qc Results

In the above printouts we can see the two sets are equal in value (not just size). However changing the splitting ratio might cause some differences with smaller datasets, since one algorithm might round up while the other will round down.