Week 4 Assignment

Deployment on Flask

Name: Omar Pablo Jazouli Martínez

Batch code: LICAN01 Submission date:18/03/2021 Submitted to: Data Glacier

In order to deploy a Machine Learning model we have taken the following four steps.

1 Create and save model

As creating a good and useful model goes beyond the scope of this assignment, we have selected the toy dataset of the Titanic passengers. The aim is to predict whether a passenger survived, taking into account their ticket class, gender and age. A decision tree was used to do so and then the model is saved in disk.

```
data = pd.read_csv('https://raw.githubusercontent.com/laxmimerit/All-CSV-ML-Data-Files-Download/master/titanic.csv')
data.dropna(inplace=True)
use_cols = ["Pclass", 'Sex', "Age"]
X_train, X_test, y_train, y_test = train_test_split(
     data[use cols],
     test size=0.3,
     random_state=0)
X_train.shape, X_test.shape
# call the model
from sklearn import tree
clf = tree.DecisionTreeClassifier(max depth=4)
# train the model
model = clf.fit(X_train[["Pclass","Sex","Age"]].values, y_train)
# make predictions on train and test set
pred_train = model.predict_proba(X_train[["Pclass", "Sex", "Age"]].values)
pred_test = model.predict_proba(X_test[["Pclass", "Sex", "Age"]].values)
print('DecisionTree roc-auc: {}'.format(roc_auc_score(y_train, pred_train[:,1])))
print('Test set')
print('DecisionTree roc-auc: {}'.format(roc_auc_score(y_test, pred_test[:,1])))
Train set
DecisionTree roc-auc: 0.8846861045442606
DecisionTree roc-auc: 0.8424908424908426
# save the model to disk
filename = 'titanic model.sav
pickle.dump(model, open(filename, 'wb'))
```

2 Flask app

Once we have created and saved the model, it is time to build a flask app to deploy it later on.

```
from flask import Flask,request,render_template
import pickle
import numpy as np
app=Flask(__name__)
model=pickle.load(open("titanic_model.sav", 'rb'))
@app.route("/")
def home():
    return render_template('index.html')
@app.route('/predict',methods=["POST"])
def predict():
    int_features=[int(x) for x in request.form.values()]
    final_features=np.array(int_features).reshape(1,-1)
    prediction=model.predict(final_features)[0]
    if prediction==1:
       output="The passenger survived"
       output="The passenger did not survived"
    return render_template("index.html",prediction_text=output)
if __name__=="__main__":
    app.run(debug=True)
```

This file is called "app.py". In it, we call the file "index.html", which contains the html code of the web application.

3 HTML and CSS templates

The html file has the form to input the data that our model will use to make its prediction (ticket class, gender and age).

It is connected to a css file ("style.css") in order to have a fancy appearance. This file is the one used in the webinar but with the following changes:

```
.login h1 { color: ■#fff; text-shadow: 0 0 10px □rgba(0,0,0,0.3); letter-spacing:1px; text-align:center; }
.login h2 { color: ■white; text-shadow: 0 0 10px □rgba(0,0,0,0); letter-spacing:1px; text-align:center; }
.login form { color: ■white; text-shadow: 0 0 10px □rgba(0,0,0,0); letter-spacing:1px; text-align:left; }
.login h3 { color: ■white; text-shadow: 0 0 10px □rgba(0,0,0,0); letter-spacing:1px; text-align:center; }
```

"Index.html" is inside a folder called "templates" and the folder containing "style.css" is called "static".

4 Deploy model

Finally, we are able to deploy our model on a web application. We run the "app.py" file with the command shell.

```
(base) MacBook-Air-de-Omar:Desktop omarjazoulimartinez$ export FLASK_APP=app.py
(base) MacBook-Air-de-Omar:Desktop omarjazoulimartinez$ flask run
* Serving Flask app "app.py"
* Environment: production
   WARNING: This is a development server. Do not use it in a production deployment.
   Use a production WSGI server instead.
* Debug mode: off
* Running on http://127.0.0.1:5000/ (Press CTRL+C to quit)
```

After using the above direction on our browser we will be redirected to the following page:



One can input the ticket class, the gender and the age of the target passenger. The model will return whether they survived or not after clicking on the "PREDICT" button.

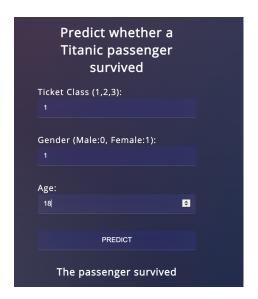


Figure 1: Survived

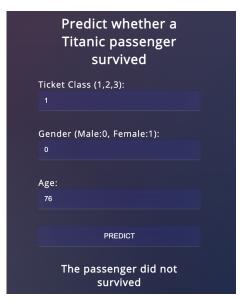


Figure 2: Did not survived