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Batch code: LISUM01.

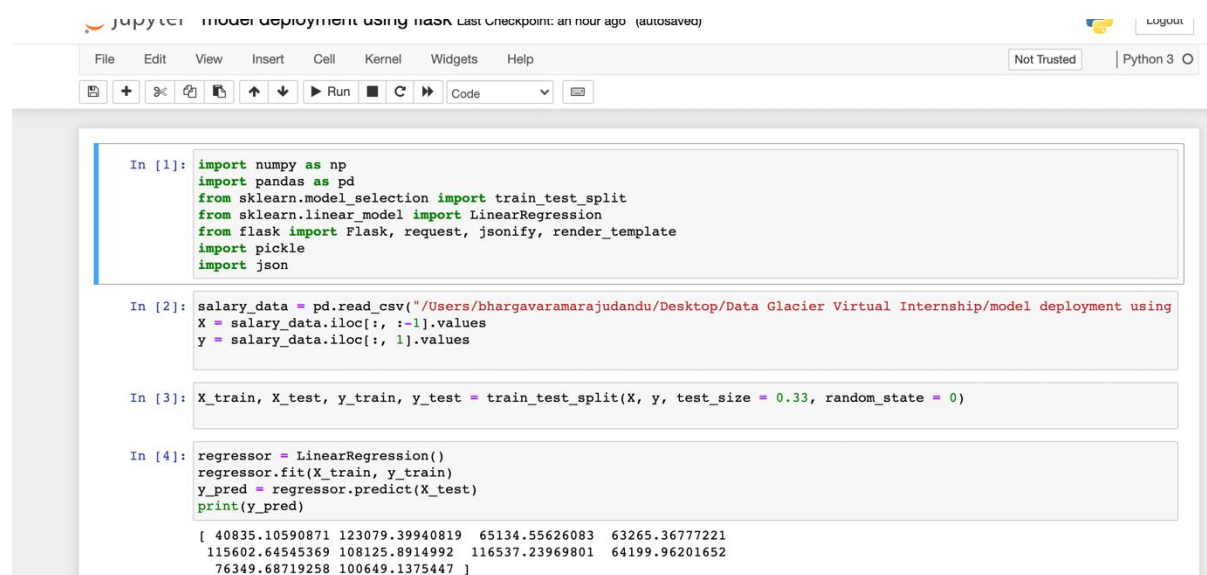
Submission Date: 10-June-2021.

Submitted to: Data Glacier.

## Deployment on Flask

### Step 1: Develop a Machine learning model.

Predict the salary of an employee using Linear Regression Model.



The image shows a Jupyter Notebook interface with the title "model deployment using flask". The notebook contains four code cells. The first cell imports necessary libraries: numpy, pandas, sklearn (train\_test\_split, LinearRegression), flask (Flask, request, jsonify, render\_template), pickle, and json. The second cell reads a CSV file "Data Glacier Virtual Internship/model deployment using" and splits it into X and y. The third cell splits the data into training and testing sets using train\_test\_split with a test\_size of 0.33. The fourth cell creates a LinearRegression model, fits it to the training data, and prints the predictions for the test data.

```
In [1]: import numpy as np
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from flask import Flask, request, jsonify, render_template
import pickle
import json

In [2]: salary_data = pd.read_csv("/Users/bhargavaramarajudandu/Desktop/Data Glacier Virtual Internship/model deployment using
X = salary_data.iloc[:, :-1].values
y = salary_data.iloc[:, 1].values

In [3]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.33, random_state = 0)

In [4]: regressor = LinearRegression()
regressor.fit(X_train, y_train)
y_pred = regressor.predict(X_test)
print(y_pred)

[ 40835.10590871 123079.39940819  65134.55626083  63265.36777221
 115602.64545369 108125.8914992  116537.23969801  64199.96201652
 76349.68719258 100649.1375447 ]
```

### Step 2: Save the trained model by using pickle library.



The image shows a Jupyter Notebook interface with the title "model deployment using flask". The notebook contains three code cells. The first cell is a comment "#Save the model in disk". The second cell uses pickle.dump to save the regressor object to a file named "model.pkl" in write-binary mode. The third cell loads the model back from "model.pkl" in read-binary mode and prints the prediction for a new input [1.8].

```
In [5]: #Save the model in disk

In [6]: pickle.dump(regressor, open('model.pkl', 'wb'))

In [7]: model = pickle.load(open('model.pkl', 'rb'))
print(model.predict([[1.8]]))

[43638.88864165]
```

### Step 3: Deployment of the model

```
In [8]: #Deployment of Model

In [9]: app = Flask(__name__)
        model = pickle.load(open('model.pkl', 'rb'))

In [10]: @app.route("/")
        def home():
            return render_template("index.html")

In [11]: @app.route('/', methods=['POST'])
        def predict():
            data = request.get_json(force=True)
            prediction = model.predict([np.array(data['exp'])])
            output = prediction[0]
            return jsonify(output)

In [12]: if __name__ == '__main__':
        app.run(port=5000)

* Serving Flask app "__main__" (lazy loading)
* 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)

In [ ]:
```

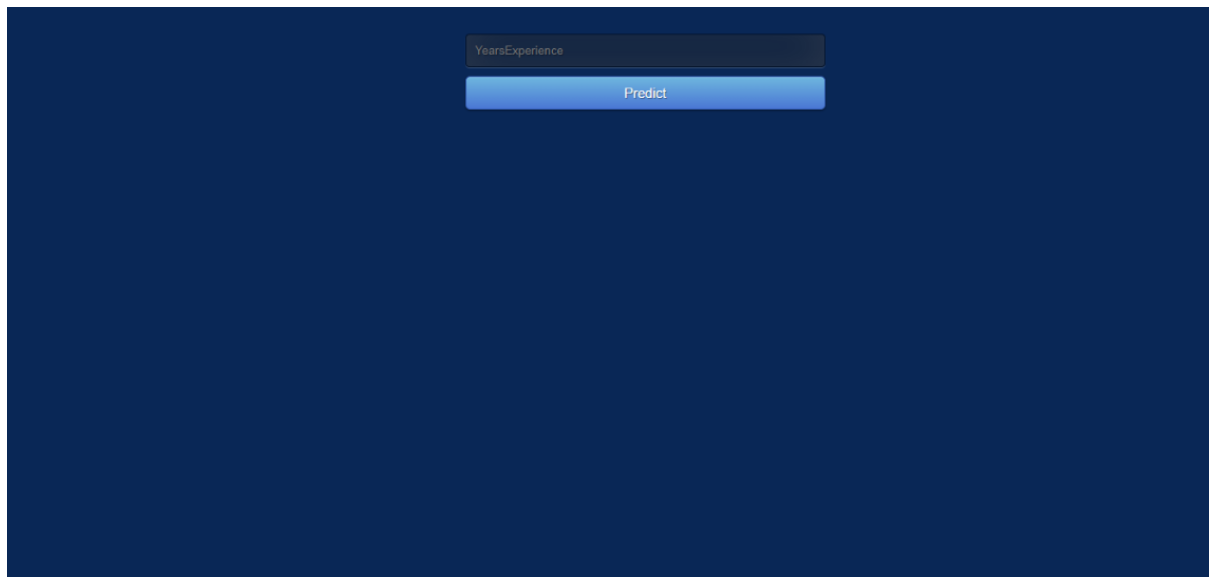
- Created the instance of the *Flask()* and loaded the model.
- Bounded “/” with the method *predict()* in which predict method gets the data from the json passed by the requestor.
- *model.predict()* method takes input from the json and converts it into 2D *numpy array* the results are stored into the variable named *output*.
- Return this variable after converting it into the json object using flask's *jsonify()* method.
- Run our server by following above code section and using port 5000.

### Step 4:

#### Checking python app.py file in terminal

```
model deployment using flask — python • python app.py — 80x24
Last login: Sat Jul 10 09:52:56 on ttys001
((base) bhargavaramarajudandu@Bhargavas-MacBook-Pro model deployment using flask) % python app.py
/Users/bhargavaramarajudandu/opt/anaconda3/lib/python3.8/site-packages/sklearn/base.py:310: UserWarning: Trying to unpickle estimator LogisticRegression from version 0.23.2 when using version 0.24.2. This might lead to breaking code or invalid results. Use at your own risk.
  warnings.warn(
* Serving Flask app "app" (lazy loading)
* Environment: production
  WARNING: This is a development server. Do not use it in a production deployment.
  Use a production WSGI server instead.
* Debug mode: on
* Running on http://127.0.0.1:5000/ (Press CTRL+C to quit)
* Restarting with fsevents reloader
/Users/bhargavaramarajudandu/opt/anaconda3/lib/python3.8/site-packages/sklearn/base.py:310: UserWarning: Trying to unpickle estimator LogisticRegression from version 0.23.2 when using version 0.24.2. This might lead to breaking code or invalid results. Use at your own risk.
  warnings.warn(
* Debugger is active!
* Debugger PIN: 150-074-756
```

## Step 5: Creating the web app using the using in browser



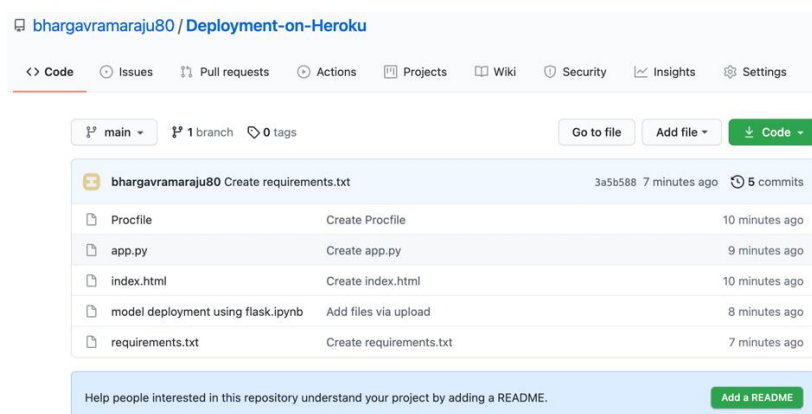
Step 6: Create Procfile which specifies the commands that are executed by the Heroku app on the startup. Web:unicorn app:app.

Running the command `pip freeze > requirements.txt` in CMD for creating requirement.txt file which will contain all of the dependencies of the flask app.

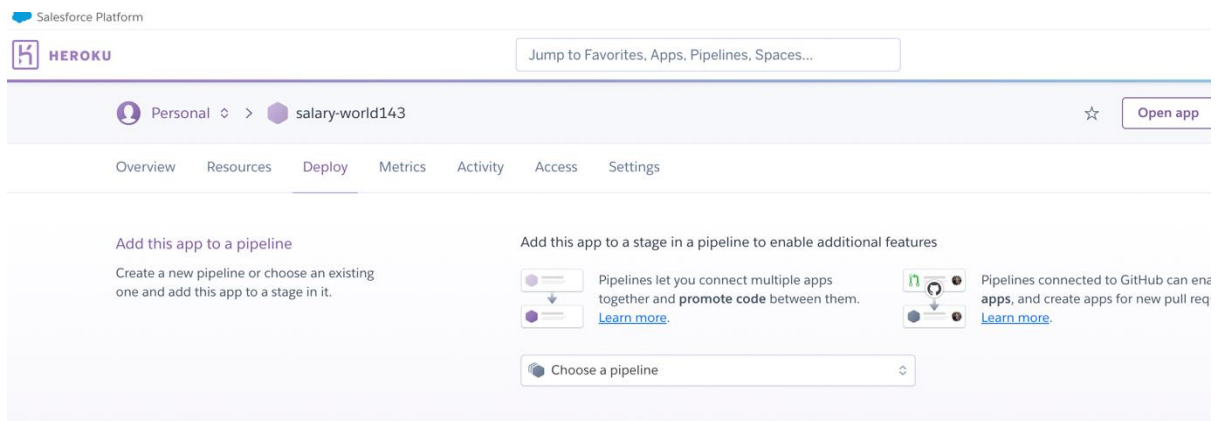
Step 7:

Create the repository in Github repository and commit the code.

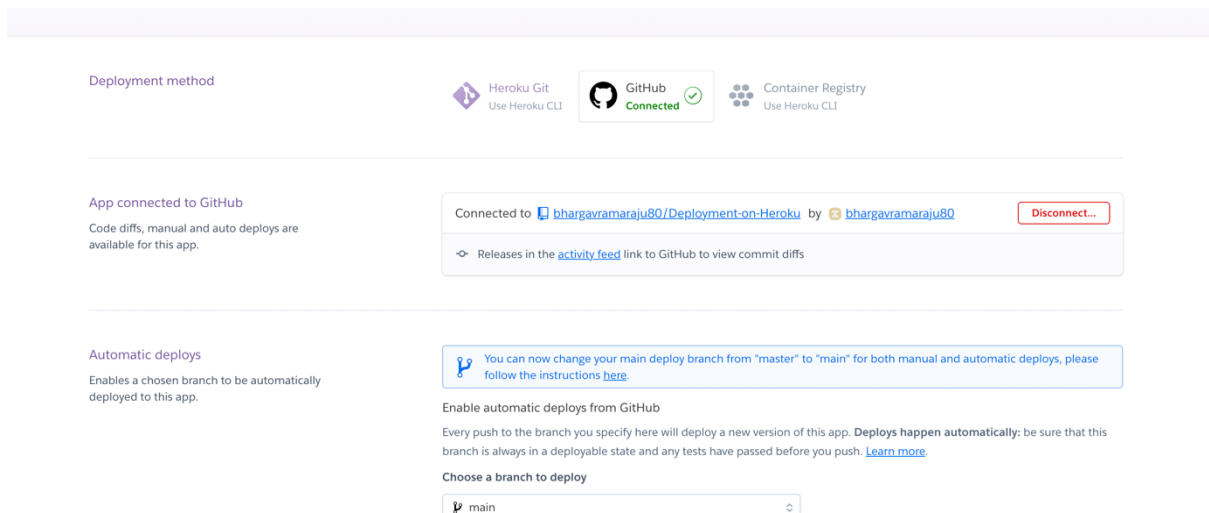
<https://github.com/bhargavramaraju80/Deployment-on-Heroku>



## Step 8: Create an account in Heroku and the create an app (Salary-world143).

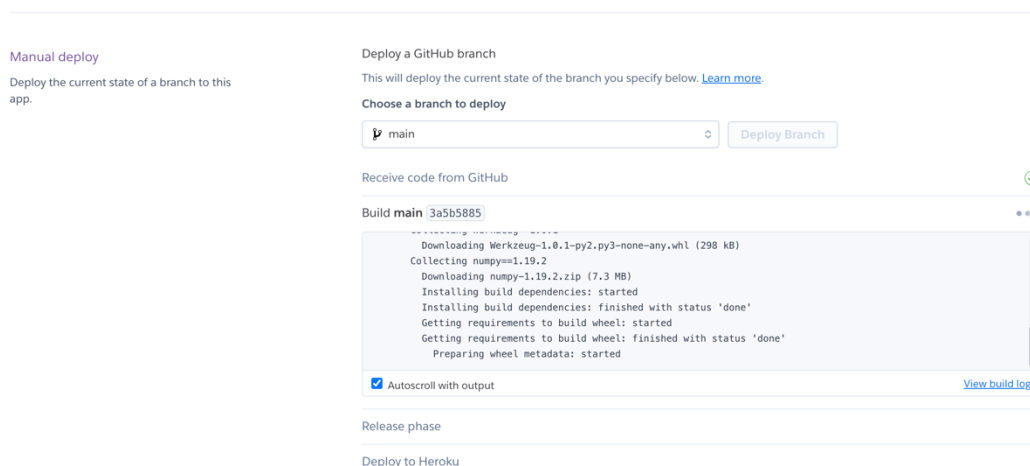


## Step 9: Link your Github account with your Heroku account and give access to your repository.



## Step 10:

## Finally Deploy the model on the Heroku app.



Step 11:

App Successfully deployed.

## App successfully deployed

[ML API \(experience-salary.herokuapp.com\)](https://experience-salary.herokuapp.com/)

