

VIRTUAL DATA INTERNSHIP

NAME: ONIFADE MICHAEL

BATCH CODE: LISUM32

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SUBMITTED TO: DATA GLACIER

Introduction

This project aims at the deployment of a machine learning model for data that expresses sales according to the type of advertisement and the size of its associated cost using a Flask framework. The proposed workflow is shown in diagram 1.1 while the data information is showcased in table 1.1

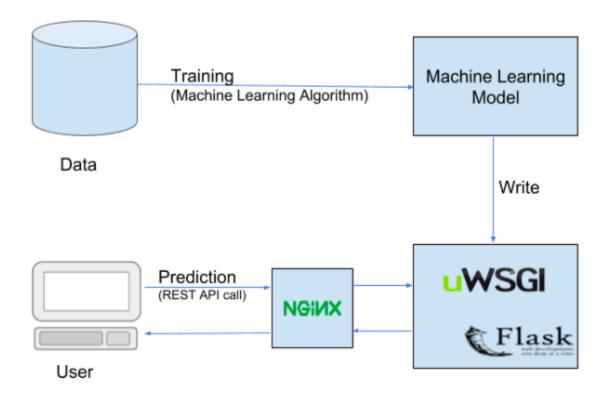


Diagram 1.1 Application Workflow

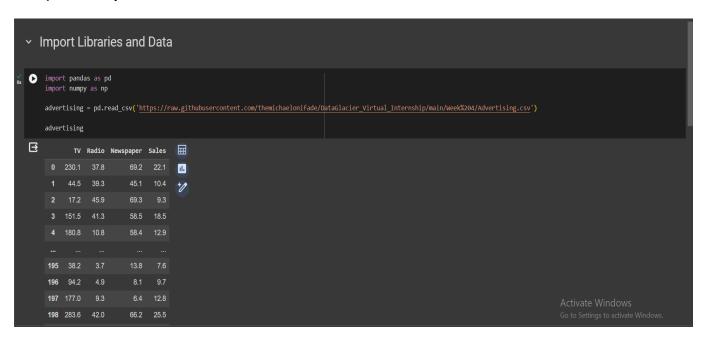
Table Data Details

Total number of observations	200
Total number of files	1
Total number of features	4
Base format of the file	Csv
Size of the data	4KB

Table 1.1 Dataset Information

The process for the deployment is summarized as follows

1) Data import



2) Model Building



```
Linear Regression

The world import Linear model import Linear Regression

model = Linear Regression()

model.fit(X_train, y_train)

Linear Regression |

Linear Regression()

Save Model

[4] import pickle pickle.dump(model.pkl', 'wb'))
```

3) Turning Model into web application using flask

```
from flask import Flask, request, render_template, jsonify
     import numpy as np
import pickle
     # Load the trained model
     model = pickle.load(open('model.pkl', 'rb'))
     app = Flask(__name__)
10
     @app.route('/')
     def index():
     return render_template('index.html')
13
     @app.route('/predict', methods=['POST'])
def predict_sales():
15
16
         data = request.get_json()
18
         tv = float(data['tv'])
radio = float(data['radio'])
19
20
21
         newspaper = float(data['newspaper'])
         prediction = model.predict(np.array([[tv, radio, newspaper]]))
25
         return jsonify({'projected sales revenue': prediction[0]})
26
27
     if __name__ == '__main__':
28
     app.run(debug=True)
```

Final output

Sales Revenue Generator
TV:
Radio:
Newpaper:
Sales Revenue