Machine Learning on Kubernetes

Creating and uploading necessary files in GCP- Cloud Shell Terminal

1. Start minikube in Google Cloud Platform

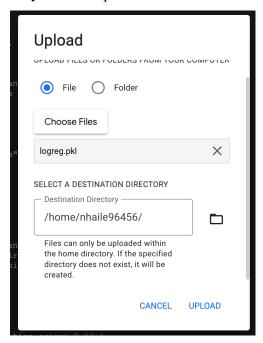
minikube start

```
nhaile96456@cloudshell:~ (cs570-big-data-analytics) $ minikube start
* minikube v1.33.1 on Ubuntu 22.04 (amd64)
  - MINIKUBE_FORCE_SYSTEMD=true
  - MINIKUBE_HOME=/google/minikube
- MINIKUBE_WANTUPDATENOTIFICATION=false
* Using the docker driver based on existing profile
* Starting "minikube" primary control-plane node in "minikube" cluster
* Pulling base image v0.0.44 ...
* Updating the running docker "minikube" container ...
* Preparing Kubernetes v1.30.0 on Docker 26.1.1 ...
   - kubelet.cgroups-per-qos=false
  - kubelet.enforce-node-allocatable=""
* Verifying Kubernetes components...
  - Using image gcr.io/k8s-minikube/storage-provisioner:v5
* Enabled addons: storage-provisioner, default-storageclass
* Done! kubectl is now configured to use "minikube" cluster and "default" namespace by default
nhaile96456@cloudshell:~ (cs570-big-data-analytics)$
```

2. Create a requirements.txt file.

```
nhaile96456@cloudshell:~ (cs570-big-data-analytics)$ vi requirements.txt
nhaile96456@cloudshell:~ (cs570-big-data-analytics)$ cat requirements.txt
Flask==1.1.1
gunicorn==19.9.0
itsdangerous==1.1.0
Jinja2==2.10.1
MarkupSafe==1.1.1
Werkzeug==0.15.5
numpy==1.19.5 # Adjusted to a version before np.float deprecation scipy>=0.15.1
scikit-learn==0.24.2 # Ensure compatibility with numpy version matplotlib>=1.4.3
pandas>=0.19
flasgger==0.9.4
nhaile96456@cloudshell:~ (cs570-big-data-analytics)$
```

3. Upload logreg.pkl file to your workspace.



4. Create a flask_api.py file and paste the below python code in it.

```
from flask import Flask, request
import numpy as np
import pickle
import pandas as pd
from flasgger import Swagger
app = Flask(__name__)
Swagger(app)
# Load the logistic regression model
pickle_in = open("logreg.pkl", "rb")
model = pickle.load(pickle_in)
@app.route('/')
def home():
 return "Welcome to the Flask API!"
@app.route('/predict', methods=["GET"])
def predict_class():
 Predict if Customer would buy the product or not.
 parameters:
  - name: age
   in: query
   type: number
   required: true
  - name: new_user
   in: query
```

```
type: number
   required: true
  - name: total_pages_visited
   in: query
   type: number
   required: true
 responses:
  200:
   description: Prediction
 age = int(request.args.get("age"))
 new_user = int(request.args.get("new_user"))
 total_pages_visited = int(request.args.get("total_pages_visited"))
 prediction = model.predict([[age, new_user, total_pages_visited]])
 return "Model prediction is " + str(prediction[0])
@app.route('/predict_file', methods=["POST"])
def prediction_test_file():
 Prediction on multiple input test file.
 parameters:
  - name: file
   in: formData
   type: file
   required: true
 responses:
  200:
   description: Test file Prediction
 df_test = pd.read_csv(request.files.get("file"))
 prediction = model.predict(df_test)
 return str(list(prediction))
if __name__ == '__main__':
 app.run(debug=True, host='0.0.0.0', port=5000)
```

```
nhaile96456@cloudshell:~ (cs570-big-data-analytics) $ vi flask_api.py
nhaile96456@cloudshell:~ (cs570-big-data-analytics) $ cat flask_api.py
from flask import Flask, request
import numpy as np
import pickle
import pandas as pd
from flasgger import Swagger

app = Flask(__name__)
Swagger(app)

# Load the logistic regression model
pickle_in = open("logreg.pkl", "rb")
model = pickle.load(pickle_in)

@app.route('/')
def home():
    return "Welcome to the Flask API!"

@app.route('/predict', methods=["GET"])
def predict_class():
    """
```

5. Create a DockerFile with the following content in it.

```
# Use the official Python image from the Docker Hub
FROM python:3.8-slim

# Set the working directory in the container
WORKDIR /app

# Copy the current directory contents into the container at /app
COPY . /app

# Install any needed packages specified in requirements.txt
RUN pip install --no-cache-dir -r requirements.txt

# Make port 5000 available to the world outside this container
EXPOSE 5000

# Define environment variable to prevent Python from writing .pyc files to disk
ENV PYTHONUNBUFFERED=1

# Run flask_api.py when the container launches
CMD ["python", "flask_api.py"]
```

```
nhaile96456@cloudshell:~ (cs570-big-data-analytics)$ vi Dockerfile
nhaile96456@cloudshell:~ (cs570-big-data-analytics) $ cat Dockerfile
# Use the official Python image from the Docker Hub
FROM python:3.8-slim
# Set the working directory in the container
WORKDIR /app
# Copy the current directory contents into the container at /app
COPY . /app
# Install any needed packages specified in requirements.txt
RUN pip install --no-cache-dir -r requirements.txt
# Make port 5000 available to the world outside this container
EXPOSE 5000
# Define environment variable to prevent Python from writing .pyc files to disk
ENV PYTHONUNBUFFERED=1
# Run flask_api.py when the container launches
CMD ["python", "flask api.py"]
nhaile96456@cloudshell:~ (cs570-big-data-analytics)$
```

Explanation

FROM python:3.8-slim: This line specifies the base image for the Docker container, which is a lightweight version of Python 3.8.

Working Directory:

WORKDIR /app: This sets the working directory to /app in the container. All subsequent commands will be run from this directory. Copy Files:

COPY . /app: This copies the contents of the current directory on your host machine to the /app directory in the container.

Install Dependencies:

RUN pip install --no-cache-dir -r requirements.txt: This installs the dependencies listed in requirements.txt using pip. The --no-cache-dir option reduces the size of the image by not caching the downloaded packages.

Expose Port:

EXPOSE 5000: This makes port 5000 available for network connections. It does not actually publish the port; it simply serves as documentation. Environment Variable:

ENV PYTHONUNBUFFERED=1: This prevents Python from buffering stdout and stderr, ensuring that logs are immediately available in Docker. Command:

CMD ["python", "flask_api.py"]: This specifies the command to run the Flask application when the container starts.

6. To build the docker image use the command.

sudo docker build -t ml app docker.

7. The following command runs a Docker container from the ml app docker image.

docker container run -p 5000:5000 ml app docker

```
nhaile96455@cloudshell:~ (cs570-big-data-analytics) $ docker container run -p 5000:5000 ml_app_docker
/usr/local/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 "flask api" (lazy loading)
* Environment: production

WARNING: This is a development server. Do not use it in a production deployment.

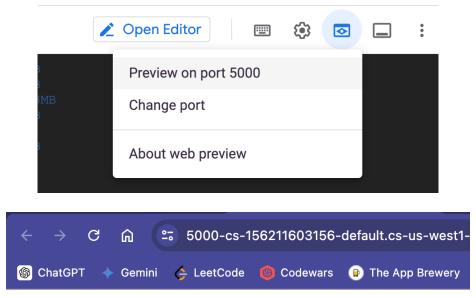
Use a production WSGI server instead.

* Debug mode: on
* Running on http://0.0.0.0:5000/ (Press CTRL+C to quit)
* Restarting with stat
/ usr/local/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: 250-833-593
```

8. On the top right side of the screen, click the little eye shaped button. Change the port number if it's not already 5000. You will expect to see some sort of a welcome message.

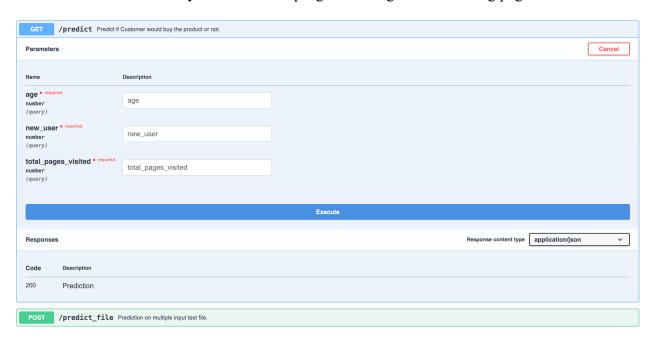


Welcome to the Flask API!

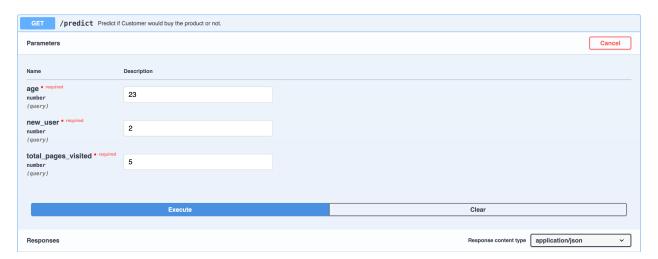
9. Add /apidocs/ at the end of the URL and you will see the home page of Swagger-UI.



10. Click 'Get' and 'Try it out' at the top right side to get the following page.

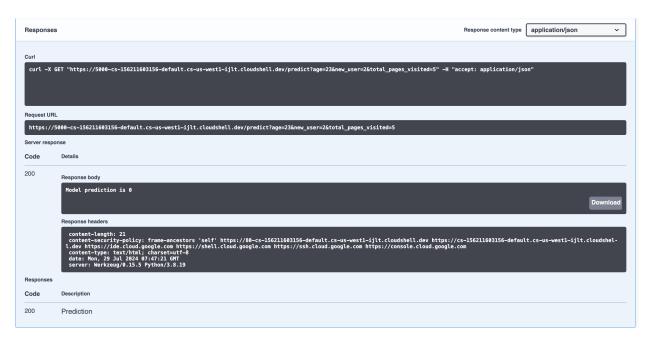


11. Enter values for the input parameters and click 'Execute.'



12. Upon the execution call, the request goes to the app and predictions are made by the model.

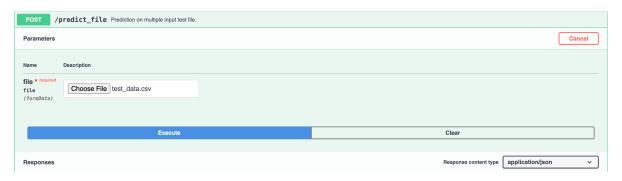
The result of the model prediction is displayed in the Prediction section of the page as following



13. Next, the app can make predictions for a group of customers (test data) by clicking 'Post'.



14. Upload the test_data.csv file and click 'Execute.'



15. The model would make the predictions, and the results would be displayed as shown below.



16. To list running docker containers to kill them, run the command docker ps.

17. Then run 'docker kill < container id>'.

```
nhaile96456@cloudshell:~ (cs570-big-data-analytics)$ docker kill 6393b5a57ff9 6393b5a57ff9 nhaile96456@cloudshell:~ (cs570-big-data-analytics)$
```