CS501- Week 10 Homework 1: Machine Learning on Kubernetes

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https://hc.labnet.sfbu.edu/~henry/sfbu/course/cloud_computing/genai/slide/exercise_kubernetes. html

Q2 ===> Machine Learning on Kubernetes

Machine Learning on Kubernetes

Creating and uploading necessary files in GCP- Cloud Shell Terminal

1. Start minikube in Google Cloud Platform

```
fba8584@cloudshell:~ (sfbu-ce571-414319)$ minikube start

* minikube v1.32.0 on Debian 11.9 (amd64)

- MINIKUBE FORCE SYSTEMD=true

- MINIKUBE FORCE SYSTEMD=true

- MINIKUBE HOME—/google/minikube

- MINIKUBE MONTE-MENOTIFICATION—false

* Using the docker driver based on existing profile

* Starting control plane node minikube in cluster minikube

* Pulling base image ...

* Updating the running docker "minikube" container ...

X Docker is nearly out of disk space, which may cause deployments to fail! (95% of capacity). You can pass '--force' to skip this check.

* Suggestion:

Try one or more of the following to free up space on the device:

1. Run "docker system prune" to remove unused Docker data (opticnally with "-a")

2. Increase the storage allocated to Docker for Desktop by clicking on:
    Docker icon > Preferences > Resources > Disk Image Size

3. Run "minikube ssh -- docker system prune" if using the Docker container runtime

* Related issue: https://github.com/kubernetes/minikube/issues/9024

* Preparing Kubernetes v1.28.3 on Docker 24.0.7 ...

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- Westfying Kubernetes components.

- Using image ggr. io/K8s-minikube/storage-provisionervs

- Bushele addons: storage-provisioner, default-storageclass

- Done! Kubectl is now configured to use "minikube" cluster and "default" namespace by default
```

- 2. Create requirements.txt file using the following command
 - nano requirements.txt

```
fba8584@cloudshell:~ (sfbu-cs571-414319) $ nano requirements.txt
```

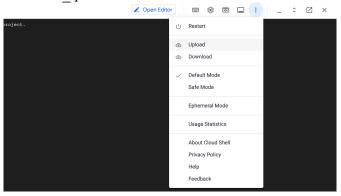
Then enter the following contents

```
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
```

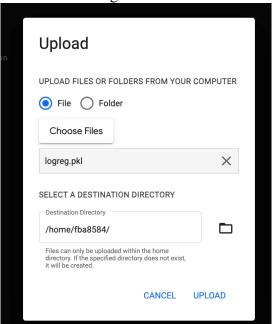
```
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Flask==1.1.1
gunicorn==19.00
itsdangerous==1.1.0
Jinja2==2.10.1
MarkupSafe==1.1.1
Werkzeug==0.15.5
numpy==1.19.5  # Adjusted to a version before np.float deprecation scipy>=0.15.1
scikit-learn==0.24.2  # Ensure compatibility with numpy version matplotlib>=1.4.3
pandas>=0.19
flasgger==0.9.4
```

3. Upload logreg.pkl file by clicking the three dots in the top-right part of the Cloud Shell Terminal and then choose upload



Then upload the logreg.pkl file as following



- 4. Create flask api.py file using the command
 - nano flask api.py

fba8584@cloudshell:~ (sfbu-cs571-414319) \$ nano flask api.py

Then enter the following contents

```
# -*- coding: utf-8 -*-
Created on Mon May 25 12:50:04 2020
@author: pramod.singh
from flask import Flask, request
import numpy as np
import pickle
import pandas as pd
from flasgger import Swagger
app = Flask( name )
Swagger(app)
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)
@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)
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   GNU nano 5.4
 Created on Mon May 25 12:50:04 2020
 @author: pramod.singh
 from flask import Flask, request
 import numpy as np
 import pickle
 import pandas as pd
 from flasgger import Swagger
 app = Flask(__name__)
 Swagger (app)
 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

```
GNU nano 5.4
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)
@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
```

```
@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))
@app.route('/apidocs')
def api docs():
    return "API Documentation goes here."
            == '__main__':
    name
    app.run (debug=True, host='0.0.0.0', port=5000)
```

Step 4: Dockerfile

- 1. Create Dockerfile using command
 - nano Dockerfile

```
fba8584@cloudshell:~ (sfbu-cs571-414319) $ nano Dockerfile
```

Then enter the following content

FROM python:3.8-slim WORKDIR /app COPY . /app EXPOSE 5000 RUN pip install -r requirements.txt CMD ["python", "flask api.py"]



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Terminal

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```
GNU nano 5.4

FROM python:3.8-slim

WORKDIR /app

COPY . /app

EXPOSE 5000

RUN pip install -r requirements.txt

CMD ["python", "flask_api.py"]
```

- 1. 'FROM python:3.8-slim'
- This line sets the base image for the Docker image you are creating. It tells Docker to start with the 'python:3.8-slim' image, which is an official Python image with Python 3.8 installed on it. The 'slim' version is a smaller version of the image that has fewer packages pre-installed, making the image size smaller.
- 2. 'WORKDIR /app'
- This instruction sets the working directory within the Docker container to */app*. All subsequent commands will be executed in this directory within the container.
- 3. 'COPY . /app'
- This line copies everything from the current directory (on the host machine where you're running the Docker build command, indicated by the first**) into the */app directory inside the Docker image (the second '/app*).
- 4. 'EXPOSE 5000'
- The 'EXPOSE' instruction informs Docker that the container listens on the specified network port at runtime. In this case, it tells Docker that the container will listen on port 5000. It's worth noting that this does not actually publish the port—it serves as documentation and is used by the 'docker run -p' command to map the container port to a port on the Docker host.
- 5. 'RUN pip install -r requirements.txt*
- This command tells Docker to run pip install' inside the container, which will install the Python dependencies listed in the requirements.tt file. These dependencies are necessary for the Flask application to run correctly. 6. CMD ["python", "flask api.py"]'
- This is the command that will be executed by default when the Docker container starts. In this case, it's telling Docker to run 'flask_api.py using Python. This is the Flask application you want to run inside the container.

Step 5: Running the Docker Container

- 1. To build the docker image use the command
 - sudo docker build -t ml app docker.

```
| Carbon | C
```

- 2. This command runs a Docker container from the ml app docker image:
 - docker container run -p 5000:5000 ml app docker

```
fba85848cloudshell:- (sfbu-cs571-414319) & docker container run -p 5000:5000 ml_app_docker

* Serving Flask app "flask apt" (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

Varifocal/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 le

ad to breaking code or invalid results. Use at your own risk.

Warnings.Warn(

* Running on http://0.0.0.0:5000/ (Press CTR1+C to quit)

* Restarting with stat

Vusr/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 le

ad to breaking code or invalid results. Use at your own risk.

Warnings.Warn(

* Debugger is active!

* Debugger 18: 297-165-261
```

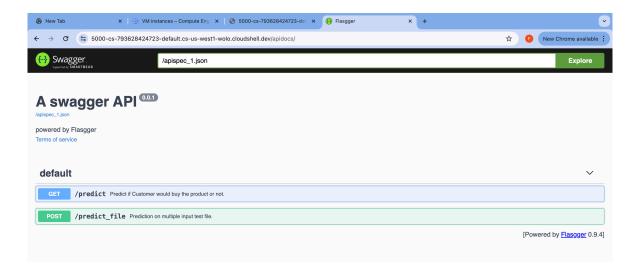
3. In the right-upper side of the terminal click the eye shaped button and then click *Preview on port 5000*. Change port if it is not 5000 by default.



4. You will see this using the web preview.

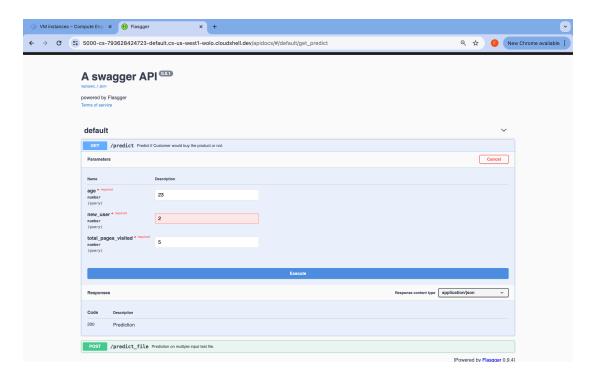


- 5. Add /apidocs/ at the end of the link to access the running ml- app as following
 - There are two tabs GET and POST.

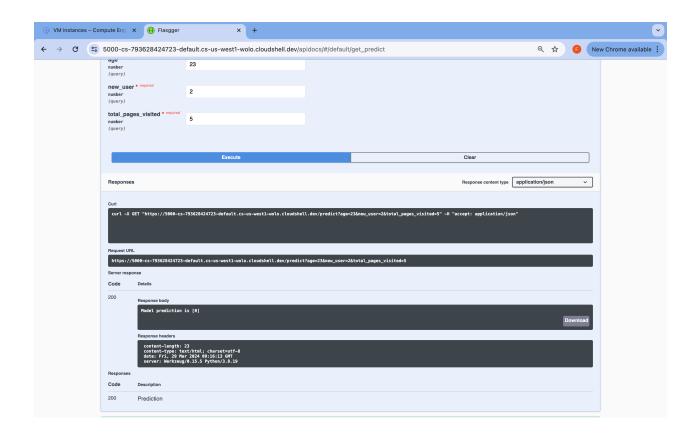




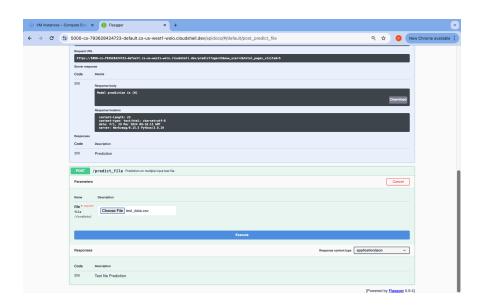
7. Fill values for the input parameters and then click Execute.



- 8. 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

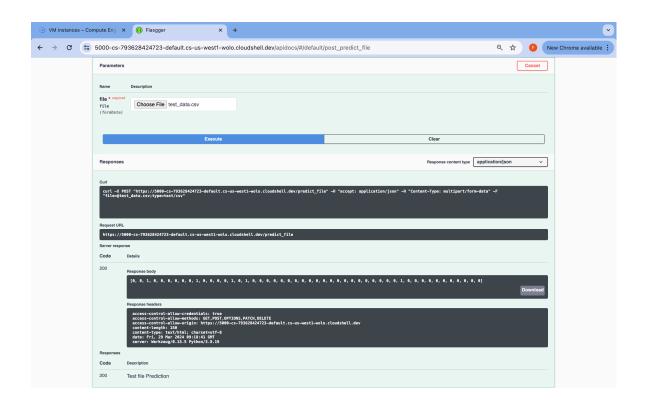


9. The next prediction that can be done is for a group of customers (test data) via a post request.



10. Upload the test data file containing the same parameters in a similar order.

The model would make the prediction, and the results would be displayed upon execute as following.



Step 6: Stopping/killing the running container

1. Use docker ps to list running Docker containers



- The CONTAINER ID is given as 367119b87a37
- 2. Use the command
 - docker kill < CONTAINER ID > to kill the running container as follows.

```
fba8584@cloudshell:~ (sfbu-cs571-414319)$ docker kill 367119b87a37
367119b87a37
fba8584@cloudshell:~ (sfbu-cs571-414319)$
```