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

```
shadgu24995@cloudshell:~ (airy-digit-437101-s8)$ kubectl get nodes
                                         STATUS
                                                   ROLES
NAME
                                                             AGE VERSION
                                                   <none>
                                                             2m14s v1.30.5-gke.1443001
gke-kubia-default-pool-021f4515-cpk8
                                         Ready
gke-kubia-default-pool-3fc93e1d-3j53
gke-kubia-default-pool-adb8682a-8jdc
                                                            2m14s
2m15s
                                         Ready
                                                   <none>
                                                                     v1.30.5-gke.1443001
                                         Ready
                                                   <none>
                                                                     v1.30.5-gke.1443001
shadgu24995@cloudshell:~ (airy-digit-437101-s8)$
```

Start minikube in Google Cloud Platform

```
shadgu24995@cloudshell:~ (airy-digit-437101-s8)$ minikube start

* minikube v1.34.0 on Ubuntu 24.04 (amd64)

- MINIKUBE_FORCE_SYSTEMD=true

- MINIKUBE_HOME=FORCE_SYSTEMD=true

- MINIKUBE_HOME=FORCE_SYSTEMD=true

- MINIKUBE_WANTUPDATENOTIFICATION=false

* Automatically selected the docker driver. Other choices: none, ssh

* Using Docker driver with root privileges

* Starting "minikube" primary control-plane node in "minikube" cluster

* Pulling base image v0.0.45 ...

* Downloading Kubernetes v1.31.0 preload ...

> preloaded-images-k8s-v1-v1..: 326.69 MiB / 326.69 MiB 100.00% 203.59

> gcr.io/k8s-minikube/kicbase..: 487.90 MiB / 487.90 MiB 100.00% 80.11 M

* Creating docker container (CFUS=2, Memory=4000MB) ...

* Preparing Kubernetes v1.31.0 on Docker 27.2.0 ...

- kubelet.cgroups-per-gos=false

- kubelet.enforce-node-allocatable=""

- Generating certificates and keys ...

- Booting up control plane ...

* Configuring RBAC rules ...

* Configuring BAC rules ...

* Configuring RbAC rules ...

* 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 "minikubbe" cluster and "default" namespace by default

shadgu24995@cloudshell:~ (airy-digit-437101-s8)$
```

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

shadgu24995@cloudshell:~ (airy-digit-437101-s8)\$ nano requirements.txt

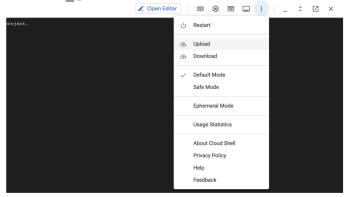
Then enter the following contents

```
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
```

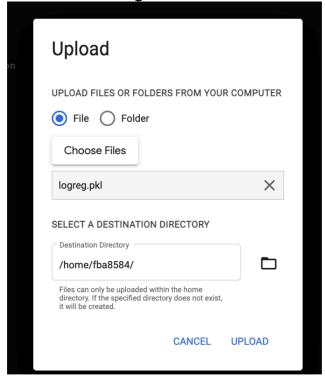
```
CLOUD SHELL
Terminal (sfbu-cs571-414319) × (sfbu-cs571-414319) × + 

GNU nano 5.4
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
```

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

shadgu24995@cloudshell:~ (airy-digit-437101-s8)\$ nano flask_api.py

```
GNU nano 5.4
# -*- coding: utf-8 -*-
Created on Mon May 25 12:50:04 2020
@author: pramod.singh
11 11 11
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."
if name == ' main ':
    app.run(debug=True, host='0.0.0.0', port=5000)
```

Step 4: Dockerfile 1. Create Dockerfile using command - nano Dockerfile

```
shadgu24995@cloudshell:~ (airy-digit-437101-s8)$ docker build -t ml_app_docker
```

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"]

```
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 .

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

```
shadgu24995@cs-342231923890-default:-/Python-3.10.128 docker run -p 5000:5000 ml_app_docker

* Serving Flask app "flask_pap!" (lazy loading)

* Environment: production
WARNING: This is a development server. Do not use it in a production deployment.
Use a production WSGO server instead.

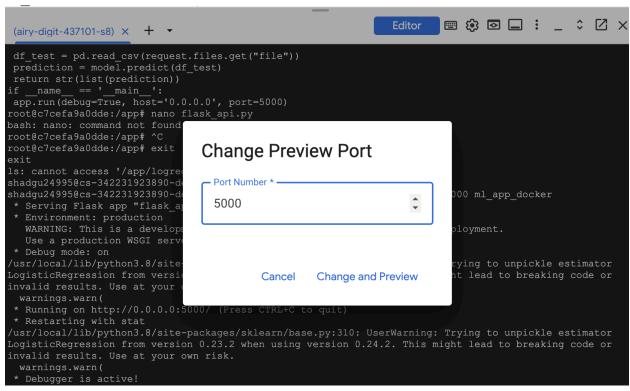
* Debug mode: on
/usr/loac1/lib/python3.8/site-packages/sklearn/base.py:310: UserWarning: Trying to unpickle estimator LogisticRegression from version 0.23.2 when using version 0.24.2. This might lead to breaking code or invalid results. Use at your own risk.

* Running on http://0.0.0.0:55000/ (Press CTRL+C to quit)

* Restarting with stat
/usr/loca1/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 tlead to breaking code or invalid results. Use at your own risk.

warnings.warn(
* Varnings.warn(
```

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.

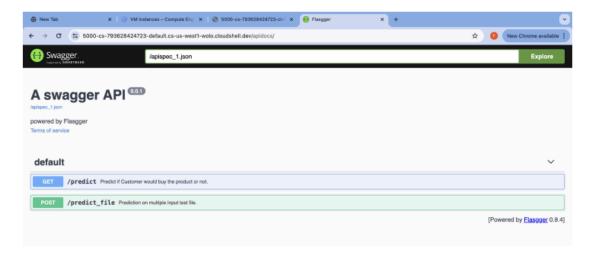


Welcome to the Flask API!

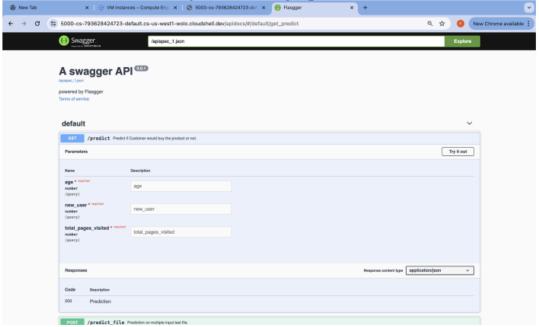
4. You will see this using the web preview.



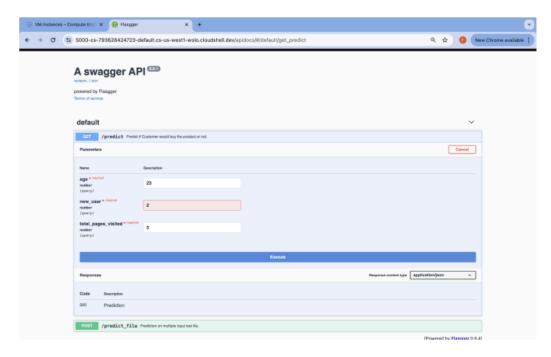
- 5. Add/apidocs/ at the end of the link to access the running ml- app as following
 - There are two tabs GET and POST.



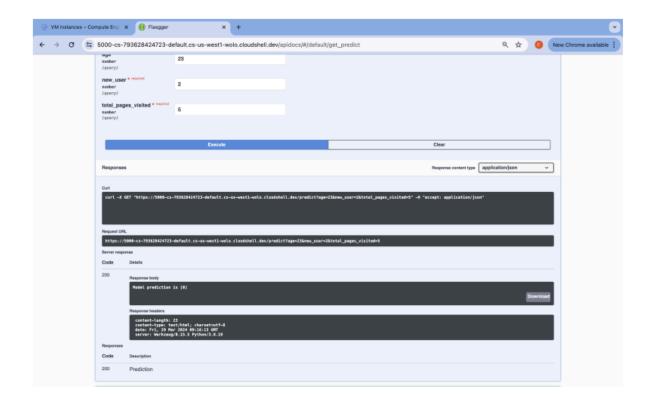
6. Click GET and then click Try it out in the top-right corner of the GET box.



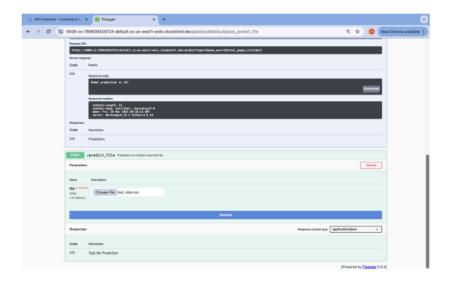
7. Fill values for the input parameters and then click Execute.



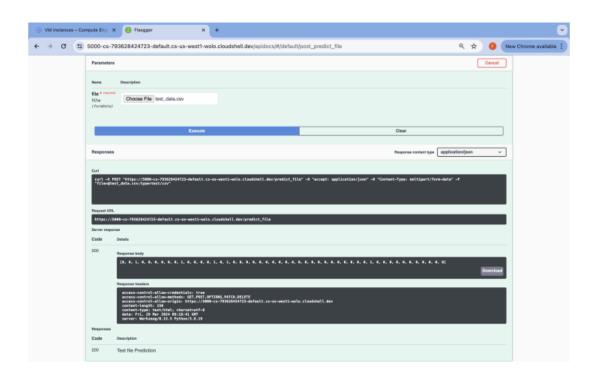
- 8. Upon the execution call, the request goes to the app, and predictions are made by the model.
 - The result of the model prediction is displayed in the Prediction section of the page as following



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



Updating Portfolio- GitHub link

https://github.com/Ghanitu/CloudComputing.git