Question:

- 1. Export the metrics (like request per second, memory usage, cpu usage etc) in the existing mini project given to Interns
- 2. Install Prometheus and Grafana using Docker (with docker-compose)
- 3. Configure prometheus (scrape configs) such way that it can scrape the metrics from default metric path of the application job
- 4. Validate the entire configuration to check if the data is coming or not in Prometheus UI
- 5. Create the Dashboards in Grafana on top of the metrics exported by adding the Prometheus as a Datasource.

Solution:

Step 1:

Application overview:

I created an ML-based Flask application that collects a comment from the user and detects whether the comment is toxic. If it is toxic, the application classifies it as severe toxicity, insult, threat, or other categories using logistic regression. I also integrated MLflow for experiment tracking, and my model achieved an accuracy of 93%.

```
from flask import Flask, render_template, request
import joblib
import pandas as pd
from sklearn.feature_extraction.text import TfidfVectorizer
import prometheus_client # Import Prometheus client for metrics
# import mlflow
app = Flask(__name__)

# model_uri = "models:/toxic_comment_model/1"
# model = mlflow.pyfunc.load model(model uri)
```

```
# Load model and vectorizer
model = joblib.load("toxic comment model.pkl")
tfidf = joblib.load("tfidf vectorizer.pkl")
# Initialize Prometheus metrics
REQUEST COUNT = prometheus client.Counter('request count',
                                                                 'Total
number of requests')
COMMENT COUNT = prometheus client.Counter('comment count', 'Total
number of comments submitted')
@app.route('/')
def home():
  return render template('index.html')
@app.route('/predict', methods=['POST'])
def predict():
  if request.method == 'POST':
      comment = request.form['comment']
       # Increment comment count
      COMMENT COUNT.inc()
       # Increment request count
      REQUEST COUNT.inc()
      comment tfidf = tfidf.transform([comment])
      prediction = model.predict(comment tfidf).tolist()[0]
          categories = ["Toxic", "Severe Toxic", "Obscene", "Threat",
"Insult", "Identity Hate"]
             result = {categories[i]: bool(prediction[i]) for i in
range(len(categories))}
              return render template('result.html', comment=comment,
result=result)
@app.route('/metrics')
def metrics():
  return prometheus_client.generate_latest()
```

<pre>ifname == 'main':</pre>	
app.run(host='0.0.0.0', port=5001, debug=True)	
Toxic Comment Classifier Enter your comment: Classify	
Comment Classification Result Comment: i hate you Classification:	
Toxic: Yes	
Severe Toxic: No	
Obscene: No	
Threat: No	
Insult: No	
Identity Hate: No Classify another comment	

Below two collect number comment and number of request

```
REQUEST_COUNT = prometheus_client.Counter('request_count', 'Total
number of requests')

COMMENT_COUNT = prometheus_client.Counter('comment_count', 'Total
number of comments submitted')
```

And then i created docker file which contain

```
FROM python:3.9-slim
WORKDIR /app
COPY . /app
RUN pip install --no-cache-dir -r requirements.txt
EXPOSE 5001
CMD ["python", "app.py"]
```

My application's Docker image

apiVersion: v1
kind: Service

vikaskarbail/toxic-comment-classification. Below are the Kubernetes deployment and service files for the application, where I exposed the service using NodePort.

```
apiVersion: apps/v1
kind: Deployment
metadata:
name: toxic-comment-classification
labels:
   app: toxic-comment-classification
spec:
replicas: 1
selector:
  matchLabels:
     app: toxic-comment-classification
template:
  metadata:
     labels:
       app: toxic-comment-classification
   spec:
     containers:
     - name: toxic-comment-classification
       image: vikaskarbail/toxic-comment-classification:latest
       - containerPort: 5001
Service.yaml
```

```
metadata:
   name: toxic-comment-classification-service
spec:
   type: NodePort
   selector:
       app: toxic-comment-classification
   ports:
       - port: 5001
        targetPort: 5001
        nodePort: 30000

# to deploy the deployment file
            kubectl apply -f todo-deployment.yaml
# to deploy the service file
            kubectl apply -f todo-service.yaml
```

Step 2: Helm Chart

To install Prometheus and Grafana in your Kubernetes cluster, use Helm. First, add the necessary Helm repositories and update them. Then, install Prometheus and Grafana using the following commands:

```
helm repo add prometheus-community
https://prometheus-community.github.io/helm-charts
helm repo add grafana https://grafana.github.io/helm-charts
helm repo update

# Install Prometheus
helm install prometheus prometheus-community/prometheus

# Install Grafana
helm install grafana grafana/grafana
```

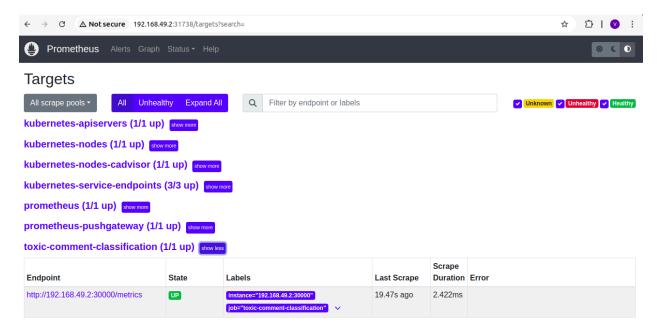
Step 3: Configure Prometheus Scrape Configs

We can customize the scrape configuration by editing the values.yaml of the Prometheus Helm release or overriding it during installation.

Step 4: Accessing and Configuring Prometheus and Grafana in Minikube

To access the Prometheus UI, Grafana UI, and Toxic Comment Application UI, use the Minikube IP along with their respective NodePorts.

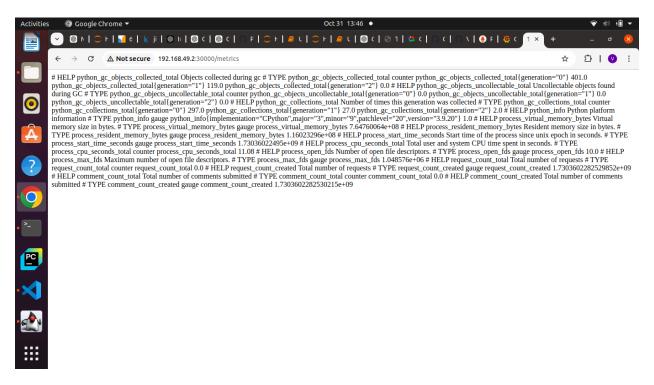
To add Prometheus as a data source in the Grafana UI, navigate to **Configuration > Data Sources**, click **Add data source**, and select **Prometheus**. Set the URL to http://prometheus-server



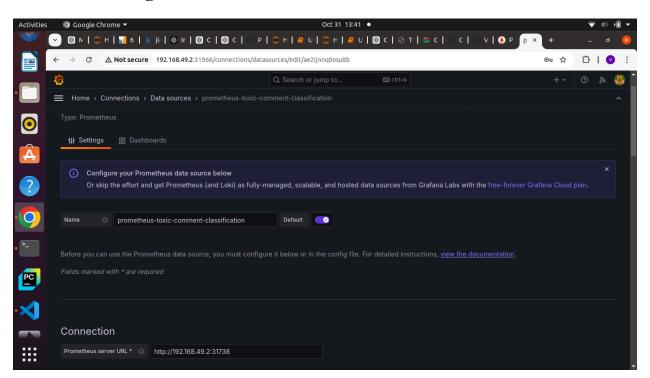
Step 5: Create Dashboards in Grafana

To create dashboards in Grafana, navigate to **Create > Dashboard**. From there, add panels to visualize key metrics such as *Total HTTP Requests*, *Memory Usage*, and *CPU Usage*.

> Metrics from the application



> Grafana Setting



> Dashboard: Cpu usage, Memory Usage, Number of comments

