#### Zakaria CHOUKRI

### Lab Flask for ML - Cloud Computing - M311

#### Installing Flask

```
> python3 -m venv flask, env
> python3 -m venv flask, env
> source flask, env/bin/activate
> pip install flask
Collecting flask
Downloading flask-3.1.2-py3-none-any.whl.metadata (3.2 kB)
Collecting flask
Downloading blinker>1.9.0 (from flask)
Downloading blinker>1.9.0-py3-none-any.whl.metadata (1.6 kB)
Collecting itsdangerous>2.2.0-py3-none-any.whl.metadata (2.6 kB)
Collecting itsdangerous>2.2.0-py3-none-any.whl.metadata (2.9 kB)
Collecting inja2>3.1.3 (from flask)
Using cached clink=8.3.3-py3-none-any.whl.metadata (2.9 kB)
Collecting markupsafe=3.0.3-cp312-cp312-manylinux2014_x86_64.manylinux_2_17_x86_64.manylinux_2_28_x86_64.whl.metadata (2.7 kB)
Collecting markupsafe=3.0.3-cp312-cp312-manylinux2014_x86_64.manylinux_2_17_x86_64.manylinux_2_28_x86_64.whl.metadata (2.7 kB)
Collecting markupsafe=3.0.3-cp312-cp312-manylinux2014_x86_64.manylinux_2_17_x86_64.manylinux_2_28_x86_64.whl.metadata (2.7 kB)
Comnloading blinker-1.9.0-py3-none-any.whl (107 kB)
Downloading blinker-1.9.0-py3-none-any.whl (107 kB)
Downloading blinker-1.9.0-py3-none-any.whl (107 kB)
Downloading markupsafe=3.0.3-cp312-cp312-manylinux2014_x86_64.manylinux_2_17_x86_64.manylinux_2_28_x86_64.whl (22 kB)
Downloading werkzeug=3.1.3-py3-none-any.whl (234 kB)
Downloading werkzeug=3.1.3-py3-none-any.whl (224 kB)

Tastalling collected packages: markupsafe, itsdangerous, click, blinker, werkzeug, jinja2, flask
Successfully installed blinker-1.9.0-click=8.3.0 flask=3.1.2 itsdangerous=2.2.0 jinja2-3.1.6 markupsafe-3.0.3 werkzeug-3.1.3

**Clink** form**

**Clink** form**

**Clink** form**

**Clink** form**

**Clink** form**

**A this definition**

**Clink** form**

**A this definition**

**A this
```

#### Example 1:

```
hello.py ×
hello.py > ...

from flask import Flask

app = Flask(__name__)

d

app.route('/')
def hello_world():
    --- return 'Hello, World!'

finame__ = '__main__':
    --- app.run()

11
```

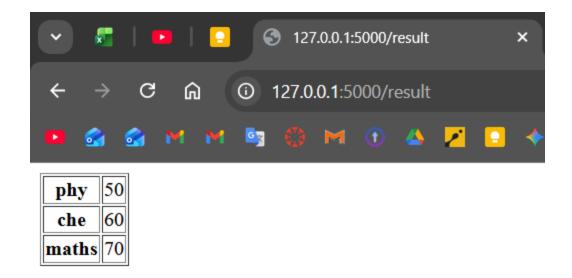


Hello, World!

Example 2:

```
app.py x
app.py >...
1 from flask import Flask, render_template
2
3 app = Flask(__name__)
4
5 @app.route('/result')
6 def result():
7 | ...dict = {'phy':50, 'che':60, 'maths':70}
8 | ...return render_template('result.html', result=dict)
9
10 if __name__ · = ·'__main__':
11 | ...app.run(debug=True)
```

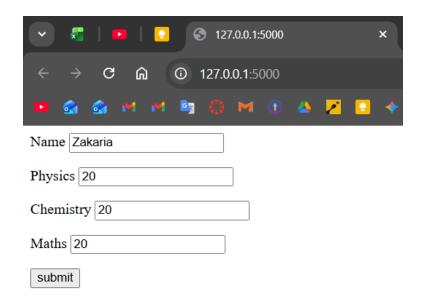
```
🥐 арр.ру
templates > 🥫 results.html > ...
 1 <!doctype html>
   <html>
 3 ----<body>
 4 
   {% for key, value in result.items() %}
 6 ····
    {{ key }} 
  10 {% endfor %}
   ·····
   </body>
   </html>
 14
```

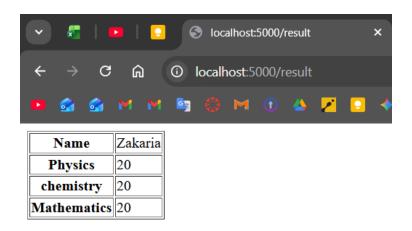


#### Example 3:

```
🥏 арр.ру
🥏 арр.ру > ...
      from flask import Flask, render_template, request
      app = Flask(__name__)
     @app.route('/')
      def student():
     return render_template('student.html')
      @app.route('/result', methods=['POST', 'GET'])
      def result():
     if request.method = 'POST':
 11
 12
      result = request.form
      return render_template('result.html', result=result)
 13
      if __name__ = '__main__':
      app.run(debug=True)
 17
```

#### Zakaria CHOUKRI





# Heroku part:

I couldn't do the Heroku part because it needs a credit card: heroku create furniture-prediction-app --region eu

# Assignment:

1. Choose a dataset, containing tabular data, of your choice I chose the Iris dataset

```
1 import pandas as pd
2 import numpy as np
 3 from sklearn.datasets import load_iris
 4 from sklearn.model_selection import train_test_split
5 from sklearn.preprocessing import StandardScaler
 6 from sklearn.linear_model import LogisticRegression
 7 from sklearn.ensemble import RandomForestClassifier
8 from sklearn.svm import SVC
9 from sklearn.metrics import accuracy score
10 import joblib
11
12 iris = load_iris()
13 X = pd.DataFrame(iris.data, columns=iris.feature_names)
14 y = pd.Series(iris.target, name='target')
16 print(f"Dataset shape: {X.shape}")
17 print(f"Missing values: {X.isnull().sum().sum()}")
18 print(f"Target distribution:\n{y.value_counts()}")
```

2. Prepare your data (treatment of missing values, etc.)

There is nothing to be done

```
Dataset shape: (150, 4)
Missing values: 0
Target distribution:
target
0 50
1 50
2 50
Name: count, dtype: int64
```

4. Use these characteristics to create the prediction page (like in the example above)

Iris Flower Classification
Sepal Length (cm)
Sepal Width (cm)
Petal Length (cm)
Petal Width (cm)
Predict

5. Compare the performance of multiple ML models to keep the best one at the end.

```
1 X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=0)
   3 scaler = StandardScaler()
   4 X_train_scaled = scaler.fit_transform(X_train)
   5 X_test_scaled = scaler.transform(X_test)
   7 models = . {
   8 ....'LogisticRegression': LogisticRegression(max_iter=200, random_state=0),
9 ....'RandomForest': RandomForestClassifier(n_estimators=100, random_state=0),
  ....dincommorest':-RandomForestClassifier(n
10 ....'SVM':-SVC(kernel='rbf',-random_state=0)
11 }
  12
  14 for name, model in models.items():
  21 best_model_name = max(scores, key=scores.get)
  22 best_model = models[best_model_name]
  23 print(f"\nBest model: {best_model_name} with accuracy {scores[best_model_name]:.4f}")
LogisticRegression: 1.0000
RandomForest: 1.0000
SVM: 1.0000
Best model: LogisticRegression with accuracy 1.0000
```

6. Create an application (i.e., a form) that predicts the target (value or class) of a record.

# **Iris Flower Classification** Sepal Length (cm) Sepal Width (cm) Petal Length (cm) Petal Width (cm) **Predict** Predicted Iris Class: virginica