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Deployment on Flask

An interface on the web allows us to communicate with the machine-learning model that is used in this assignment.

- Building ML model
- Deployment using Flask

Introduction:

This document provides an overview of a Python codebase developed to deploy a machine learning model using the Flask framework. The machine learning model is trained on the Iris dataset, a popular dataset in the field of machine learning and data science. The Iris dataset is commonly used for classification tasks and consists of measurements of iris flowers' sepal and petal dimensions, along with their corresponding species.

The primary objective of this codebase is to train a machine learning model capable of predicting the species of iris flowers based on their sepal length, sepal width, petal length, and petal width. Once the model is trained, it is deployed as a web application using the Flask framework, allowing users to interactively input the iris flower's measurements and receive predictions on their species in real-time.

This project serves as an example of deploying machine learning models into production environments, enabling seamless integration of predictive analytics into web applications. By leveraging the Flask framework, users can access the predictive capabilities of the trained model through a user-friendly interface, facilitating practical applications of machine learning in various domains.

Snapshot of each step of deployment:

ML model:

ML Model Result:

Deploying Flask

```
dimport numpy as np
from flask import Flask, request, jsonify, render_template
dimport pickle

# Create a Flask app
app = Flask(__name__)

# Load the model for making predictions
model = pickle.load(open("model.pkl", "rb"))

@app.route("/")
Didef home():
# Render the index.html template
return render_template("index.html")

@app.route("/predict", methods=["POST"])
Odef predict():
# Retrieve the input values from the form
float_features = [float(x) for x in request.form.values()]
features = [np.array(float_features)]

# Make predictions using the loaded model
prediction = model.predict(features)

# Render the index.html template with the prediction result
return render_template("index.html", prediction_text=f"The flower species is {prediction}")

> Sif __name__ == "__main__":
# Run the Flask app
app.run(debug=True, port=5000)
```

Output:

```
/Users/jef/Desktop/PYTHON/pythonProject5/venv/bin/python /Users/jef/Desktop/PYTHON/pythonProject5/app.py

* Serving Flask app 'app'

* Debug mode: on
WARNING: This is a development server. Do not use it in a production deployment. Use a production WSGI server instead.

* Running on http://127.8.8.1:5888

Press CTRL+C to quit

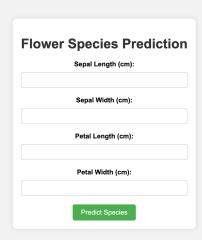
* Restarting with stat

* Debugger is active!

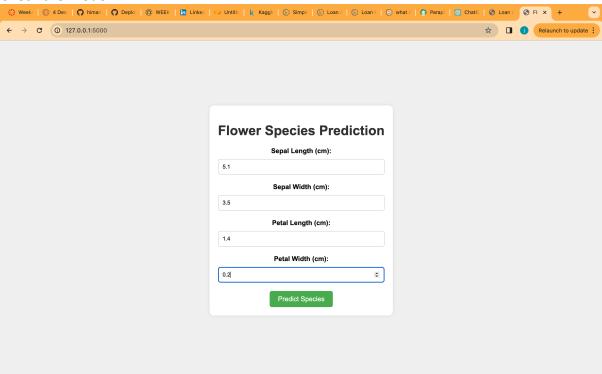
* Debugger PIN: 133-739-767
```

HTML (index.html)

Browser Output:



Check the Model:



Prediction:

