# Heroku Cloud Deployment, Week 5 Project

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### **Objective**

The objective of this project is to train and deploy a machine learning model using a simple toy dataset. The model was developed, saved, and deployed as both an API and a web application on **Heroku** using free-tier resources. A detailed PDF document, including snapshots of each deployment step, was created to document the process. The project demonstrates the complete workflow from model training to cloud deployment, ensuring accessibility and usability through a web-based interface.

### **Key Components**

### 1. Machine Learning Model:

- A pre-trained Logistic Regression model (titanic\_model.pkl) was used to make survival predictions.
- The model was trained using features such as:
  - Passenger Class (Pclass)
  - Gender (Sex)
  - Age
  - Number of Siblings/Spouse aboard (SibSp)
  - Number of Parents/Children aboard (Parch)
  - Fare paid (Fare).

#### 2. Flask API:

- o A Flask application was built to host the predictive model.
- The /predict endpoint accepts POST requests with passenger data in JSON format and returns predictions in JSON form.

 The home route (/) serves as a connection point for static files and basic API confirmation.

#### 3. Web Interface:

- An interactive HTML front-end allows users to input passenger data through a form and receive predictions in real time.
- The background of the page features an image of the Titanic to enhance user experience.
- o JavaScript handles form submission and interacts with the Flask API for predictions.

### 4. File Organization:

- The Titanic image file is stored in a static folder and linked in the HTML file to ensure proper rendering.
- Flask was configured to serve static files and correctly route requests to the titanic.html front-end.

## Steps Taken

## 1. Data Preparation and Model Training:

 A Logistic Regression model was trained using the Titanic dataset, exported as titanic\_model.pkl.

### 2. Backend Development:

 A Flask app was created to load the model, define prediction logic, and serve the front-end.

#### 3. Front-End Design:

 A responsive HTML file was developed to capture user inputs and display predictions.

# 4. Integration and Deployment:

- The Flask app was tested locally, with HTML and static assets linked properly.
- End-to-end functionality was confirmed with accurate predictions based on sample inputs.

# **Sample Input and Output**

## Input:

Pclass: 1

• Sex: Female (0)

• Age: 25

SibSp: 0

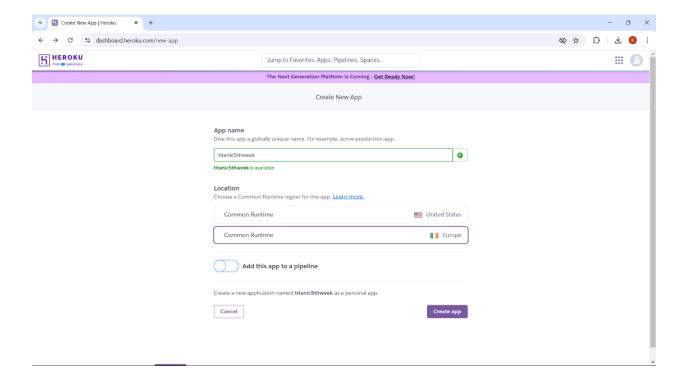
• Parch: 1

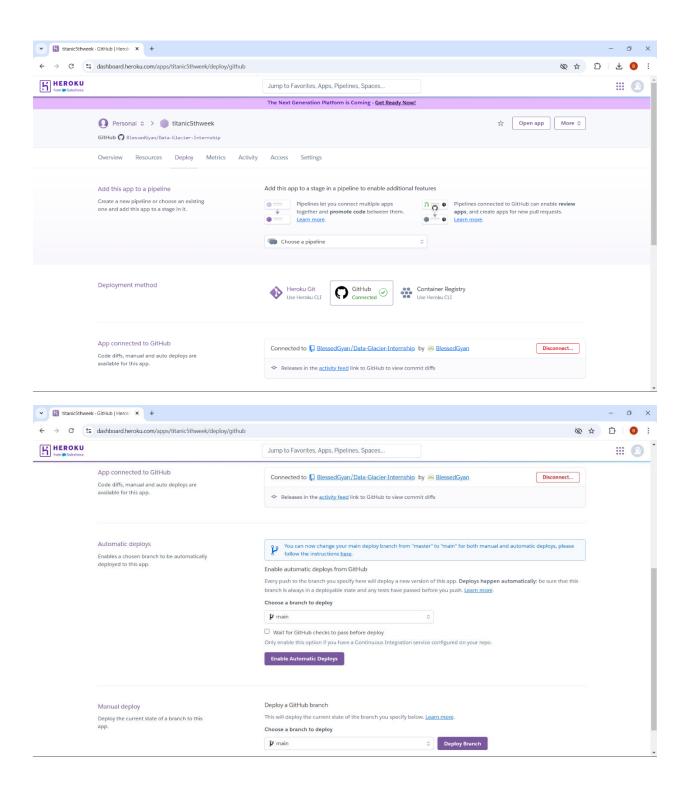
• Fare: 50.0

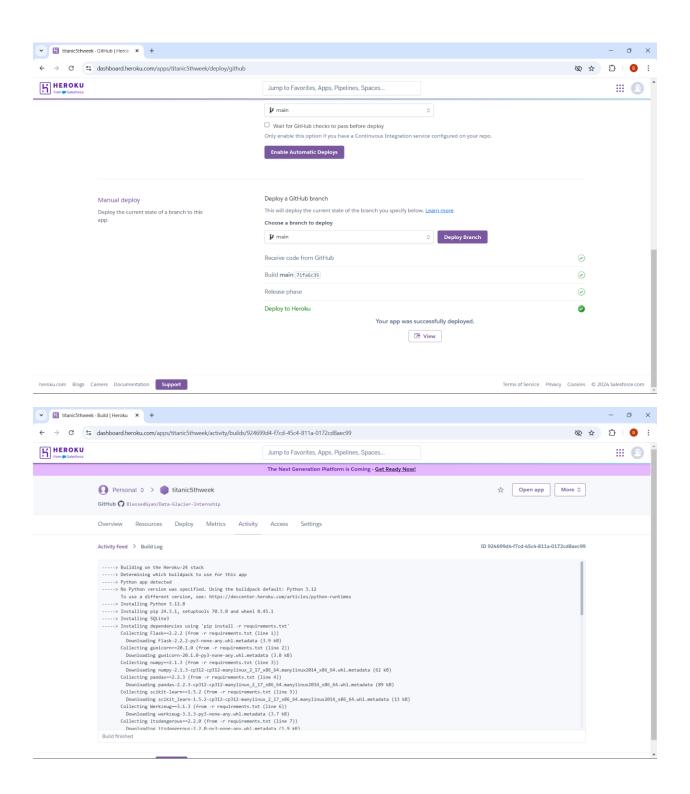
## **Output:**

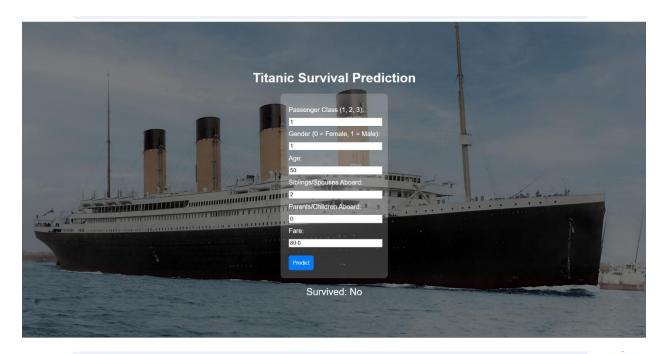
JSON Response: { "Survived": 1 }

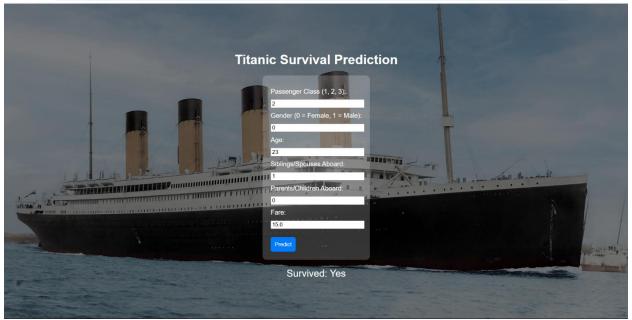
• Display: "Survived: Yes"











## Conclusion

The deployment was **successful**, and the model is now accessible via both API and web application on **Heroku**. The process provided hands-on experience in model deployment, cloud computing, and API integration. This project showcases an efficient approach to deploying machine learning models in a cloud environment, making them scalable and accessible.