

Machine Learning Model Deployment

Deploying Flask Apps

Using Heroku



#### Steps:

- Create and Pickle a Model Using Titanic Data
- Create Flask App
- Test Flask App in Production (optional)
- Create GitHub Repository (optional)
- Deploy to Heroku
- Test Working App



# **Creating and Training a Model**

**Titanic Data set** 



## #serializing our model to a file called model.pkl

import pickle
pickle.dump(dt\_clf\_gini, open(".../Logistic\_Regression\_model.pkl","wb"))

The <u>pickle</u> module implements a fundamental, but powerful algorithm for serializing and de-serializing a Python object structure. "Pickling" is the process whereby a Python object hierarchy is converted into a byte stream, and "unpickling" is the inverse operation, whereby a byte stream is converted back into an object hierarchy. Pickling (and unpickling) is alternatively known as "serialization", "marshalling," 1 or "flattening", however, to avoid confusion, the terms used here are "pickling" and "unpickling".

As a data scientist, you will use sets of data in the form of dictionaries, DataFrames, or any other data type. When working with those, you might want to save them to a file, so you can use them later on or send them to someone else. This is what Python's pickle module is for: it serializes objects so they can be saved to a file, and loaded in a program again later on.



# Creating a Simple Web Application using Flask

- Flask is a python based microframework used for developing small scale websites.
- It can create a REST API that allows you to send data, and receive a prediction as a response.
- Similar to what Django REST framework .





# Creating a Simple HTML Form for WebApp

Index.html

#### Titanic Prediction Form

Age						
Sex	Ma	ale	-			
Fare						
Pclas	šS	1	▼			
Submit						



# **Installing Flask**

!pip install flask

Create a project directory i.e flask-app

Put Logistic\_Regression\_model.pkl in project directory **flask-app** 

mkdir **templates** under directory flask-app

Put index.html in templates directory



# Create Logistic\_Regression\_model.py file and put under Titanic\_Flask\_Prediction directory

```
import os
import numpy as np
import flask
import pickle
from flask import Flask, render_template, request
#creating instance of the class
app=Flask(__name__)
#to tell flask what url should trigger the function index()
@app.route('/')
@app.route('/index')
def index():
  return flask.render_template('index.html')
if name == ' main ':
  app.run(port = 5000, debug=True)
```

- app=Flask(\_\_name\_\_) create an instance of flask.
- @app.route('/') is used to tell flask what url should trigger the function index()
- In the function index we use render\_template('index.html') to display the script index.html in the browser.



#### Now run the application from jupyter terminal

python Logistic\_Regression\_model.py

This should run the application and launch a simple server. Open <a href="http://127.0.0.1:5000/">http://127.0.0.1:5000/</a> to see the html form.

```
PS C:\Users\DELL> python Logistic_Regression_model.py

* Serving Flask app "Logistic_Regression_model" (lazy loading)

* Environment: production

WARNING: This is a development server. Do not use it in a production deployment.

Use a production WSGI server instead.

* Debug mode: on

* Restarting with stat

* Debugger is active!

* Debugger PIN: 229-278-700

* Running on http://127.0.0.1:5000/ (Press CTRL+C to quit)
```



#### **Putting the Prediction part in file**

#### Logistic\_Regression\_model.py

```
#prediction function
def ValuePredictor(to_predict_list):
  to_predict = np.array(to_predict_list).reshape(1,4)
  loaded model =
pickle.load(open("Logistic_Regression_model.pkl","rb"))
  result = loaded_model.predict(to predict)
  return result(0)
@app.route('/result',methods = ['POST'])
def result():
  if request.method == 'POST':
     to predict list = request.form.to dict()
     to_predict_list=list(to_predict_list.values())
     to_predict_list = list(map(int, to_predict_list))
     result = ValuePredictor(to predict list)
```

- Here after the form is submitted, the form values are stored in variable *to\_predict\_list* in the form of dictionary.
- We convert it into a list of the dictionary's values and pass it as an argument to *ValuePredictor()* function.
- In this function, we load the Logistic\_Regression\_model.pkl file and predict the new values and return the result.
- This result/prediction is then passed as an argument to the template engine with the html page to be displayed.



Create the following *result.html* file and add it to templates directory.

- Run the application again
- python Logistic\_Regression\_model.py
- It should predict the income after submitting the form.



## This is how our project layout looks like



## Successfully created the Webapp Now it's time to use heroku to deploy it



**Heroku i**s a platform as a service (PaaS) that enables developers to build, run, and operate applications entirely in the cloud.



## **Install gunicorn**

#### !pip install gunicorn

Gunicorn handles requests and takes care of complicated things like threading very easily in real production.



Three common building blocks when deploying a Python web application to production are:

- A web server (like nginx)
- A WSGI (Web Server Gateway Interface) application server (like Gunicorn)
- Your actual application (written using a developer-friendly framework like Django)
- The web server accepts requests, takes care of general domain logic and takes care of handling https connections. Only requests which are meant to arrive at the application are passed on toward the application server and the application itself.
   The application code does not care about anything except being able to process single requests.
- Gunicorn takes care of everything which happens in-between the web server and your web application. This way, when coding up your a Django application you don't need to find your own solutions for:
  - communicating with multiple web servers
  - reacting to lots of web requests at once and distributing the load
  - keepiung multiple processes of the web application running



#### Creating a requirements.txt file and put under project directory

# pip freeze > requirements.txt

- The requirements.txt file will contain all of the dependencies for the flask app.
- In our local machine, we have installed a lot of libraries and other important files like flask, gunicorn, sklearn etc. We need to tell heroku that our project requires all these libraries to successfully run the application. This is done by creating a **requirements.txt** file.



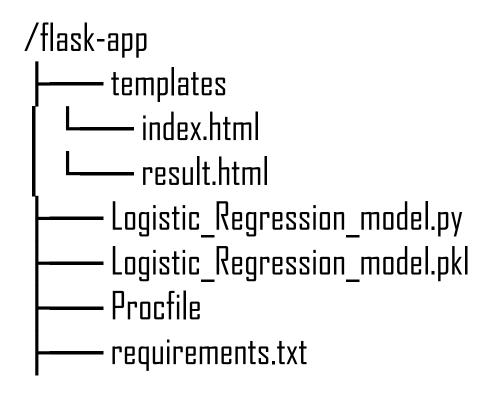
#### Creating Procfile and put under project directory

- A Procfile specifies the commands that are executed by a Heroku app on startup.
- To create one, open up a new file named Procfile (no extension) in the working directory and paste the following.

web: gunicorn app:app

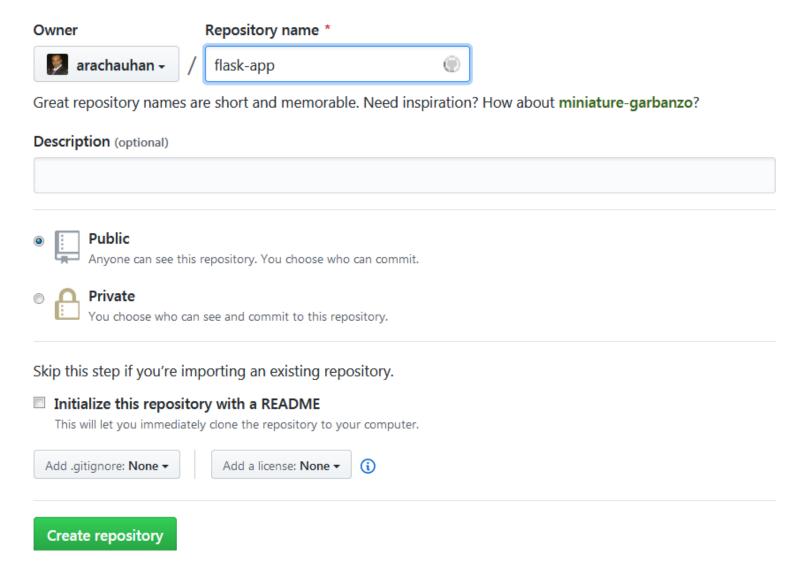


## This is how our project layout looks like now





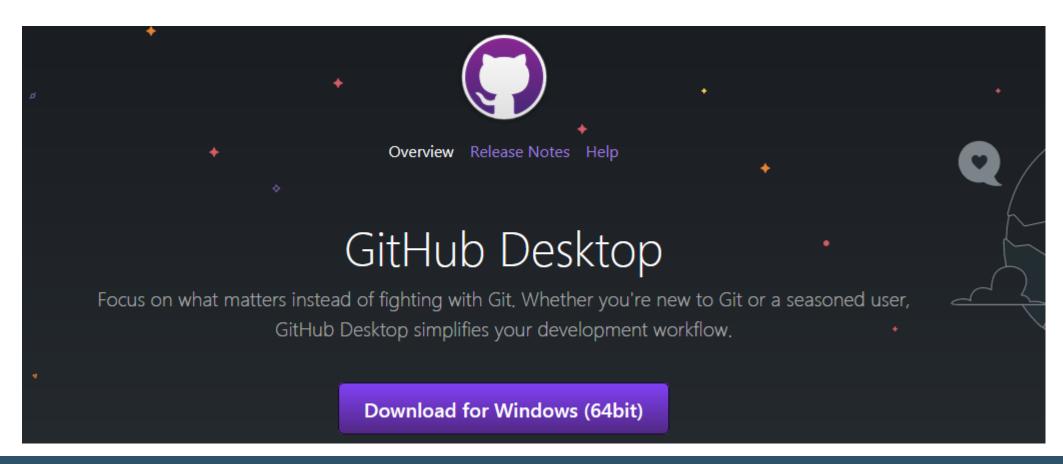
#### **Create New GitHub Repository**





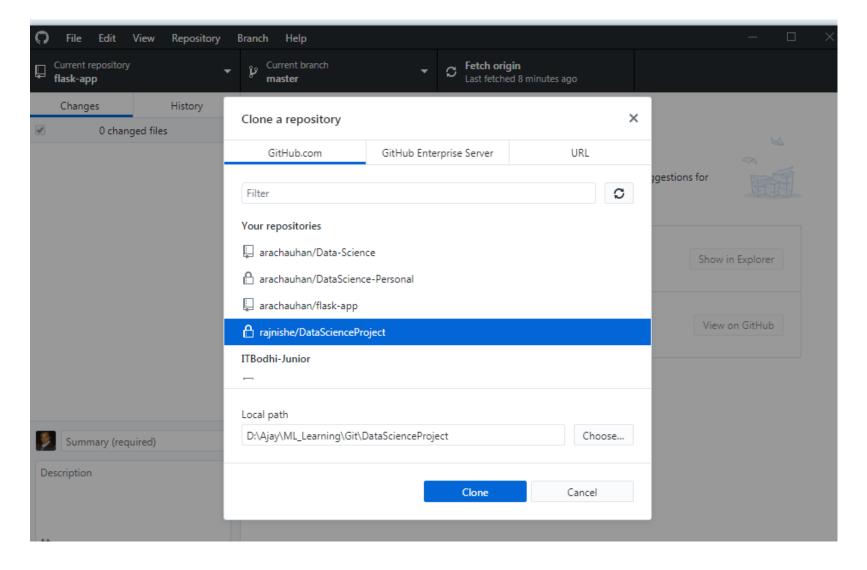
## **Using GitHUb Desktop for Rich Experience**

- Download and install it
- We will be COMMIT complete project directory to GitHub and Connect Heroku to GitHub for WebApp deployment.
- You can also use <u>heroku git</u> that uses command line to deploy models.



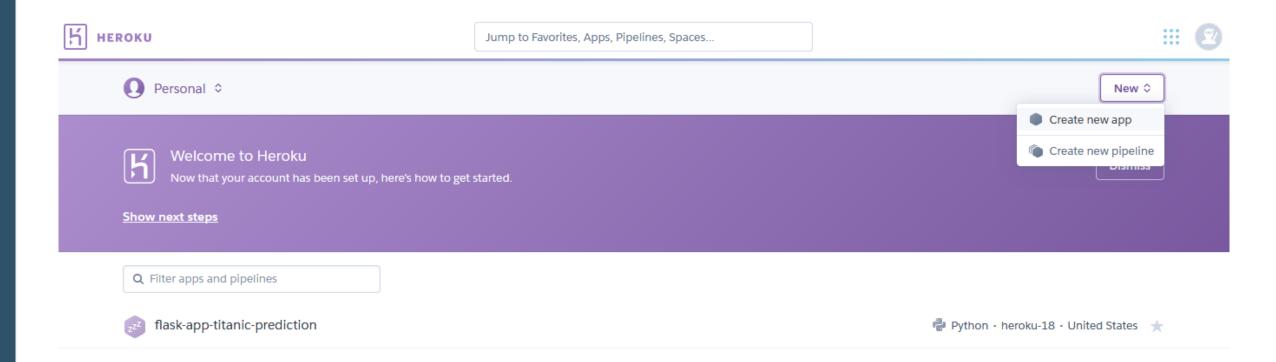


## Clone GitHub repository to local project directory



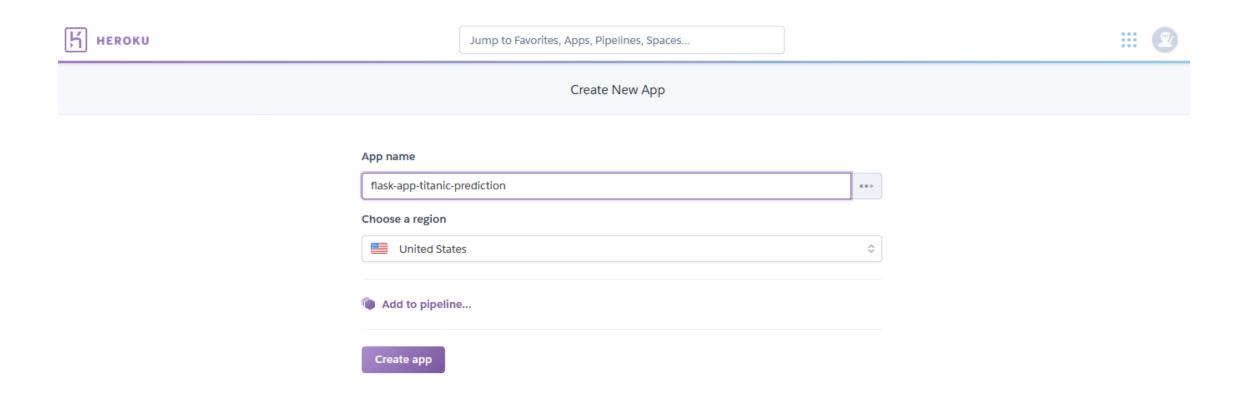


- Create Fee account at www.heroku.com
- Create a new app simply by choosing a name and clicking "create app". This name doesn't matter but it does have to be unique.



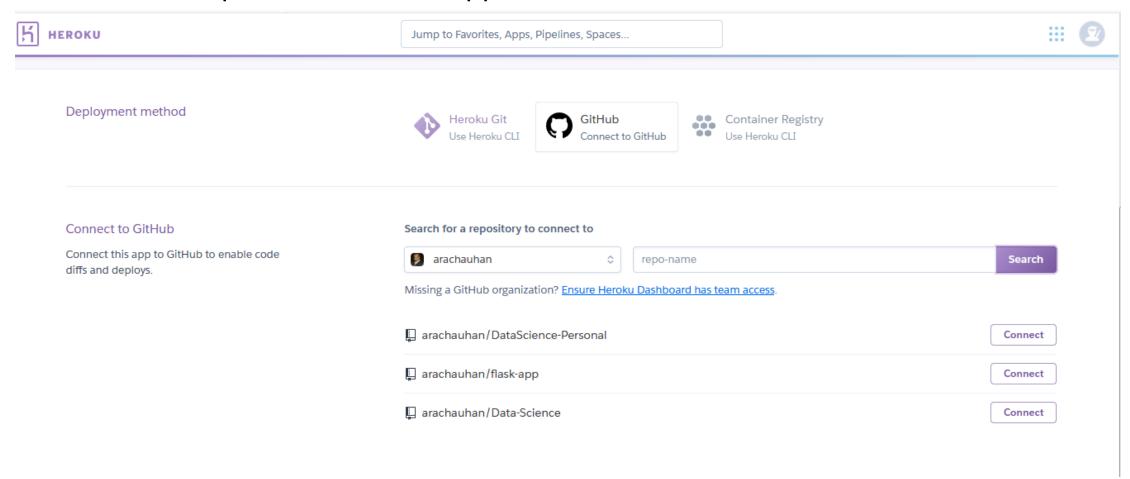


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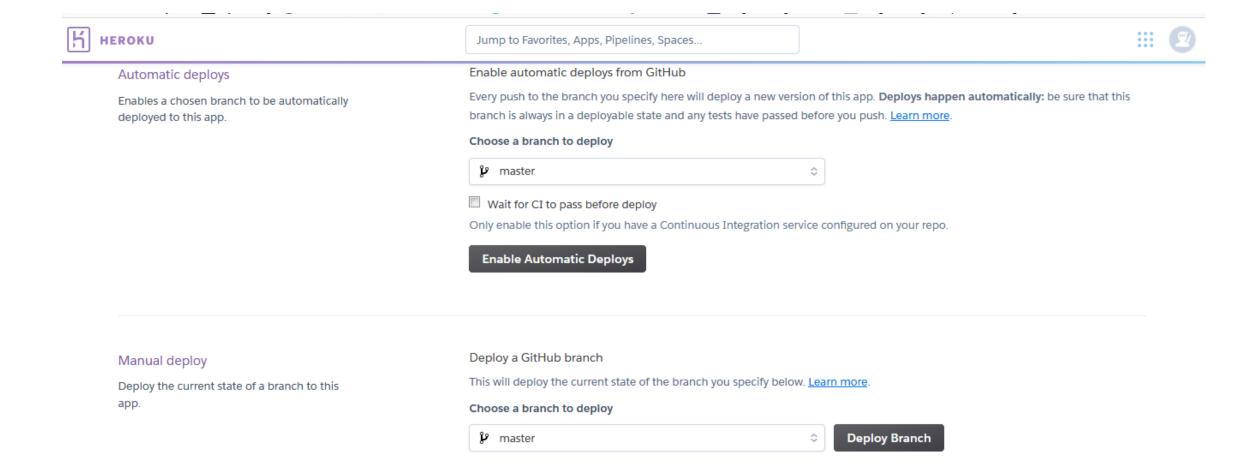


- You have a few options for the way you can deploy the app.
- Heroku CLI and GitHub
- Select Github
- Search repo-name as flask-app and connect





#### Just scroll to the bottom of the page and click "Deploy Branch"





#### If everything Goes well, You should see below message

Your app was successfully deployed.



If something went wrong, check your requirements.txt, delete and dependencies that are giving you problems, and try again



#### **Deploying with Heroku CLI**

- In the Heroku CLI section, you will see these instructions to follow for deployment.
- Paste each command into your terminal and follow any prompts like logging in.
- Pay attention to any commands you will need to modify, such as cd my-project/ — where my-project/ should actually be your project directory.
- The git remote should be set to the app name from Heroku EXACTLY.



#### Install the Heroku CLI

Download and install the Heroku CLI.

If you haven't already, log in to your Heroku account and follow the prompts to create a new SSH public key.

```
$ heroku login
```

#### Create a new Git repository

Initialize a git repository in a new or existing directory

```
$ cd my-project/
$ git init
$ heroku git:remote -a flak-app
```

#### Deploy your application

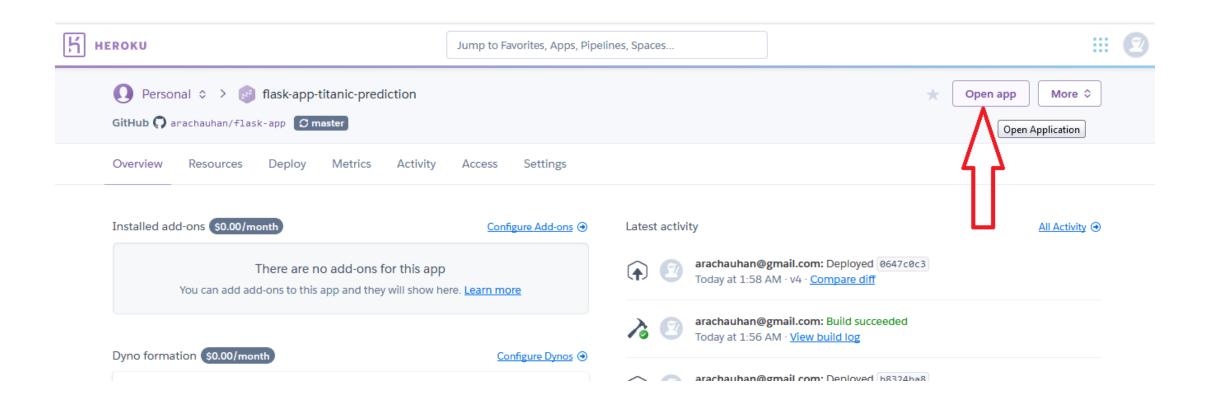
Commit your code to the repository and deploy it to Heroku using Git.

```
$ git add .
$ git commit -am "make it better"
$ git push heroku master
```

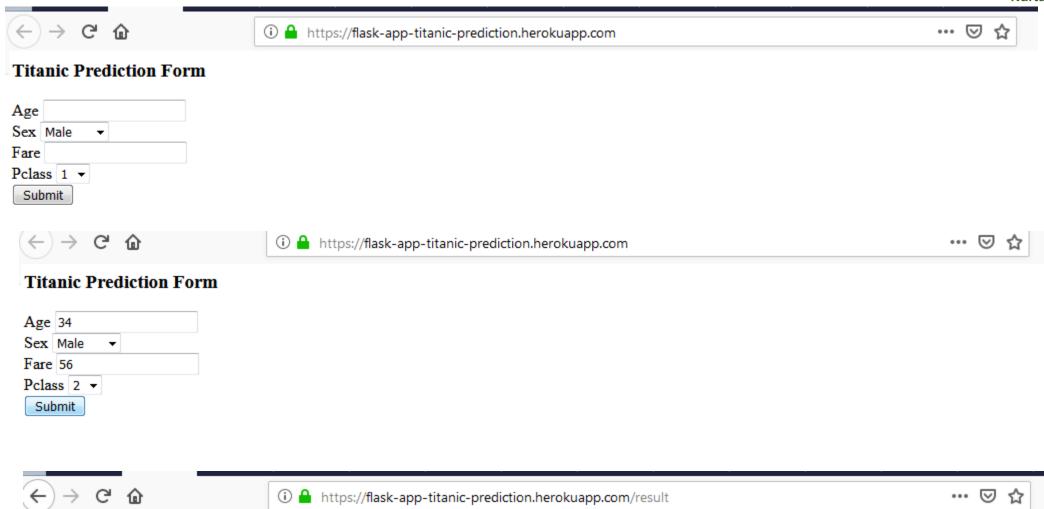


#### **Test the Deployed Model & Generate Prediction**

#### Click on Open App







#### Pessenger NOT survived