

Deploying Neural Networks on the Web

Vladimir Haltakov

Notebook:
Sam Ray



About me

Vladimir Haltakov

- Self-driving car engineer at BMW (localization)
- PhD in Computer Vision and Machine Learning from TU Munich
- Helping people on Twitter to get started with Machine Learning
- Traveling and photography



[@haltakov](https://twitter.com/haltakov)



Deploying Neural Networks on the Web

- Designing and training a neural network is only a part of the game
- From a pytorch model to a web application
- The Web - the easiest way to reach millions of people
- Tutorial by Sam Ray and Vladimir Haltakov
- Other contributors: Konrad Kording, Spiros Chavlis

What you are going to learn?

- How to use Flask for serving web pages?
- How to apply the MVVM design pattern to write maintainable code?
- How to build a REST API?
- How to create an interactive UI for your service?
- How to integrate your deep learning model?
- How to deploy your service on Heroku?

Flask

- Python web application microframework
- Lightweight, easy to use, scalable
- Big community and many extensions
- Easier to learn than Django
- Examples: Pinterest and LinkedIn



Simple Flask App

```
app = Flask(__name__)

@app.route("/")
def home():
    return "<h1>Welcome to Neuromatch</h1>"

app.run()
```

Using ngrok

```
from flask_ngrok import run_with_ngrok

# ...

run_with_ngrok(app)
app.run()
```

- The problem with running Flask in a notebook
- URL <http://127.0.0.1:5000/> not accessible from the Internet
- Create a tunnel from your notebook to the Internet with ngrok
- URL like <http://33ca1c4cb1f9.ngrok.io>

Alternative web frameworks

- FastAPI - focused on speed
- Bottle - another framework coming with the standard library
- Tornado - an asynchronous web framework

Jinja2 Templates

- Fast, expressive, extensible templating engine
- Separate HTML code and data
- Main features:

- Variables `{{ value }}`
- If statements `{% if value > 0 %} ... {% else %}`
- Loops `{% for value in list %}`
- Inheritance `{% extends "layout.html" %}`
- Modules `{% include helper %}`

Jinja2 Templates - example

Data

```
OrderedDict([('system', 'Linux'),  
             ('node', '3a2677501e8b'),  
             ('release', '5.4.104+'),  
             ('version', '#1 SMP Sat Jun 5 09:50:34 PDT 2021'),  
             ('machine', 'x86_64'),  
             ('processor', 'x86_64')])
```

Template

```
<table>  
  <tr>  
    <th>Property</th>  
    <th>Value</th>  
  </tr>  
  {% for key, value in platform.items() %}  
    <tr>  
      <td>{{ key }}</td>  
      <td>{{ value }}</td>  
    </tr>  
  {% endfor %}  
</table>
```

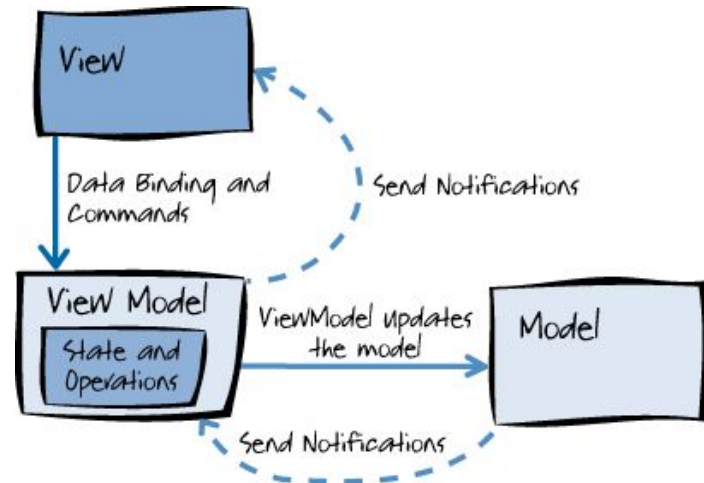
Property Value

system	Linux
node	3a2677501e8b
release	5.4.104+
version	#1 SMP Sat Jun 5 09:50:34 PDT 2021
machine	x86_64
processor	x86_64

Rendered HTML page

The MVVM Design Pattern

- Model View View-Model (MVVM)
- Powerful design pattern for web apps
 - View - the user interacts
 - Model - the data
 - View-Model - binding between the View and the application state



MVVM - the Model

- The model stores your data
- Implements the access to the data (for example in a database)
- ORM (Object Relational Mapper)
- SQLAlchemy

```
class PointModel:  
  
    def __init__(self, x, y):  
        self.x = x  
        self.y = y
```

MVVM - the View

- Structure, layout, and appearance
- Implements the interaction with the user (for example input)
- Implements the rendering HTML page

```
class PointView(Resource):  
    def get(self):  
        point = PointViewModel.get_sample_data()  
        return f"Point: (x={point.x}, y={point.y})"
```

MVVM - the View-Model

- Contains the state of the application
- Implements an automatic binding from the View to the state
- Handles the communication between the view and the state.

```
class PointViewModel:

    @classmethod
    def get_sample_data(cls):
        return SamplePointModel(2, 5)

    def setup(self, api):
        api.add_resource(PointView, '/')
```

REST API

- Representational State Transfer (REST)
- Enable communication and interaction with other web services
- Rules and constraints for designing APIs
 - Uniform interface
 - Statelessness
 - Cacheability
 - Layered system
 - Code on demand



Example - ML classifier REST API

- GET /info - information about your model
 - {"model": "ResNet", "parameters": 1000000}
- GET /classes - list of the supported classes
 - ["dog", "cat", "car"]
- POST /classify - classify an image
 - {"dog": 0.93, "cat": 0.05, "car": 0.02}

REST API - documentation

GET **/platform** This examples uses PlatformView Resource

It works also with swag_from, schemas and spec_dict

Parameters Try it out

No parameters

Responses Response content type application/json

Code	Description
200	A single Machine item

Example Value Model

```
{
  "machine": "string",
  "node": "string",
  "processor": "string",
  "system": "string"
}
```

```
import flasgger

# ...

swg = flasgger.Swagger(app)
```

Docs URL: <http://xxxxxxx.ngrok.io/apidocs/>



Interactive UI - Vue.js



- Create interactive UI with Vue.js
- MVVM front end JavaScript framework
- Building UIs and single page web apps
- Alternatives
 - React
 - Angular

Vue.js - app overview

- Back end (Flask)
 - Serve REST API at `/platform`
 - Serve Vue.js template at `/`
- Front end (Vue.js)
 - Define HTML template of the webpage
 - Fetch and display data from `/platform` when initialized

Vue.js - include libraries

- We first need to include the vue.js and the axios.js libraries.
- We are going to use axios to fetch data from our API

```
<head>  
  <script src="https://cdn.jsdelivr.net/npm/vue/dist/vue.js"></script>  
  <script src="https://cdnjs.cloudflare.com/ajax/libs/axios/0.21.1/axios.min.js"></script>  
</head>
```

Vue.js - define the template

- We can define how the data is displayed
- We can display variables using {{ value }}

```
<div id="app">
  <ul>
    <li><strong>System:    </strong>{{ platform.system }}</li>
    <li><strong>Machine:   </strong>{{ platform.machine }}</li>
    <li><strong>Processor: </strong>{{ platform.processor }}</li>
    <li><strong>Node:      </strong>{{ platform.node }}</li>
  </ul>
</div>
```

Vue.js - define the template

- Initialize the Vue application
- Define the data function
- When initialized (`mounted()`),
fetch data from the platform API
(`/platform`) using axios

```
var app = new Vue({  
  el: '#app',  
  data() {  
    return {  
      platform: null  
    }  
  },  
  mounted () {  
    axios.get('/platform')  
      .then(response => (this.platform = response.data))  
  }  
});
```

Vue.js - why not use Jinja?

- We did the same with Jinja, why do we need Vue.js?
 - Jinja templates are rendered in the back end
 - Vue templates are rendered dynamically
 - We will add dynamic functionality in the next section

Model Presentation

- Flask web application that classifies images
- Two entry points
 - `/` serve an interactive UI for uploading images
 - `/predict` classify the input image
- Use a pre-trained DenseNet model from torchvision

Torchvision

- Official PyTorch package
 - Datasets (ImageNet, MNIST, COCO, ...)
 - Models (AlexNet, ResNet, DenseNet, ...)
 - Image Transformations
 - Other computer vision operators and utilities for working with images

Alternative: TorchServe

- Official PyTorch package to serve models as web servers
- REST API for classifying images
- Requires a separate deployment of the API

Loading a pre-trained model

```
from torchvision import models

model = models.densenet121(pretrained=True)
model.eval()

class_labels_url = "https://raw.githubusercontent.com/pytorch/hub/master/imagenet_classes.txt"
class_labels = urlopen(class_labels_url).read().decode("utf-8").split("\n")

transform = transforms.Compose([
    transforms.Resize(256),
    transforms.CenterCrop(224),
    transforms.ToTensor(),
    transforms.Normalize(mean=[0.485, 0.456, 0.406], std=[0.229, 0.224, 0.225]),
])
```

Classifying an image

```
def predict(model, transform, image, class_labels):  
    # Transform the image and convert it to a tensor  
    image_tensor = transform(image).unsqueeze(0)  
  
    # Pass the image through the model  
    with torch.no_grad():  
        output = model(image_tensor)  
  
    # Select the class with the highest probability and look up the name  
    class_id = torch.argmax(output).item()  
    class_name = class_labels[class_id]  
  
    # Return the class name  
    return class_name
```

Classifying an image



Photo by Marliese Streefland on Unsplash



English foxhound

Creating the UI

Select image to classify: cat.jpg



tiger cat

```
<div id="app" style="width: 50%; margin: 200px auto">
  <form id="imageForm" enctype="multipart/form-data" method="POST">

    <label for="imageFile">Select image to classify:</label>
    <input id="imageFile" name="file" type="file" />

    <div v-if="prediction">
      {{ prediction }}
    </div>

    <input v-if="image" type="submit" value="Classify Image" />

  </form>
</div>
```

The Vue.js Application

Select image to classify: cat.jpg



tiger cat

```
var app = new Vue({  
  el: "#app",  
  data() {  
    return {  
      image: null,  
      prediction: null,  
    };  
  },  
});
```

Using the Classification API

```
document.getElementById("imageForm").addEventListener("submit", (e) => {  
  axios  
    .post("/predict", new FormData(document.getElementById("imageForm")), {  
      headers: {  
        "Content-Type": "multipart/form-data",  
      },  
    })  
    .then((response) => (app.prediction = response.data));  
  e.preventDefault();  
});
```


Using the Classification API

```
document.getElementById("imageFile").addEventListener("change", (e) => {  
  const [file] = document.getElementById("imageFile").files;  
  if (file) {  
    app.image = URL.createObjectURL(file);  
  }  
});
```

Using the Classification API

```
app = Flask(__name__)

@app.route("/")
def home():
    return index_template

@app.route("/predict", methods=['POST'])
def predict_api():
    image_file = request.files['file']
    image_bytes = image_file.read()
    image = Image.open(io.BytesIO(image_bytes))

    class_name = predict(model, transform, image, class_labels)

    return class_name

run_with_ngrok(app)
app.run()
```



Deploying on Heroku

- What is Heroku?
 - Public cloud provider
 - Platforms-as-a-Service
 - Pre-configured environments
 - Very easy deployment
 - Scalable and flexible
 - Free tier



Deploying on Heroku - Steps

1. Create a local application (outside of Colab)
2. Create a Heroku account
3. Install the Heroku CLI
4. Create a new Heroku application
5. Initialize a Git repository
6. Push to deploy



Prepare Python Environment

1. Create a Python environment

```
python -m venv .venv
```

2. Activate the environment

```
source .venv/bin/activateor .venv\Scripts\activate.bat
```

3. Install dependencies

```
pip install flask Pillow gunicorn
```

4. Install PyTorch (without torchaudio)

```
pip install torch torchvision
```



Prepare Python Environment

1. Create a Python environment

```
python -m venv .venv
```

2. Activate the environment

```
source .venv/bin/activate or .venv\Scripts\activate.bat
```

3. Install dependencies

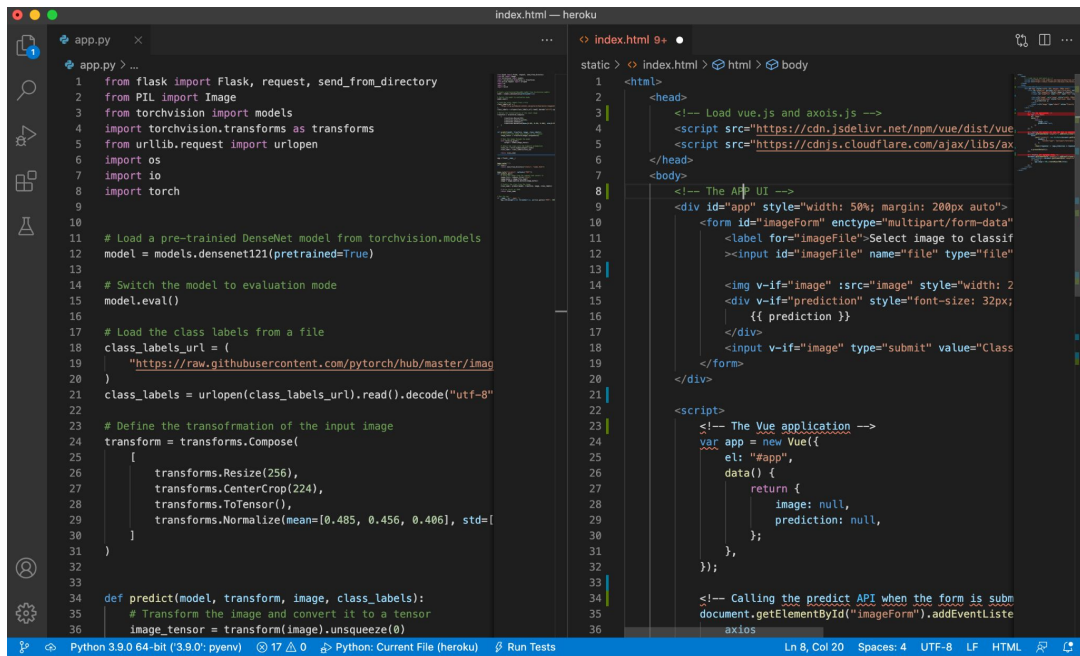
```
pip install flask Pillow gunicorn
```

4. Install PyTorch (without torchaudio)

```
pip install torch torchvision
```



Create a Local Application



```
app.py
1 from flask import Flask, request, send_from_directory
2 from PIL import Image
3 from torchvision import models
4 import torchvision.transforms as transforms
5 from urllib.request import urlopen
6 import os
7 import io
8 import torch
9
10
11 # Load a pre-trained DenseNet model from torchvision.models
12 model = models.densenet121(pretrained=True)
13
14 # Switch the model to evaluation mode
15 model.eval()
16
17 # Load the class labels from a file
18 class_labels_url = (
19     "https://raw.githubusercontent.com/pytorch/hub/master/imag
20 )
21 class_labels = urlopen(class_labels_url).read().decode("utf-8")
22
23 # Define the transformation of the input image
24 transform = transforms.Compose(
25     [
26         transforms.Resize(256),
27         transforms.CenterCrop(224),
28         transforms.ToTensor(),
29         transforms.Normalize(mean=[0.485, 0.456, 0.406], std=[
30     ])
31 )
32
33
34 def predict(model, transform, image, class_labels):
35     # Transform the image and convert it to a tensor
36     image_tensor = transform(image).unsqueeze(0)
```

```
index.html
1 <html>
2 <head>
3   <!-- Load vue.js and axios.js -->
4   <script src="https://cdn.jsdelivr.net/npm/vue/dist/vue.js"></script>
5   <script src="https://cdn.jsdelivr.net/npm/axios/dist/axios.min.js"></script>
6 </head>
7 <body>
8   <!-- The App UI -->
9   <div id="app" style="width: 50%; margin: 20px auto">
10     <form id="imageForm" enctype="multipart/form-data">
11       <label for="imageFile">Select image to classify</label>
12       <input id="imageFile" name="file" type="file">
13     </form>
14     
15     <div v-if="prediction" style="font-size: 32px; margin-top: 10px;">
16       {{ prediction }}
17     </div>
18     <input v-if="image" type="submit" value="Classify">
19   </div>
20
21   <!-- The Vue application -->
22   <script>
23     var app = new Vue({
24       el: "#app",
25       data() {
26         return {
27           image: null,
28           prediction: null,
29         };
30       },
31       methods: {
32         predict() {
33           // Calling the predict API when the form is submitted
34           document.getElementById("imageForm").addEventListener("submit", this.predict);
35         }
36       }
37     });
```

Test it locally:
python app.py



Preparing for Heroku

requirements.txt

```
click==8.0.1
Flask==2.0.1
gunicorn==20.1.0
itsdangerous==2.0.1
Jinja2==3.0.1
MarkupSafe==2.0.1
numpy==1.21.1
Pillow==8.3.1
torch==1.9.0
torchvision==0.10.0
typing-extensions==3.10.0.0
Werkzeug==2.0.1
```

Too large because they include
both GPU and CPU code

Procfile

```
web: gunicorn app:app
```


Preparing for Heroku

requirements.txt

```
-f https://download.pytorch.org/whl/torch_stable.html
click==8.0.1
Flask==2.0.1
gunicorn==20.1.0
itsdangerous==2.0.1
Jinja2==3.0.1
MarkupSafe==2.0.1
numpy==1.21.1
Pillow==8.3.1
torch==1.9.0+cpu
torchvision==0.10.0+cpu
typing-extensions==3.10.0.0
Werkzeug==2.0.1
```

Procfile

```
web: gunicorn app:app
```

Deploy Your App

1. Create a Heroku account
2. Install the Heroku CLI
3. Login to Heroku

```
heroku login
```

4. Create a new Heroku application

```
heroku create <application name>
```

5. Initialize a git repository

6. Push to deploy

```
git push heroku master
```

```
git init
git add app.py Procfile requirements.txt static
git commit -m "Initial commit"
heroku git:remote -a <application name>
```



Heroku Dashboard

<https://dashboard.heroku.com/apps>

The screenshot shows the Heroku Dashboard for the application 'vladimir-classifier-app'. The top navigation bar includes a user profile icon, 'Personal', and the app name. On the right, there are buttons for 'Open app' and 'More'. Below the navigation bar, a tab bar shows 'Overview' (selected), 'Resources', 'Deploy', 'Metrics', 'Activity', 'Access', and 'Settings'. The main content area features a promotional banner for Heroku Pipelines with a 'Create a Heroku Pipeline' button. Below this, the 'Installed add-ons' section shows a message: 'There are no add-ons for this app. You can add add-ons to this app and they will show here. [Learn more](#)'. To the right, the 'Latest activity' section lists two events: a deployment by 'vladimir.haltakov@gmail.com' on Aug 3 at 10:22 PM, and a successful build by the same user on Aug 3 at 10:21 PM, with a link to 'View build log'.

Change the Domain Name?

Domains

You can add custom domains to any Heroku app, then visit [Configuring DNS](#) to setup your DNS target.

Your app can be found at <https://vladimir-classifier-app.herokuapp.com/>

Add domain

🔍 Filter domains

Custom domains will appear here

Custom domains allow you to access your app via one or more non-Heroku domain names (for example, `www.yourcustomdomain.com`)

Summary

- Create Flask applications
- Apply the MVVM design pattern
- Create REST APIs
- Interactive UI with Vue.js
- Integrate a PyTorch model into Flask
- Deploy on Heroku

