

This project is developed based on Python 3.9. Therefore, when downloading Python, please ensure its version is  $\geq 3.9$ . You can download Python from this link: <https://www.python.org/downloads/> We highly recommend using a Python IDE to run our project. We suggest using PyCharm. You can download it from this link: <https://www.jetbrains.com/pycharm/>

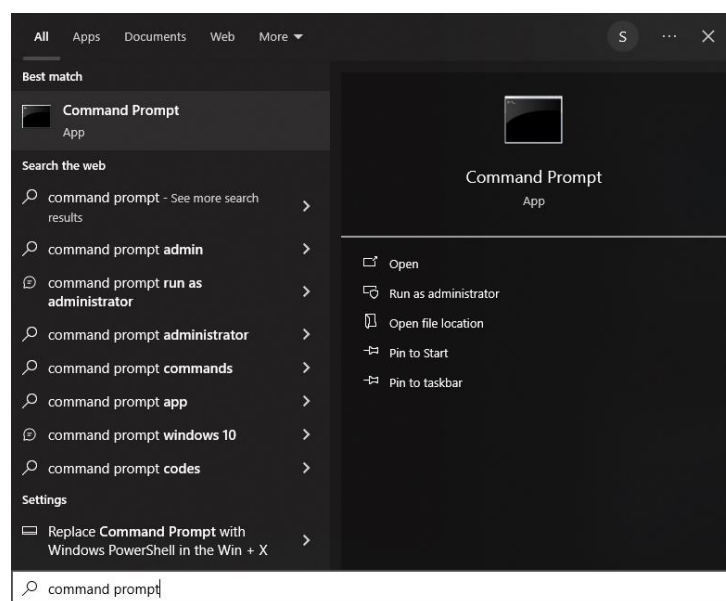
1;

## How to download all 300 FELO Leaflets (Web scraping):

\*Tested on Windows 10

### Pre-requisites:

1. Download [Google Chrome](#)
2. Have [Python 3](#) installed
3. Have the necessary packages installed. To install the packages, open Command Prompt via Start. without quotation marks, run '`pip install requests beautifulsoup4 selenium`'



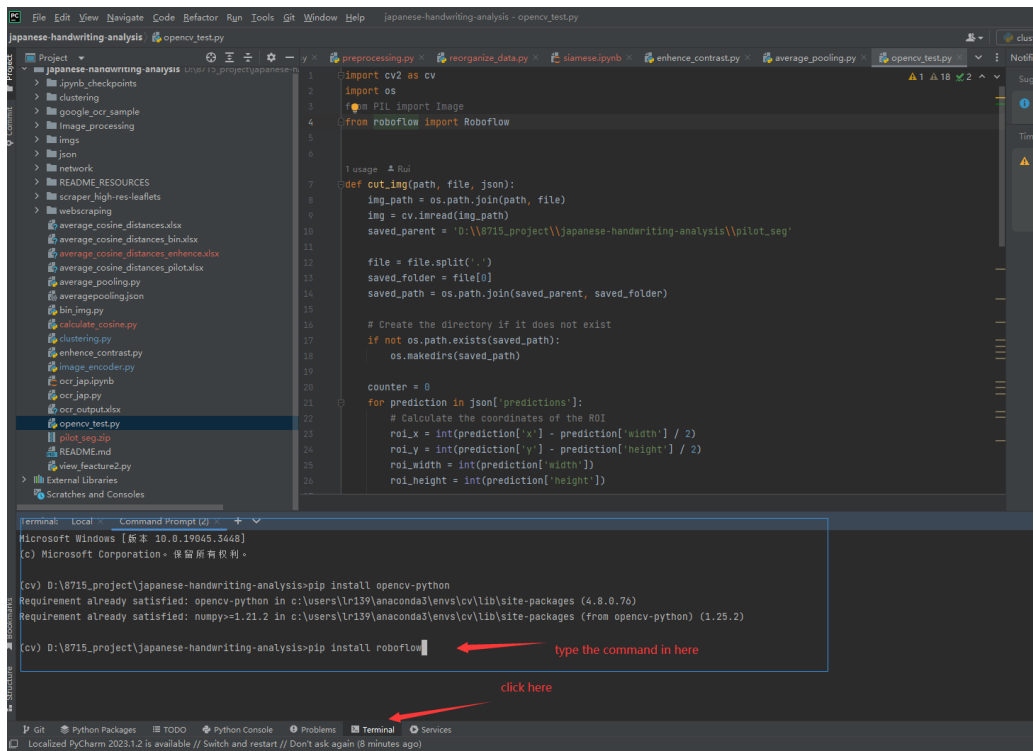
### How to download the leaflets:

1. Within the Command Prompt, navigate to `/scraper_high-res-leaflets/` folder.
2. Without quotation marks, run '`python scraper.py`' or '`python3 scraper.py`'

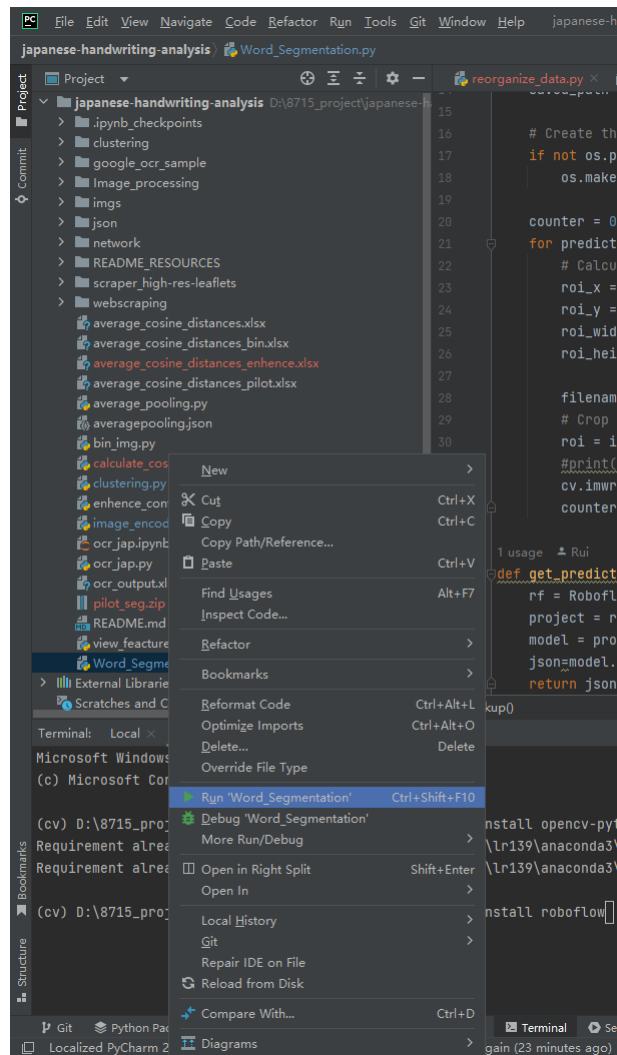
Leaflets will be downloaded within the same directory the script is ran. Based on FELO\_leaflets spreadsheet, there should be 300 leaflets downloaded. Download speed may vary depending on internet speed.

## Word Segmentation:

- Use the following commands to install the required libraries  
pip install opencv-python  
pip install roboflow



- Open the Word Segmentation.py file. In this file, we will invoke the model from RoboFlow to implement Word Segmentation. What you need to modify are:
  - Line 10's saved\_parent: This is the path where the output images will be saved.
  - Line 42's path: This is the path for the input images.
  - Line 36's api key: Refer to the RoboFlow section for this
- The path for the input images should be in the following format:
  - japanese-handwriting-analysis
  - input\_image\_folder
  - img1.jpg
  - img2.jpg
  - ...jpg
  - other\_folder
  - Word\_Segmentation.py

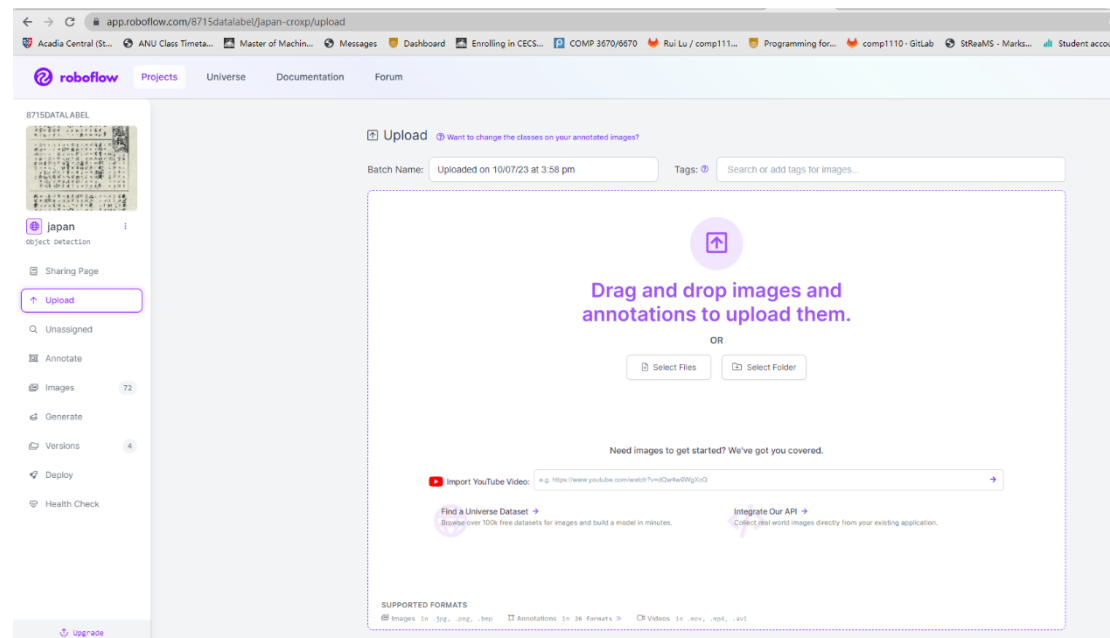


- In the left navigation pane, select the file and right-click, then choose Run

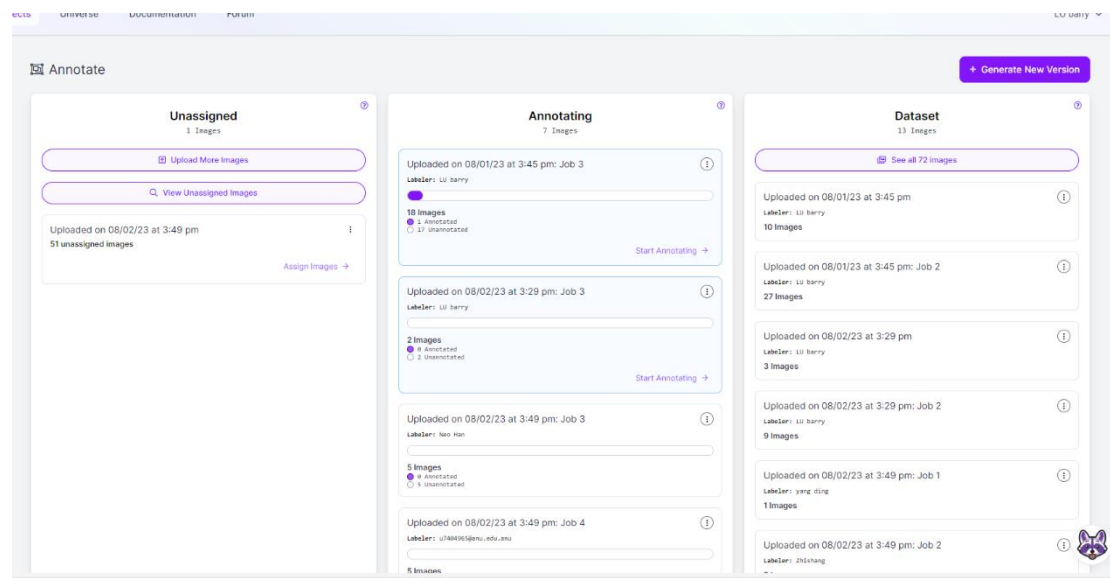
## 2: RoboFlow

RoboFlow offers an integrated platform for annotating data, training models, and deploying models. Our current RoboFlow project is nearing its API call limit. Therefore, you might consider creating a new RoboFlow project and transferring the annotated data into it. The annotated data can be found on Google Drive under the filename 'japan.v5i.yolov8'.

Train model using RobowFlow:



First, open your RoboFlow project, then go to the 'upload' page to upload your dataset



If you want to annotate more data, go to the 'annotated' page. Click on 'assign image' to allocate the images to yourself or your team members. Then, click 'start annotating'. After completing the annotations, click on 'generate new version'.

+

Generate New Version

VERSIONS

2023-08-07 11:21am  
v5 Aug 7, 2023

2023-08-07 11:03am  
v4 Aug 7, 2023

2023-08-02 7:49pm  
v3 Aug 2, 2023

2023-08-01 4:16pm  
v1 Aug 1, 2023

Generating New Version

Prepare your images and data for training by compiling them into a version.  
Experiment with different configurations to achieve better training results.

✓

Source Images

Images: 72

Classes: 1

Unannotated: 0

Edit

✓

Train/Test Split

Training Set: 50 images

Validation Set: 9 images

Testing Set: 13 images

3

Preprocessing

What can preprocessing do?

Decrease training time and increase performance by applying image transformations to all images in this dataset.

Auto-Orient

Edit

×

Resize

Stretch to 640×640

Edit

×

+

Add Preprocessing Step

Continue

Follow the instructions on this page, and then you can train your model.

• A Python script using the Roboflow SDK.

Below, we have instructions on how to use our deployment options.

Hosted API

My Machine

Luxonis OAK

Enterprise

Python

cURL

Javascript

Swift

.NET

Infer on Local and Hosted Images

To install dependencies, `pip install roboflow`

```

from roboflow import RoboFlow
rf = RoboFlow(api_key="KHZmYcucQKFOHB6wFL17")
project = rf.workspace().project("japan-croxp")
model = project.version(5).model

# infer on a local image
print(model.predict("your_image.jpg", confidence=40, overlap=30).json())

# visualize your prediction
# model.predict("your_image.jpg", confidence=40, overlap=30).save("prediction.jpg")

# infer on an image hosted elsewhere
# print(model.predict("URL_OF_YOUR_IMAGE", hosted=True, confidence=40, overlap=30).json())

```

api key

model version

Use with Snap AR's Lens Studio

Export and use this model to create custom lenses within Snap AR's Lens Studio. [Read More](#)

Roboflow Documentation

Look through our full documentation for more information and resources on how to utilize this model.

Example Web App

Use this model with a full fledged web application that has all sample code included.

Video Inference Script

Our example script performs inference on a video file with Roboflow Infer.

Deploy to NVIDIA Jetson

Perform inference at the edge with a Jetson via our Docker

After completing the training, click 'deploy', and you can then invoke your model. You can modify the content in Word Segmentation.py based on the Python code they provide.

### 3: Clustering

#### OCR branch:

See the full plan in research report on confluence [Report on OCR method progress, plan and constraints - International Collaboration on Handwriting - Confluence \(atlassian.net\)](#)

Still lacking training and testing. Only finished finding OCR and use the model.

The result of seg\_letter is in ocr\_output.xlsx, if you want to use the model again. You can go to clustering/google\_ocr.ipynb. Run every cell except last one, remember to change the input and output directory! The seg\_letter is on google drive so probably download it to your computer and lead the path to it.

Ask Zhishang Bian if you have any questions.

#### Cluster with K-means:

This is work from semester 1 2023. New methods don't use it now. You probably don't need it

Open with Pycharm or Jupyter lab.

Jupyter lab: Run every cell and it will automatically download the library you need

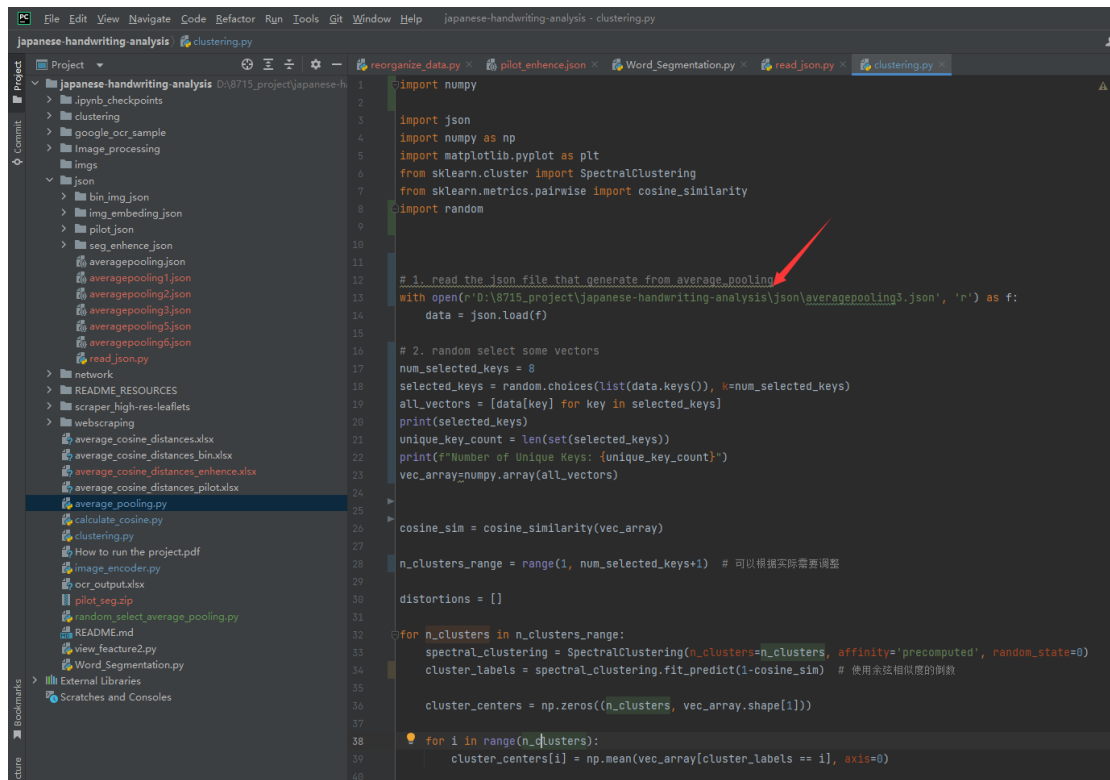
Pycharm: Run main() and it will automatically download the library you need

Found it here:

<https://gitlab.cecs.anu.edu.au/u7434576/japanese-handwriting-analysis/-/tree/main/clustering>

#### Spectral clustering:

First, run the Word Segmentation.py, image\_encoder.py, averagepooling.py



```
1 import numpy
2
3 import json
4 import numpy as np
5 import matplotlib.pyplot as plt
6 from sklearn.cluster import SpectralClustering
7 from sklearn.metrics.pairwise import cosine_similarity
8 import random
9
10
11
12 # 1. read the json file that generate from average_pooling
13 with open("D:\8715_project\japanese-handwriting-analysis\json\averagepooling3.json", 'r') as f:
14     data = json.load(f)
15
16 # 2. random select some vectors
17 num_selected_keys = 8
18 selected_keys = random.choices(list(data.keys()), k=num_selected_keys)
19 all_vectors = [data[key] for key in selected_keys]
20 print(selected_keys)
21 unique_key_count = len(set(selected_keys))
22 print(f"Number of Unique Keys: {unique_key_count}")
23 vec_array = numpy.array(all_vectors)
24
25
26 cosine_sim = cosine_similarity(vec_array)
27
28 n_clusters_range = range(1, num_selected_keys+1) # 可以根据实际需要调整
29
30 distortions = []
31
32 for n_clusters in n_clusters_range:
33     spectral_clustering = SpectralClustering(n_clusters=n_clusters, affinity='precomputed', random_state=0)
34     cluster_labels = spectral_clustering.fit_predict(1-cosine_sim) # 使用余弦相似度的倒数
35
36     cluster_centers = np.zeros((n_clusters, vec_array.shape[1]))
37
38     for i in range(n_clusters):
39         cluster_centers[i] = np.mean(vec_array[cluster_labels == i], axis=0)
```

Modify the path on line 13. You can generate the json file yourself or use the one in gitlab. Then you can run the program

#### 4: Image encoder:

Install the following package pytorch, pil,clip

The segamentaed image can be downloaded from here:

<https://drive.google.com/drive/folders/1hecluxT4ad-6tQBP7q64SHWG3DRKsFKW>

seg\_letter.zip is the leaflets. Pilot\_seg.zip is the test data. You can ou can unzip them in the root directory.

```
data.py × siamese.ipynb × average_pooling.py × image_encoder.py × pilot_enhance.json × Word_Segmentation.py × read_json.
def get_image_features(model, preprocess, folder_path):
    image_features_dict = {}

    for subdir, _, files in os.walk(folder_path):
        for file in files:
            if file.endswith(('png', 'jpg', 'jpeg')):
                # get_path
                full_path = os.path.join(subdir, file)

                # use clip to get the features
                image_tensor = preprocess(Image.open(full_path)).unsqueeze(0).to(device)
                with torch.no_grad():
                    image_features = model.encode_image(image_tensor)
                    image_features = image_features.cpu().numpy().tolist()

                # save feature vector
                image_features_dict[full_path] = image_features[0]

    return image_features_dict

if __name__ == "__main__":
    device = "cuda" if torch.cuda.is_available() else "cpu"
    model, preprocess = clip.load("ViT-B/32", device=device)

    folders = {
        r"D:\8715_project\japanese-handwriting-analysis\seg_letter_enhance": r"D:\8715_project\japanese-handwriting-ana
    }

    for input_folder, output_folder in folders.items():
        image_features_dict = get_image_features(model, preprocess, input_folder)
        # save to json format
        if not os.path.exists(output_folder):
            os.makedirs(output_folder)
        output_file = os.path.join(output_folder, os.path.basename(input_folder) + ".json")
        with open(output_file, 'w') as f:
            json.dump(image_features_dict, f, indent=4)
```

Modify the path on line 33. Then you can run it.

## 5: AveragePooling



```
9
10
11 1 usage  Rui *
12 def process_json(input_path, output_path):
13     with open(input_path, 'r') as f:
14         data = json.load(f)
15
16     # 结果数据初始化
17     result_data = {}
18
19     # key is the path of the img, values are the vectors
20     dir_vectors = {}
21     for img_path, vector in data.items():
22
23         dir_name = os.path.basename(os.path.dirname(img_path))
24
25         if dir_name not in dir_vectors:
26             dir_vectors[dir_name] = []
27
28         dir_vectors[dir_name].append(vector)
29
30     for dir_name, vectors in dir_vectors.items():
31         # select 30 vectors
32         selected_vectors = vectors[:30]
33
34         avg_vector = average_pooling(selected_vectors)
35
36         result_data[dir_name] = avg_vector
37
38     # save to json format
39     with open(output_path, 'w') as f:
40         json.dump(result_data, f)
41
42
43
44
45 input_path = 'D:\8715_project\japanese-handwriting-analysis\json\pilot_json\pilot_enhence.json'
46 output_path = 'json/averagepooling3.json'
47 process_json(input_path, output_path)
48
```

Modify the path on line 45 and 46. Then run the file.