



Data Glacier

Your Deep Learning Partner

Resume Extraction project.

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Executive summary:

The Client:

HR departements.

Data Set:

A json file containing 200 resumes.

Dataset link: <https://gist.github.com/Rahulrky/b57ad459545c896231c4b770bd8d22ef>

General Description :

This is flask web app used for resumes annotation and parsing.

Product-Service Description:

This app is dedicated to HR managers to help them:

- Converting hours of labor into seconds.
- Increase recruiters' efficiency and availability.
- Reducing the need for more employees.
- Avoiding errors.

Data science objectives:

- Identifying the suitable technologies for our business objectives.
- Training and deploying fast and efficient Deep Learning models.

Problem statement :

Problem:

Resumes contain surfeit information that is not relevant for the HR/authority, and they have to manually process the resumes to shortlist the promising candidates for them. This EDA is dedicated to them to help them enhance their performance.



Data comprehension :

The data that has been provided is a json file.

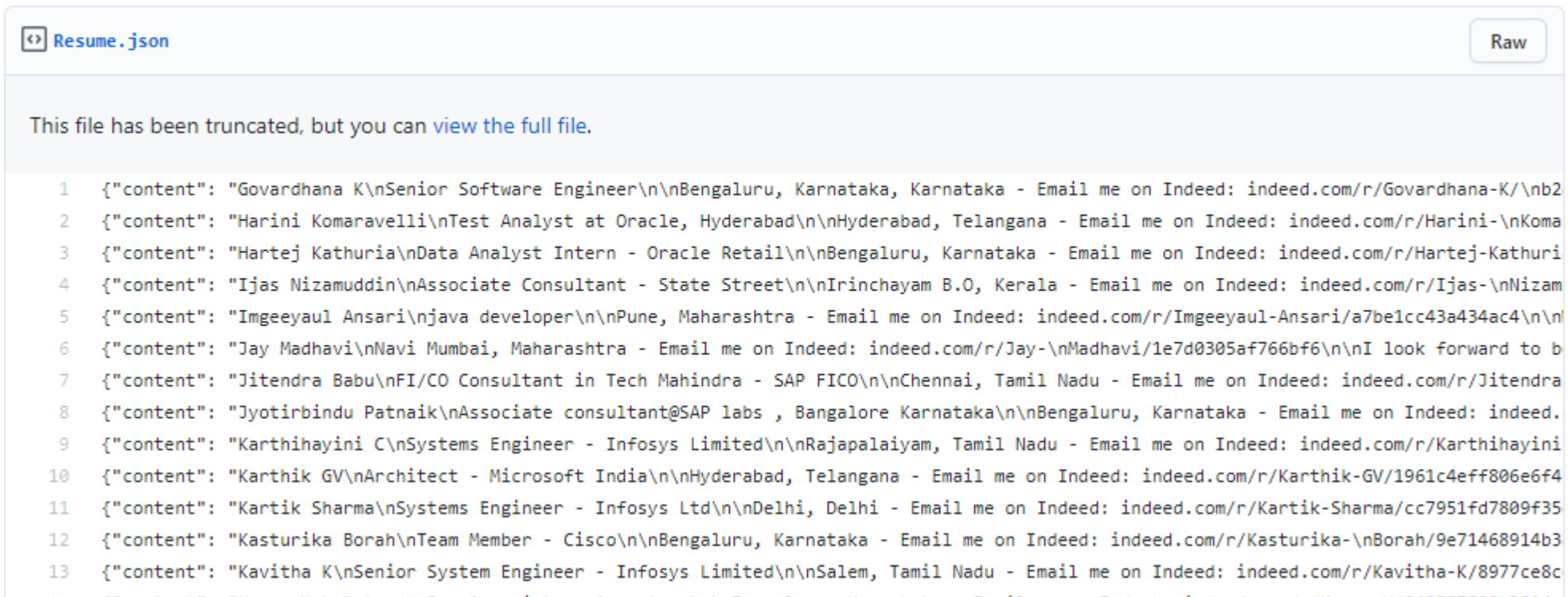
Its s composed of 200 resumes. Each line represents a 2 keys dictionary:

“annotation” is the key to the labeled resume.

“content”: is the key to the plain resume text.

Each resume feature is represented with a dictionary: `dict_keys(['label', 'points'])`

'points' is the key to a dictionary that looks like this: `[{'start': 1749, 'end': 1754, 'text': 'Oracle'}]`



Data preparation :

Steps:

- Importing the data and transforming it into a dataframe.
- Getting rid of the trailing white spaces from the plain text resumes
- Putting entities of the annotation into lists composed of : label, starting position , ending position.
- Getting rid of trailing spaces and unnecessary punctuation from the entities .
- Tokenization.
- Splitting the data into testing and training sets.
- Transforming the data to be fed to the model to torch tensors.

```
[ ] for i in range(len(data)):
    tokenized_texts = tokenizer.texts_to_sequences(data["content"])

    print(tokenized_texts[0])
```

```
[1979, 893, 187, 49, 60, 168, 73, 73, 62, 57, 8, 11, 11, 13, 16, 1979, 893, 3170, 569,
```

EDA

Analysis plan:

- Graduation year analysis.
- Candidates location analysis.
- Companies Candidates worked at.
- Candidates degrees and universities studied at analysis.

EDA summary:

- The EDA performed on our data has shown that the candidates who have applied are quite different, they don't share to the same field of study nor the same degrees. They went to different colleges and they don't have the same skills.
- Depending on these insights, the client's HR department can request the elimination of some candidates' categories depending on the job profile needed (example: the client needs candidates with engineering degree)

Recommendation:

This EDA has shown that those who applied for this job are quite different especially when talking about the fields of study, I recommend that the HR department takes more care of the job description and the requirements provided in order to have a more accurate candidates resumes.

Modeling and evaluation:

Depending on the data we have, our business and data science objectives I BERT model to work with in order to build the resume extraction.

Bidirectional Encoder Representations from Transformers (BERT) is a transformer-based machine learning technique for natural language processing (NLP) pre training developed by Google.

Epochs: 6

Optimizer:Adam

As we can see the accuracy we had is : 0.9138

F1- score: 0.94

```
Epoch: 0%|          | 0/5 [00:00<?, ?it/s]Train loss: 0.8996409103274345
Epoch: 20%|██        | 1/5 [00:11<00:46, 11.65s/it]Validation loss: 0.5537128895521164
Validation Accuracy: 0.9138541666666666
F1-Score: 0.9417320178767181
Train loss: 0.5108052641153336
Epoch: 40%|████      | 2/5 [00:23<00:35, 11.67s/it]Validation loss: 0.517583355307579
Validation Accuracy: 0.9138541666666666
F1-Score: 0.9417320178767181
Train loss: 0.4752659574151039
Epoch: 60%|██████    | 3/5 [00:35<00:23, 11.75s/it]Validation loss: 0.4547186344861984
Validation Accuracy: 0.9138541666666666
F1-Score: 0.9417320178767181
Train loss: 0.4253872831662496
Epoch: 80%|████████  | 4/5 [00:47<00:11, 11.83s/it]Validation loss: 0.42342111468315125
Validation Accuracy: 0.9171875
F1-Score: 0.9430812041487477
Train loss: 0.40889669706424076
Epoch: 100%|██████████| 5/5 [00:59<00:00, 11.90s/it]Validation loss: 0.4165296256542206
Validation Accuracy: 0.9138541666666666
F1-Score: 0.9417320178767181
```


Deployment:

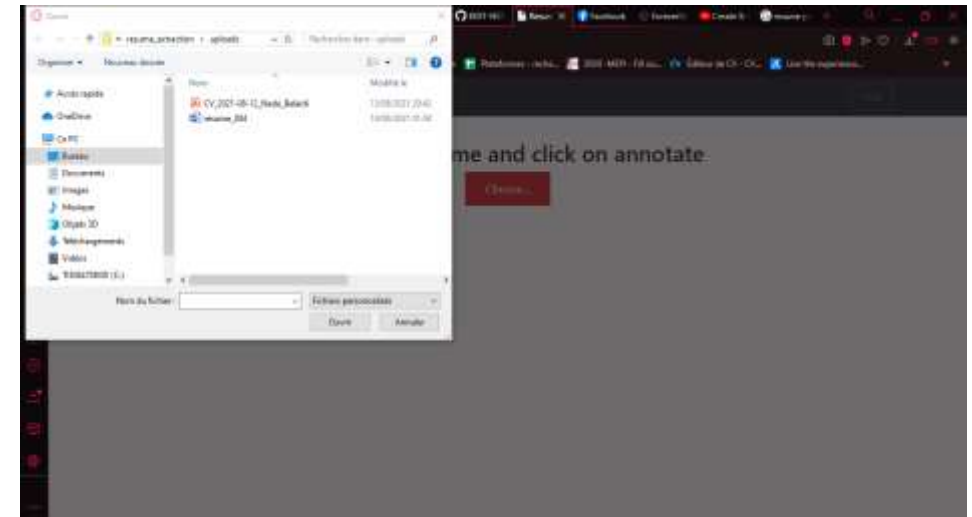
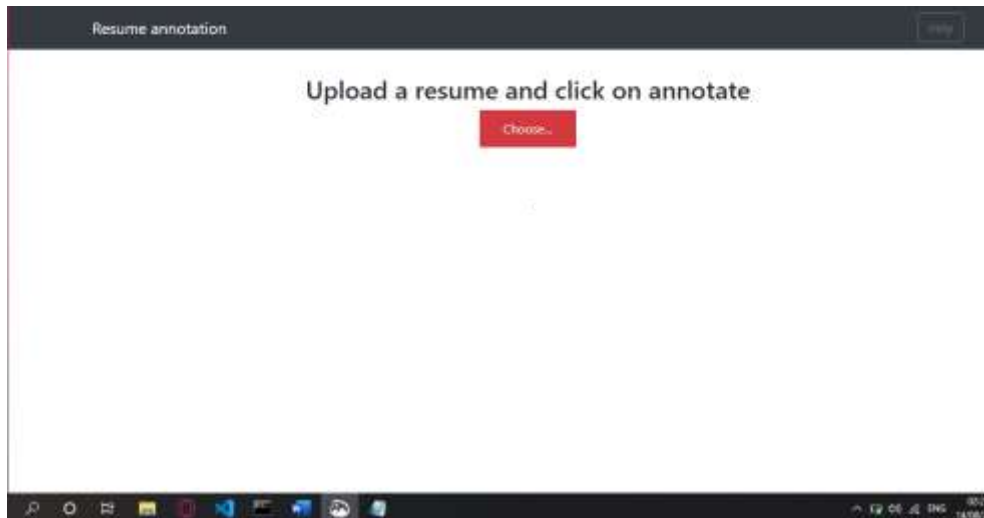
I used Flask, an open-source micro framework for web development in Python.

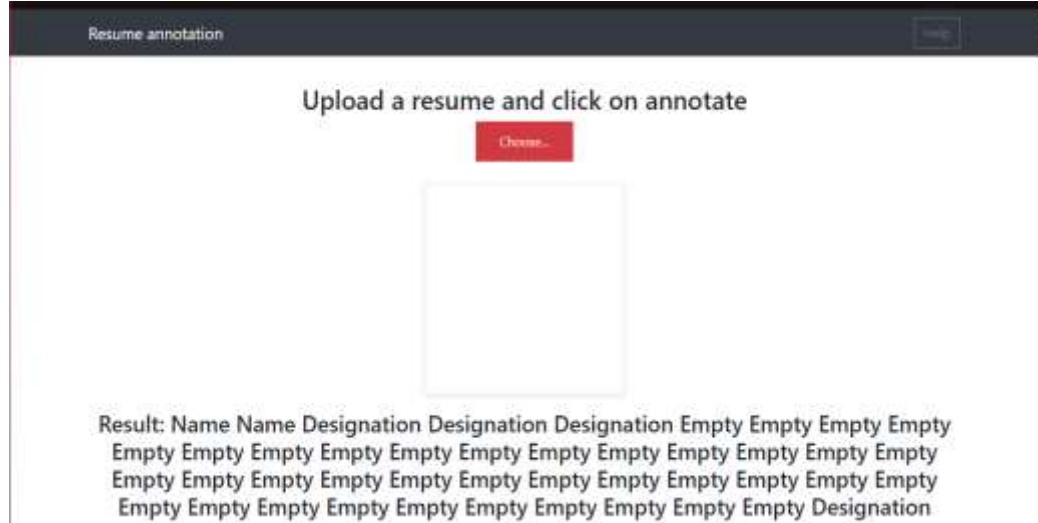
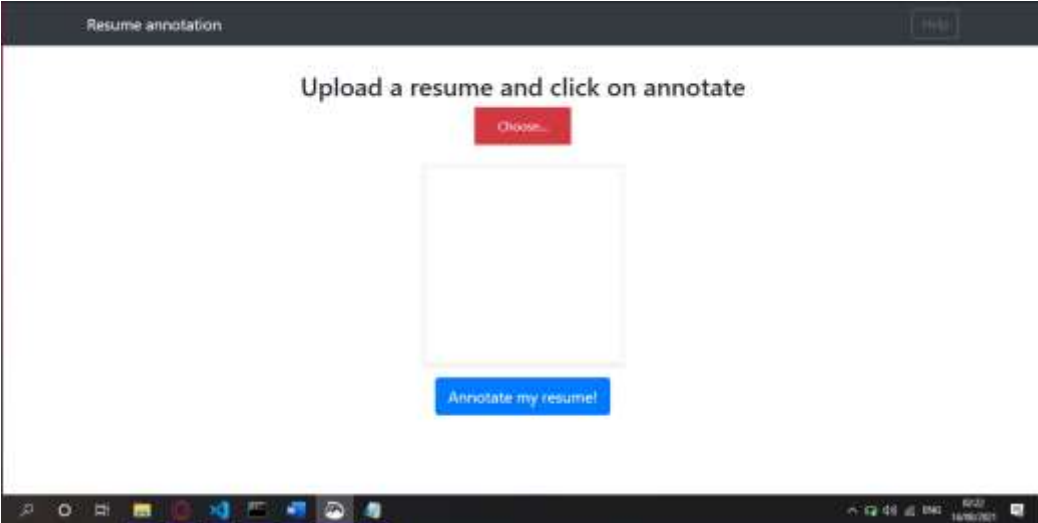


I started with creating additional functions dealing with mining text from files in different Formats (pdf, docx,json..)

I used HTML,CSS and JS for the front-end part.

This is a quick demo:





Conclusion:

This was my NLP project that I worked on while being an intern at Data Glacier during one month. This is an app that will make it easier for HR employer to find the perfect candidate for the job in less time while avoiding errors and with less employees needs.

GitHub Repo link: <https://github.com/NadaBelaidi/NLP-Resume-Extraction>

Thank You