Data Mining & Machine Learning Project Resume Classification

Daniele Laporta







Project steps

- **1. Design:** Dataset description, preprocessing, EDA and feature transformation, sentiment score, word embeddings.
- **2. Model Implementation:** Resume binary classification (STEM/ not STEM) with stratified k-fold cross validation.
- 3. Classification Results
- 4. Conclusions
- 5. Possible Improvements
- 6. Additional study

Developed with python on coldb.







Dataset Description

Resumes dataset from **Kaggle**:

real people linkedin resumes publicly available. Each resume has many categorical features and a ground truth, 'category', which is the job sector.

category	linkedin	profile picture	description	Experience	Name	position	location	skills	clean skills
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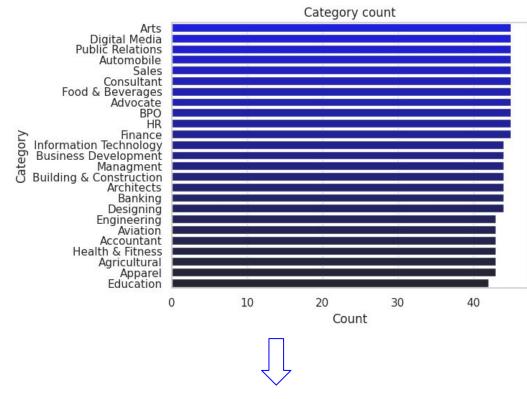
(Initial dataset: 1251 rowsx 11 columns \rightarrow Final dataset: 1103 rows x 9 columns)





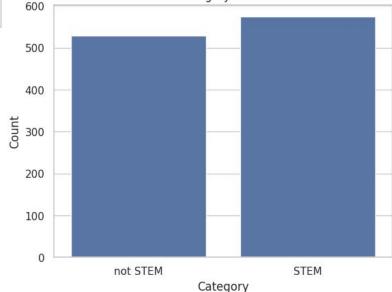


EDA



25 classes, 40 samples per class. Not enough data to perform multi classification!

For simplicity, I reduced the ground truth to 2 classes:
STEM/ not STEM



Category Count





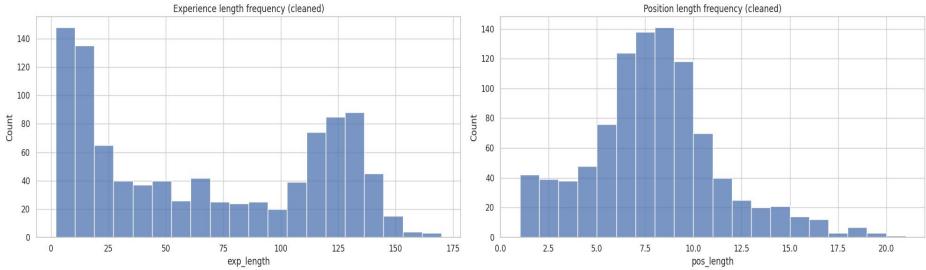


EDA

'skills' Word Cloud

'experience' and 'position' length frequency



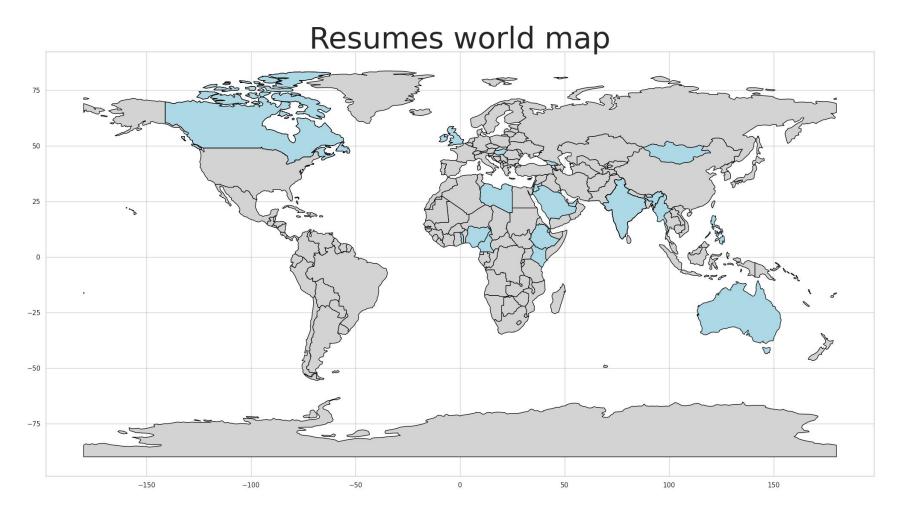








EDA



* not all countries are recognised in the map.



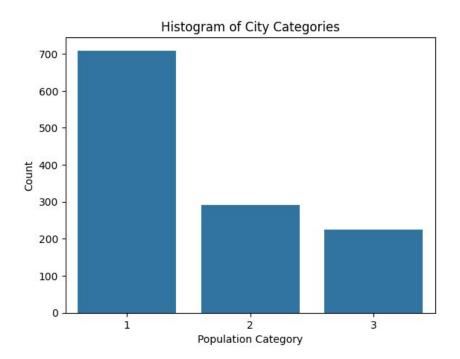




Feature Transformation

The following are performed on all the dataset.

- Transform 'location' column
 I used a dataset (Geonames) of around
 147k cities with their population size.
 The aim is to substitute the cities in my
 Kaggle dataset with a discretization of
 the population in terms of:
 - population <1M: 'small' (1)
 - 1M ≤population<5M: <u>'medium' (2)</u>,
 - population ≥5M: <u>'big' (3).</u>



- Sentiment score for 'experience', 'position' and 'skills' columns using 2 pre-trained Hugging Face models, specific for job skills and descriptions.

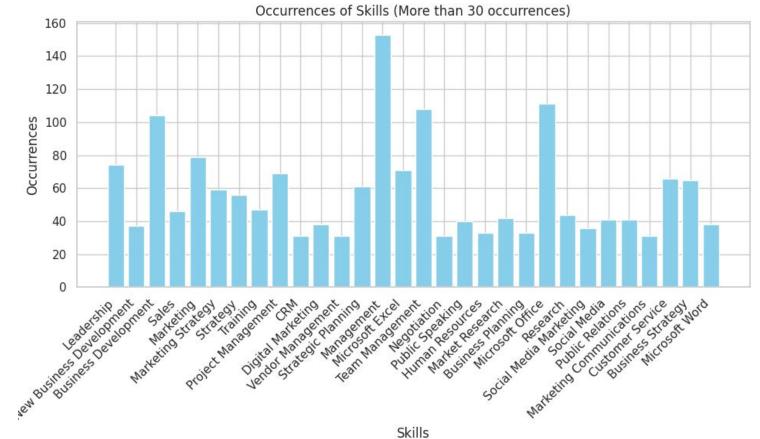






Feature Transformation

- Transform (weight) 'skills' column: If a skill appears more times, it becomes less important for the model. A skill that appears very few times can be considered rare.

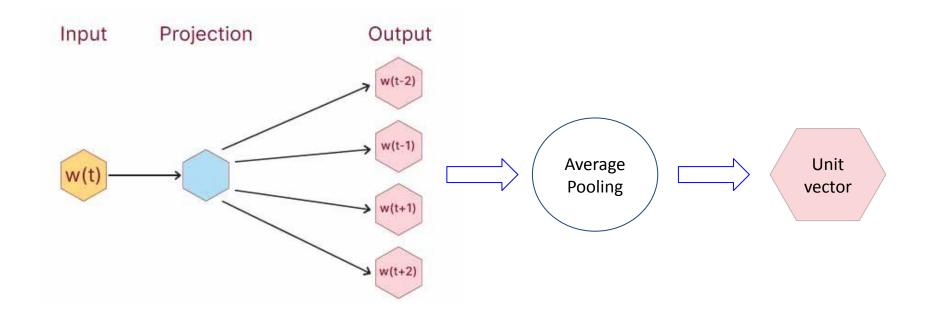








Word Embeddings



- Word2Vec, Doc2Vec and BERT for the 'skills' feature.
- Word2Vec, Doc2Vec for the 'experience' feature.
- Similar words → similar embeddings.







Classification Results

- Stratified K-Fold Cross Validation (K=5). 2 classes: STEM (1) / not STEM (0)
- Dataset <u>basic</u> configuration: 9 features: 'exp_length', 'pos_length', 'population_category', 'skills_scores_1', 'skills_scores_2', 'position_scores_1', 'position_scores_2', 'experience_scores_1', 'experience_scores_2'.
- Dataset <u>+3we</u> configuration: same 9 features + 3 word embeddings on 'skills' feature.
- Dataset <u>+5we</u> configuration: +3we + 2 word embeddings on 'experience' feature.

Dataset Configuration	Averaged metrics	LOG REG	RF	SVC	MULTIN. NB	XGBOOST
basic		0.60 0.62 0.63 0.62	0.60 0.62 0.59 0.60	0.61 0.62 0.68 0.64	0.56 0.55 <mark>0.85</mark> 0.67	0.59 0.61 0.57 0.59
+3we	Accuracy Precision Recall F1 Score	0.59 0.60 0.68 0.63	0.59 0.61 0.59 0.60	0.59 0.59 0.71 0.64	0.57 0.56 <mark>0.82</mark> 0.66	0.58 0.59 0.65 0.61
+5we		0.66 0.68 0.68 0.67	0.67 0.72 0.61 0.63	0.66 0.68 0.71 <mark>0.68</mark>	0.58 0.57 <mark>0.80</mark> 0.66	0.67 0.70 0.66 0.67







Possible Improvements

- Data augmentation or generally more data at disposal (E.g. salary, cost and quality of living).
- Focus on a niche job market (additional study).
- Study historical and evolving trends of required skills per job.
- Develop an application for resume-job matching.

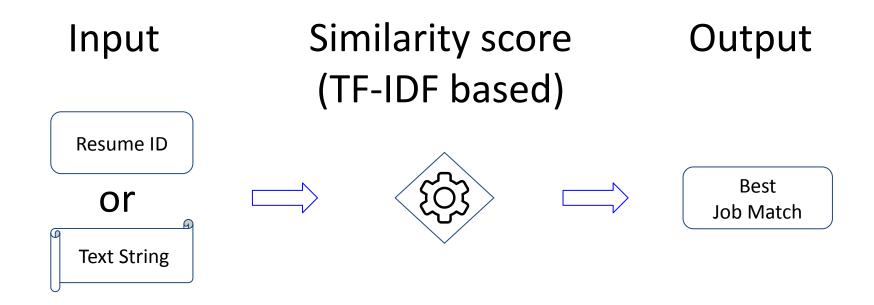






Additional Study: Job Recommender

Source: Job postings dataset from <u>Kaggle</u> (data science and tech job offers)









Thank you for the attention!





