# CS-GY 6513 Project Report: Resume Matching using Text Processing Techniques

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#### 1. Introduction

In recent years, text processing techniques has become very useful to solve some practical problems. For example, text classification, text searching, sentiment analysis, recommending system etc. After transferring texts into vector space, we can use multiple techniques to find some properties of the texts or quickly searching some crucial information of the given requests. These techniques include word embedding, N-Gram, TF-IDF or base on some machine learning model such as SVM, K-Means, Random Forests etc.

In this project, we aim to match a given resume to some job information which were crawled from a job hunting website. This will help the job-hunters to find some jobs which is most related to their skills, background and experience. We also plan to do some analysis on the data that collected from the website, for example, what kind of techniques are needed most in a job? The location distribution of job demanding etc. We will visualize the results and find some interesting patterns of these data.

#### 2. Methods

The data for the project is gathered from a job hunting website http://www.monster.com. This website provides us with tons of job information by querying the job and preferred working location. In the project, we collect the information of 20 kinds of job. By searching one job we can get tens of thousands of records. Gather detailed information and store them into .csv file and MySQL in Azure.

Crawling in python use modules of requests, lxml and multi-threading. The website of Monster is using Ajax. Thus we simulate the GET response with the

headers. There will be 25 records in one page. Traversing all the pages it come out can get a list of job detail links. Access all the urls and enter second-level page. Using xpath to match the elements such as job title, company name, working location, detailed description etc. Clear the data. And store them in different tables named by the jobs in an Azure MySQL database. Write a client to query the table base on location, job, company, description.

After gathering and storing the data, we can use some text processing techniques to finish our tasks.

We analyse word and document frequency using a technique called TF-IDF. TF stands for term frequency, that is how frequently a word occurs in a document. There are words in a document, however, that occur many times but may not be important. in English, these are probably words like "the", "is", "of", and so forth. We might take the approach of adding words like these to a list of stop words and removing them before analysis, but it is possible that some of words might be more important in some documents than others. To distinguish these words, we use another method called inverse document frequency (IDF), which decreases the weight for commonly used words and increases the weight for words that are not used very much in a collection of documents.

In practical, we define a new vector space representation. For document i, we construct a vector  $x_i$  such that the j-th coordinate is given by:

$$x_i(j) = tf_i(j) \cdot idf(j)$$

The term  $tf_i(j)$  represents term-frequency, which counts the number of occurrences of word j in document i. The term idf(j) represents inverse-document frequency, and is defined as follows.

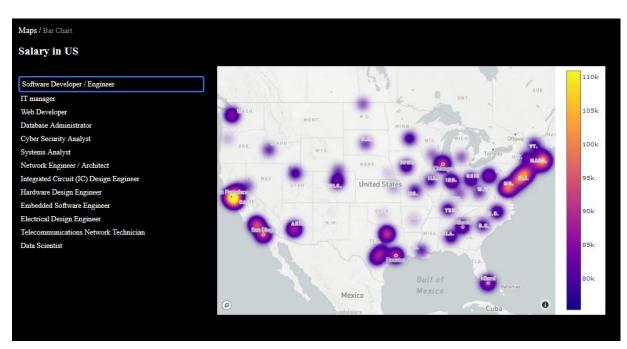
1

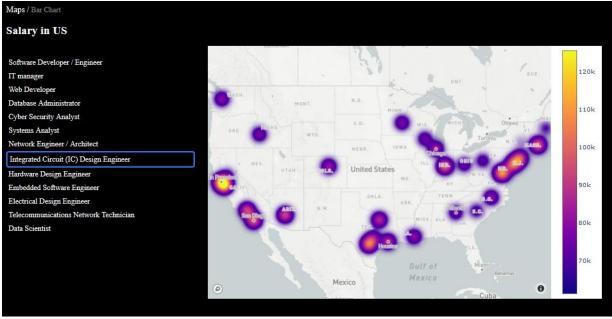
Let  $n_j$  be the number of documents in the database which contain at least one occurrence of word j. Then,

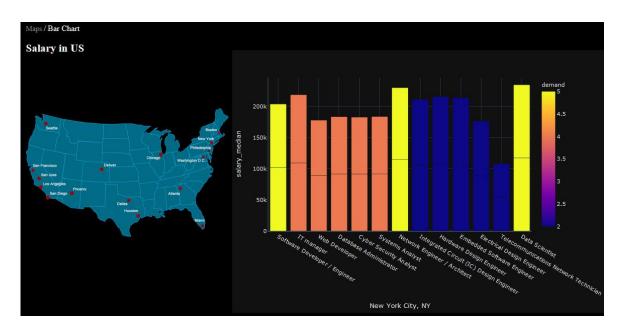
$$idf(j) = log \frac{n}{n_j + 1}$$

Finally, we calculate the cosine-similarity between query vector and the TF-IDF matrix which is defined above.

## 3. Visualization









## 4. Results

In this project, all the jobs we get from the websites are related to field in Electronic and Computer Engineering(ECE) or Computer Science(CS). Finally we use about 20,000 jobs for matching the resume, and the resume we used are also from ECE major or CS major.

For implementing the resume matching system, we use PySpark in Python, which is a great language for performing exploratory data analysis at scale, building machine learning pipelines, and creating ETLs for a data

platform. The first benefit of using PySpark is that it has an in-memory computation architecture, which helps us increase the speed of processing. Secondly, Being dynamic in nature, it helps us to develop a parallel application, as Spark provides 80 high-level operators. For these reasons, data scientists always use it to build a recommendation system or machine learning system.

These are some jobs recommended to a resume which shows a solid background in the field of Machine Learning:

job	company	location	similarity
Machine Learning	CyberCoders	New York, New Yor	0. 23400453719699693
omputer vision e	Brambles USA Inc	Orlando, Florida	0.23260229031089125
Machine Learning	Brambles USA Inc	Orlando, Florida	0.23260229031089125
data-scientist	Microsoft Corpora	Redmond, Washingt	0.23218043070360098
ython Software E	CHEP	Orlando, Florida	0.2318661081229323
omputer vision e	CHEP	Orlando, Florida	0.2318661081229323
Machine Learning	CHEP	Orlando, Florida	0.23161338123623557
data-scientist	Randstad	Raleigh, North Ca	0.2265064669531539
data-scientist	ServiceNow, Inc.	Kirkland, Washing	0.22312532521787307
data-scientist	SPECTRUM	Golden, Colorado	0.21862107413506884
Spark Engineer	SPECTRUM	Englewood, Colorado	0.21862107413506884
Machine Learning	SPECTRUM	Wheat Ridge, Colo	0.21862107413506884
FPGA Engineer	SPECTRUM	Pine, Colorado	0.21862107413506884
NLP engineer	SPECTRUM	Englewood, Colorado	0.21862107413506884
data-scientist	Paycom	Oklahoma City, Ok	0.21478529137798136
Machine Learning	Paycom	Oklahoma City, Ok	0.21478529137798136
data-scientist	Randstad Technolo	San Francisco, Ca	0.21284766243301098
data-scientist	Exelon	OAK BROOK, Illinois	0.21057569022732878
NLP engineer	Trexquant Investment	Stamford, Connect	0.20802822570551724
NLP engineer	Tailored Management	South San Francis	0.20401481468053828

(1) As is shown above, the most similar jobs are Machine Leaning, Computer Vision Engineer and Data Scientist .etc, which are all in the field of Machine Learning.

The following are all job recommendations for different kinds of resumes:

job	company	location	similarity	
SQL Developer	AbleForce, Inc.	  SAN DIEGO, Califo	0.264033861383034	
Database Engineer	AbleForce, Inc.	SAN DIEGO, Califo	0.264033861383034	
SQL Developer	Tekmark Global So	Ohio	0.26021566245042727	
SQL Developer	Accede Solutions Inc	Buffalo Grove, Il	0.23786697071082044	
SQL Developer	Shulman Fleming a	Jersey City, New	0.23768455539319813	
Database Engineer	Shulman Fleming a	Jersey City, New	0.23768455539319813	
Python Software E	Shulman Fleming a	Jersey City, New	0.23768455539319813	
database administ	AbleForce, Inc.	SAN DIEGO, Califo	0.2314174228770613	
IT manager	Zachary Piper LLC	Norristown, Penns	0.22206289327132797	
SQL Developer	Questa Technology	Durham, North Car	0.2212274635445375	
Database Engineer	Randstad Technolo	Brea, California	0.20937696887132082	
SQL Developer	Randstad Technolo	Brea, California	0.20937696887132082	
database administ	Netsource, Inc.	Bellevue, Washington	0.20896717872484252	
Database Engineer	Randstad Technolo	Phoenix, Arizona	0.1854604088665849	
SQL Developer	Randstad Technolo	Phoenix, Arizona	0.1854604088665849	
database administ	The Maxis Group	Phoenix, Arizona	0.18379570062459383	
python	Shulman Fleming a	Jersey City, New	0.18032984907850536	
SQL Developer	Randstad	Brea, California	0.17606677224380562	
Database Engineer	Randstad	Brea, California	0.17606677224380562	
Spark Engineer	Zachary Piper LLC	Horsham, Pennsylv	0.1749505043029096	

(2) Resume for a Database Engineer employee.

Top 20 matched jobs:

similarity	location	company	job
. 19837398245555088	Charlotte, North	Gulfstream Strate	Electrical Design
. 19837398245555088	Raleigh, North Ca	Gulfstream Strate	Electrical Design
. 19837398245555088	Jacksonville, Flo	Gulfstream Strate	Electrical Design
. 19837398245555088	Jacksonville, Flo	Gulfstream Strate	Electrical Design
. 19837398245555088	Raleigh, North Ca	Gulfstream Strate	Electrical Design
. 19837398245555088	Sacramento, Calif	Gulfstream Strate	Electrical Design
. 19837398245555088	Charlotte, North	Gulfstream Strate	Electrical Design
. 19837398245555088	Sacramento, Calif	Gulfstream Strate	Electrical Design
. 16304672360326572	Cincinnati, Ohio	Surf Search	FPGA Engineer
. 16304672360326572	Cincinnati, Ohio	Surf Search	Electrical Design
. 15299152687517378	NEW BEDFORD, Mass	Northern Wind, Inc.	electrical test
0.1504037889182138	Columbus, Ohio 43235	Synerfac Technica	Electrical Design
. 13440340264780684	Erie, Pennsylvania	Hunt, Guillot and	PLC Technician
0.1342985896378198	Huntsville, Alaba	Tech USA	electrical test
. 13388333096894836	SOUTH EL MONTE, C	Sterling Machiner	electrical test
. 13380706384992894	Hampstead, Maryla	Synerfac Technica	Electrical Design
. 13007857259261205	FLAGSTAFF, Arizon	Lowell Obervatory	electrical test
	Dallas, Texas 76001		Electrical Design
. 12694260039065755	Hope Hull, Alabam	Spherion	electrical test
0.1263053518144696	Perth Amboy, New	Randstad	electrical test

## (3) Resume for a Electrical Engineering major employee.

Top 20 matched jobs:

similarity	location	company	job
  0.08116147399903528	San Antonio, Texas	Southwest Researc	Machine Learning
0.08064671429779627	San Jose, California	Microchip Technology	FPGA Engineer
0.08037405195916836	Fremont, Californ	Lam Research Corp	FPGA Engineer
0.08037405195916836	Fremont, Californ	Lam Research Corp	DSP engineer
0.07606167591026433	Ashaway, Rhode Is	Randstad	FPGA Engineer
0.071527836903023	San Antonio, Texas	Southwest Researc	Embedded Systems
0.06983257516394467	San Jose, California	ON Semiconductor	DSP engineer
0.06794424029819622	Lexington, Massac	Odyssey Systems C	FPGA Engineer
0.06794424029819622	Lexington, Massac	Odyssey Systems C	Embedded Systems
0.06794424029819622	Lexington, Massac	Odyssey Systems C	DSP engineer
0.06794424029819622	Lexington, Massac	Odyssey Systems C	ARM engineer
0.06712068324775063	Ashaway, Rhode Is	Randstad Technolo	Circuit Design En
0.06712068324775063	Ashaway, Rhode Is	Randstad Technolo	FPGA Engineer
0.06575373417683203	Huntsville, Alabama	Dynetics	FPGA Engineer
0.06496953463692742	San Antonio, Texas	Southwest Researc	Machine Learning
0.06496953463692742	San Antonio, Texas	Southwest Researc	computer vision e
0.06496923964857998	Irvine, Californi	CyberCoders	computer vision e
0.06496923964857998	Irvine, Californi	CyberCoders	Python Software E
0.0647745018675982	San Antonio, Texas	Southwest Researc	Machine Learning
0.0647745018675982	San Antonio, Texas	Southwest Researc	computer vision e

(4) Resume for a Computer Engineering major employee.

In addition, we can also use SQL query in PySpark to find some recommended jobs in a specific city, you can choose any big city you like in USA, for example, the recommended jobs in New York City for a CE major student is shown below:

similarity	cation	10		l	company	job
0.04988706287672642	Ne	City,	York	New	Clarapath Inc.	Telecommunication
0.04988706287672642	Ne	City,	York	New	Clarapath Inc.	FPGA Engineer
0.035548372632214956	Ne	City,	York	New	Synechron	python
0.034945310902609356	Ne	City,	York	New	Clarapath Inc.	Telecommunication
0.034945310902609356	Ne	City,	York	New	Clarapath Inc.	web developer
0.03454556273853086	Ne	City,	York	New	Synechron	Database Engineer
0.03454556273853086	Ne	City,	York	New	Synechron	Python Software E
0.03454556273853086	Ne	City,	York	New	Synechron	Architect
0.03115421079601298	Ne	City,	York	New	Technovision	Software Product
0.029346975702960926	Ne	City,	York	New	ness Informat	database administ
0.027419386578140224	Ne	City,	York	New	Hudson Inter	Test Automation E
0.02657937467733803	Ne	City,	York	New	CloudFlare	DSP engineer
0.023481358658936607	Ne	City,	York	New	Grow	data-scientist
0.023481358658936607	Ne	City,	York	New	Grow	Machine Learning
0.021909819169386807	Ne	City,	York	New	Apex Systems	Spark Engineer
0.020684520483680347	Ne	City,	York	New	NPD Group, Inc.	data-scientist
0.02004689255561453	Ne	City,	York	New	Data Inc	Java Software Eng
0.019200753034171885	Ne	City,	York	New	F PETERS ASSO	Data Analyst
0.01902748198912478	Ne	City,	York	New	Case Interactive	JavaScript Developer
0.01902748198912478	Ne	City,	York	New	Case Interactive	Python Software E

Another SQL query example is to find the rank of a specific job given your resume. For example, a CE major student could find some similar job information for Embedded Systems Engineer:

		job	company	location	similarity
Embedded S	Systems .		Southwest Researc	San Antonio, Texas	0.11408710602803729
Embedded S	Systems .		US ARMY Ground Ve	WARREN, Michigan	0.10491342174520059
Embedded S	Systems .		Abbott	Alameda, California	0.09937305121916463
Embedded S	Systems .		Talentlab	OTTAWA, ON	0.09733485801632007
Embedded S	Systems .		Oculii Corp.	Beavercreek, Ohio	0.0955485982066785
Embedded S	Systems .		Kumu Networks	Sunnyvale, Califo	0.0954019401067412
Embedded S	Systems		D3 Engineering	Rochester, New Yo	0.09532752286044031
Embedded S	Systems			Santa Clara, Cali	
Embedded S	Systems		CyberCoders	San Antonio, Texas	0.08739002114928836
Embedded S	Systems		Southwest Researc	San Antonio, Texas	0.08734039335372538
Embedded S	Systems .		Neteffects	Creve Coeur, Miss	0.0839969616373131
Embedded S	Systems .		Novanta	North Syracuse, N	0.08396024753874272
Embedded S	Systems .		CyberCoders	Hayward, Californ	0.0820860831573231
Embedded S				Colorado Springs,	그 그 그 집에 나가 있는데 하지만 하지만 하시네요?
Embedded S	Svstems .		IntelliPro Group		
			GCR Professional		Q (2)
			Enphase Energy	네는 그리아 100 100 100 100 100 100 100 100 100 10	[1] [1] [1] [1] [1] [1] [1] [1] [1] [1]
	3 <u>-</u> 71001 (1710)		CyberCoders	형	그러는 그렇게 되어 있다는 것이 없는 나를 가지 않는데 살아 있다.
Embedded S	·			Annapolis Junctio	구 ^ ' 이번 ^ ' ' 전에 ' ' 이번 가게 되었다. 그렇게 되었다. 그렇게 되었다.
Embedded S	379			Brighton, Massach	(i) (ii) (ii) (ii) (iii)

Finally, for the running time of this resume matching system, it will take about 30 seconds to finish the recommendation for you, and the query time is within 10 seconds. This shows the benefit of using Spark to process data.

#### 5. Further Work

There are so many ways to improve our project and a lot of additional work we can do to make this project more complex and wonderful. For example, we can add more different kinds of jobs to our SQL system

to match more resumes from other majors. In this case, we should modify our code to process huge amount of data in short time. Currently the volume our data is not so big and the processing time need to be speed up as well.