

Assignment 1 Report

Topic:

Fintech Hiring trends in the largest banks in the US

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Objective:

Financial institutions in the US are changing rapidly. Changing business models and the technological revolution has fueled the growth of a new breed of financial products and services collectively known as Fintech. With changing demographics, automation efforts and demand for new products and services, large financial institutions are realizing the power of technologies like data science, AI, cloud technologies and machine learning and are heavily investing to upgrade their technological platforms to cater to the upcoming revolution. Technology has been a key player in helping drive this revolution. In a 2017 report by CB Insights more than 46% of the job openings were in the technology section. Things are fast evolving and as we enter 2019, it is interesting to understand the hiring trends in the top financial institutions in the US. Our goal in this case study is to conduct a study on the job openings in the top US Banks in the United States and analyze trends in the industry particularly in the area of Fintech.

Dataset:

We have to extract Fintech keywords from these PDFs.

1. http://www3.weforum.org/docs/Beyond_Fintech__A_Pragmatic_Assessment_of_Disruptive_Potential_in_Financial_Services.pdf
2. http://www3.weforum.org/docs/WEF_The_future_of_financial_services.pdf
3. http://www3.weforum.org/docs/WEF_The_future_of_financial_infrastructure.pdf
4. http://www3.weforum.org/docs/WEF_A_Blueprint_for_Digital_Identity.pdf

We have to Scrape data from these links.

- <https://usbank.taleo.net/careersection/10000/jobsearch.ftl?lang=en&keyword=campus>
- <https://jobs.americanexpress.com/jobs?page=1>

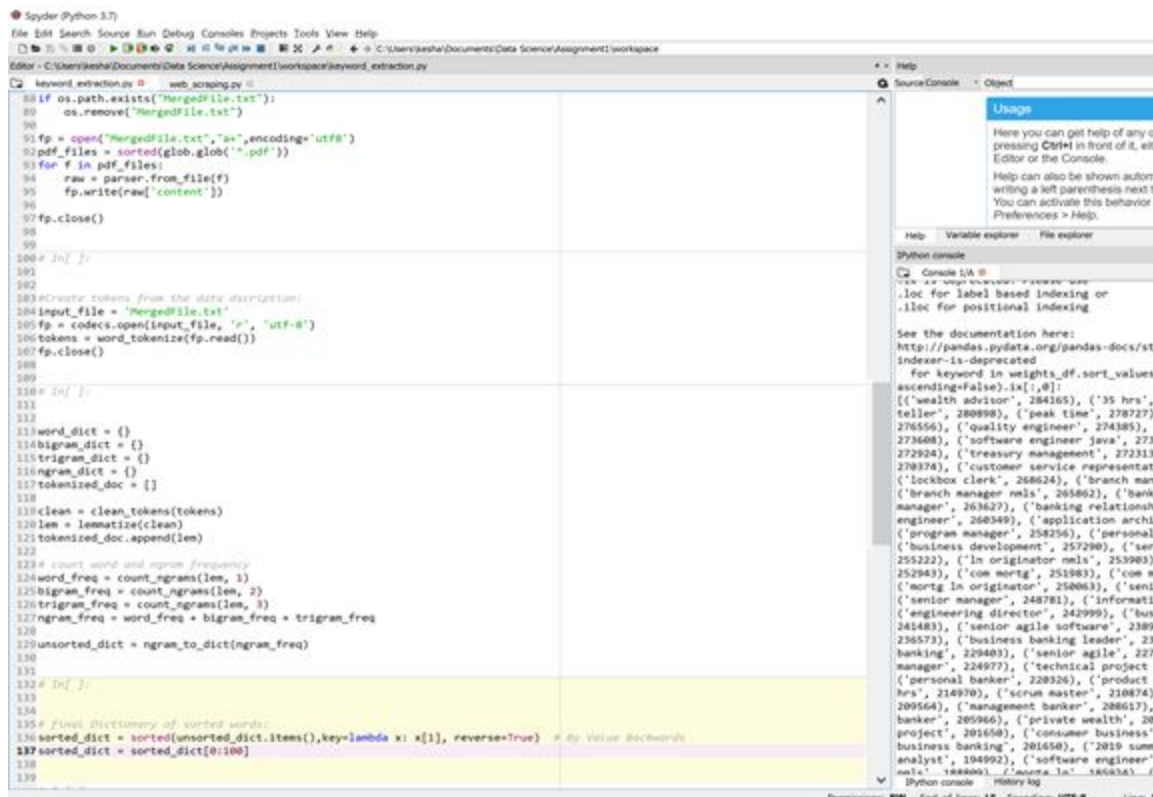
Data Preparation:

Data Merge:

- Parsed Fintech documents from World Economic Forum using tika library.
- Merged all documents into single text file.
- Read merged text and tokenize into words using nltk library

Data Cleaning:

- Clean Tokens: Ignoring case and punctuation. Remove stopwords provided by nltk library and custom stopwords to manually clear junk words or text.
- Lemmatize: Sort words by grouping inflected or variant forms of the same word.



```
178 if os.path.exists("mergedFile.txt"):
179     os.remove("mergedFile.txt")
180
181 #1 fp = open("mergedFile.txt", "a+", encoding="utf8")
182 pdf_files = sorted(glob.glob("*.pdf"))
183 for f in pdf_files:
184     raw = parser.from_file(f)
185     fp.write(raw['content'])
186 fp.close()
187
188 #2 # In[ ]:
189
190 #Create tokens from the data description:
191 input_file = "mergedFile.txt"
192 fp = codecs.open(input_file, 'r', 'utf-8')
193 tokens = word_tokenize(fp.read())
194 fp.close()
195
196 # In[ ]:
197
198 word_dict = {}
199 bigram_dict = {}
200 trigram_dict = {}
201 ngram_dict = {}
202 tokenized_doc = []
203
204 clean = clean_tokens(tokens)
205 lem = lemmatize(clean)
206 tokenized_doc.append(lem)
207
208 # Count word and ngram frequency
209 word_freq = count_ngrams(lem, 1)
210 bigram_freq = count_ngrams(lem, 2)
211 trigram_freq = count_ngrams(lem, 3)
212 ngram_freq = word_freq + bigram_freq + trigram_freq
213
214 unsorted_dict = ngram_to_dict(ngram_freq)
215
216 # In[ ]:
217
218 # Final Dictionary of sorted words:
219 sorted_dict = sorted(unsorted_dict.items(), key=lambda x: x[1], reverse=True) # By Value backwards
220 sorted_dict = sorted_dict[0:100]
221
222
```

Keyword extraction and ranking:

Word Frequency:

- Use clean tokens to create a list of words, bigrams, trigrams.
- Use nltk find most common 100 for each of the n-grams.
- Manually filter top 100 relevant using 300 n-grams.

```
140 # In[ ]:
141
142
143 #Creating CSV file:
144
145 if os.path.exists("Data/ds1.csv"):
146     os.remove("Data/ds1.csv")
147
148 csvData= [['Word', 'Score']]
149
150 for word, frequency in sorted_dict:
151     csvData.append([word,frequency])
152
153 with open('Data/ds1.csv', 'w', newline='',encoding='utf8') as csvFile:
154     writer = csv.writer(csvFile)
155     writer.writerows(csvData)
156
157 csvFile.close()
158
159
160 # ## TextRank
161
162 # In[ ]:
163
```

TF-IDF Score:

- Using stop_words and ngram_range execute TfidfVectorizer from sklearn and get tfidf score.
- Convert to dataframe and sort by score.

```
162 # In[ ]:
163
164
165 #Analyse the data to extract keywords
166 fp = codecs.open(input_file, 'r', 'utf-8')
167 tr4w = TextRank4Keyword()
168 tr4w.analyze(text=fp.read(),lower=True, window=3, pagerank_config={'alpha':0.85})
169
170 df=pd.DataFrame()
171 for item in tr4w.get_keywords(100, word_min_len=2):
172     df=df.append({'Word':item.word,'Score':item.weight},ignore_index=True)
173
174 #Extract the top 100 keywords from the analysed data based on weights:
175
176 df=df.nlargest(100,columns=['Score'])
177
178 if os.path.exists("Data/ds3.csv"):
179     os.remove("Data/ds3.csv")
180
181 columnsTitles = ['Word', 'Score']
182 df = df.reindex(columns=columnsTitles)
183
184 #Create the CSV:
185 df.to_csv("Data/ds3.csv",index=False)
186
187
188 # ## TF-IDF
189
190 # In[ ]:
```

TextRank:

- In order to find relevant keywords, the textrank algorithm constructs a word network. This network is constructed by looking which words follow one another.
- A link is set up between two words if they follow one another, the link gets a higher weight if these 2 words occur more frequently next to each other in the text.

Web Scraping:

Selenium webdriver is used to get the dynamic data from Java-script in the web pages

```
: driver = webdriver.Chrome("C:/chromedriver.exe")
amex_joblistings = []
url = 'https://jobs.americanexpress.com/jobs?page=1'
driver.get(url)
time.sleep(2)
body = driver.find_element_by_tag_name("body").get_attribute('innerHTML')
soup = BeautifulSoup(body, "html.parser")
#Getting total number of pages
pgno = soup.find(class_="mat-paginator-range-label").get_text()
last_page = math.ceil(int(pgno[10:])/10)+1
```

Scraping all URL's of pages where jobs are posted using beautiful soup

```
: for i in range(1,last_page):
    url = 'https://jobs.americanexpress.com/jobs?page='+str(i)
    driver.get(url)
    time.sleep(2)
    body = driver.find_element_by_tag_name("body").get_attribute('innerHTML')
    soup = BeautifulSoup(body, "html.parser")
    links=soup.find_all(class_="job-title-link")
```

- Web Scraping done for US Bank and American Express
- Data extracted from webpage

Top Job Titles in Fintech Industry:

```

Editor - C:\Users\Aayush\Documents\NEU Notes\Data Science\Assignment 1\Graph\Graph-Top fintech oriented jobs.py
Scraped.py dataanalysis.py Graph-Top fintech oriented jobs.py Top Categories in Fintech jobs.py keyword_extraction_manual_list.py

22 def lemmatize(tokens):
23     """ Removes plurals """
24     return [lemmatizer.lemmatize(token) for token in tokens]
25
26 #function to create ngram, bigram, trigram
27 def count_ngrams(tokens, n):
28     n_grams = ngrams(tokens, n)
29     ngram_freq = collections.Counter(n_grams)
30     ngram_freq = ngram_freq.most_common()
31     return ngram_freq
32
33 #function to create dictionary of words and frequencies:
34 def ngram_to_dict(ngram_freq):
35     l = []
36     for t in ngram_freq:
37         l.append(' '.join(t[0]),t[1]))
38     return dict(l)
39
40
41 jobTitles = []
42 finalDataCSV = pd.read_csv("C:\\Users\\Aayush\\Documents\\NEU Notes\\Data Science\\Assignment 1\\Graph\\FinalData.csv",encoding = "ISO-8859-1")
43 jobTitles = finalDataCSV['Job Title'][:]
44 tokens = nltk.word_tokenize(str(jobTitles))
45 clean = clean_tokens(tokens)
46 lem = lemmatize(clean)
47
48 bigram_dict = {}
49 trigram_dict = {}
50 quadgram_dict = {}
51
52
53 bigram_freq = count_ngrams(lem, 2)
54 trigram_freq = count_ngrams(lem, 3)
55 quad_freq = count_ngrams(lem, 4)
56 ngram_freq = bigram_freq + trigram_freq + quad_freq
57
58 unsorted_dict = ngram_to_dict(ngram_freq)
59 sorted_dict = sorted(unsorted_dict.items(),key=lambda x: x[1], reverse=True)
60
61 tfidf = TfidfVectorizer(analyzer='word', stop_words='english', ngram_range=(2,3))
62 response = tfidf.fit_transform(jobTitles)
63 weights = np.asarray(response.mean(axis=0)).ravel().tolist()
64 weights_df = pd.DataFrame({'word': tfidf.get_feature_names(), 'Score': weights})
65 weights_df = weights_df.sort_values(by='Score', ascending=False)
66
67 x = 0
68 mydict = {}
69 for keyword in weights_df.sort_values(by='Score', ascending=False).ix[:,0]:
70     for index,row in finalDataCSV.iterrows():
71         x += finalDataCSV.loc[index, 'l': '100'].sum() if (keyword in str(finalDataCSV['Job Title'][index]).lower()) else 0
72
73     mydict[keyword] = x
74
75 import os
76 import csv
77 sorted_dict = sorted(mydict.items(),key=lambda x: x[1], reverse=True)
78

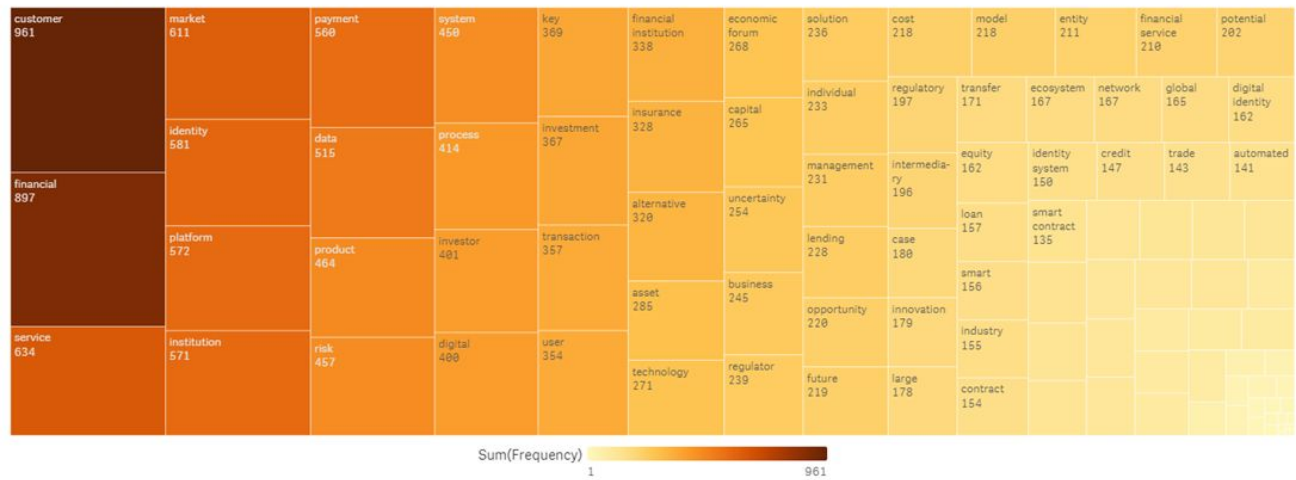
```

- Compiled a list of every bigrams, trigrams in job titles in the final data
- Combined similar job titles and compared them with each job listing and deduced the final fintech score for each job title.
- Hence extracted top fintech oriented job titles.

Insights:

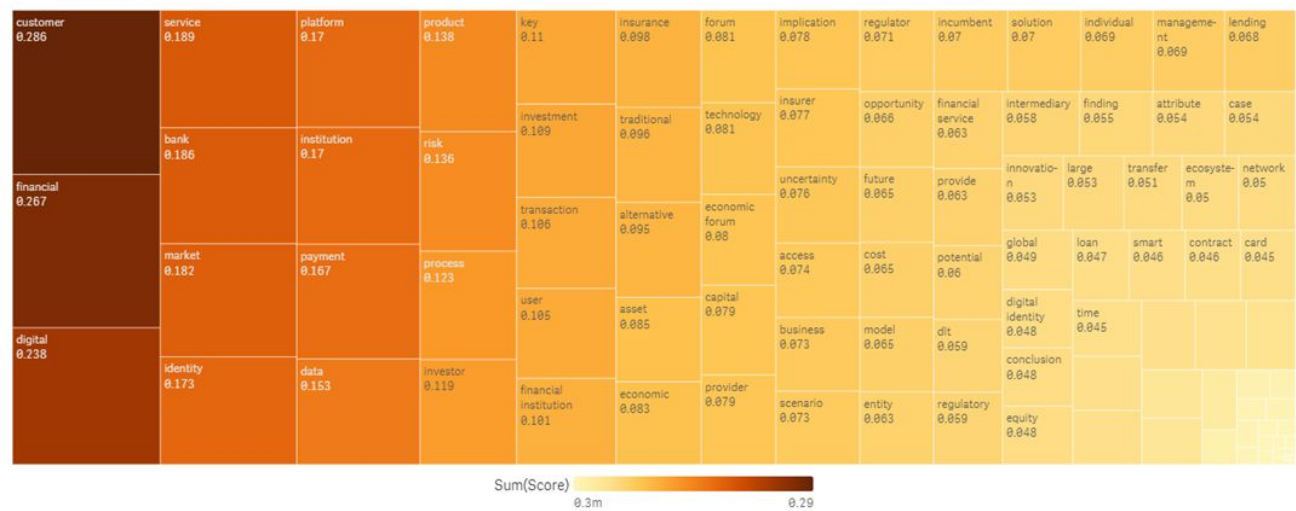
Part 1 A Wordcount Analysis for 100 words

Part 1 Wordcount Analysis



Part 1B TF/IDF score for 100 words

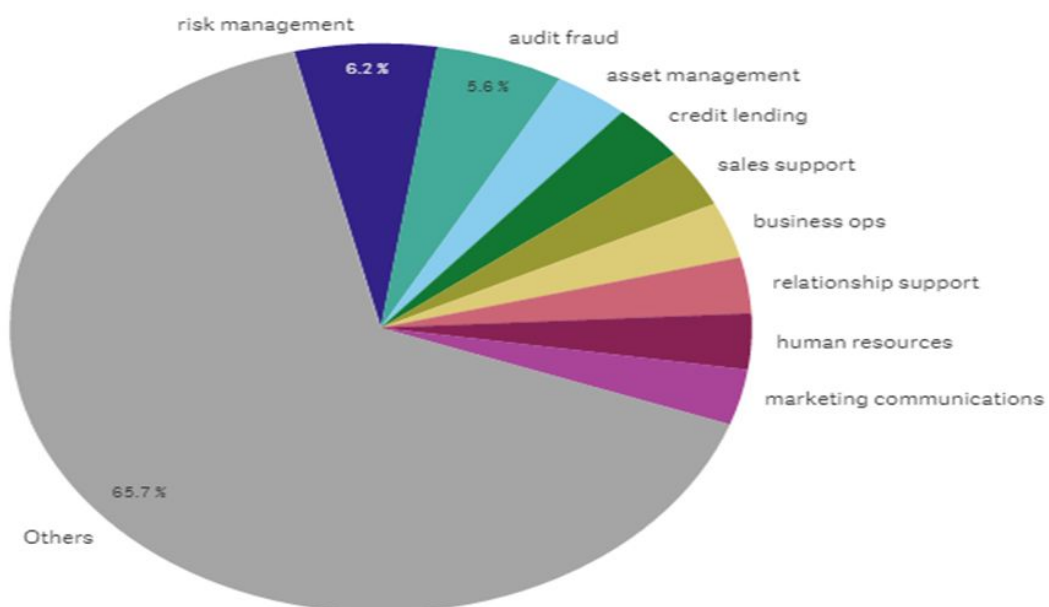
Part 1 TF/IDF score for 100 words

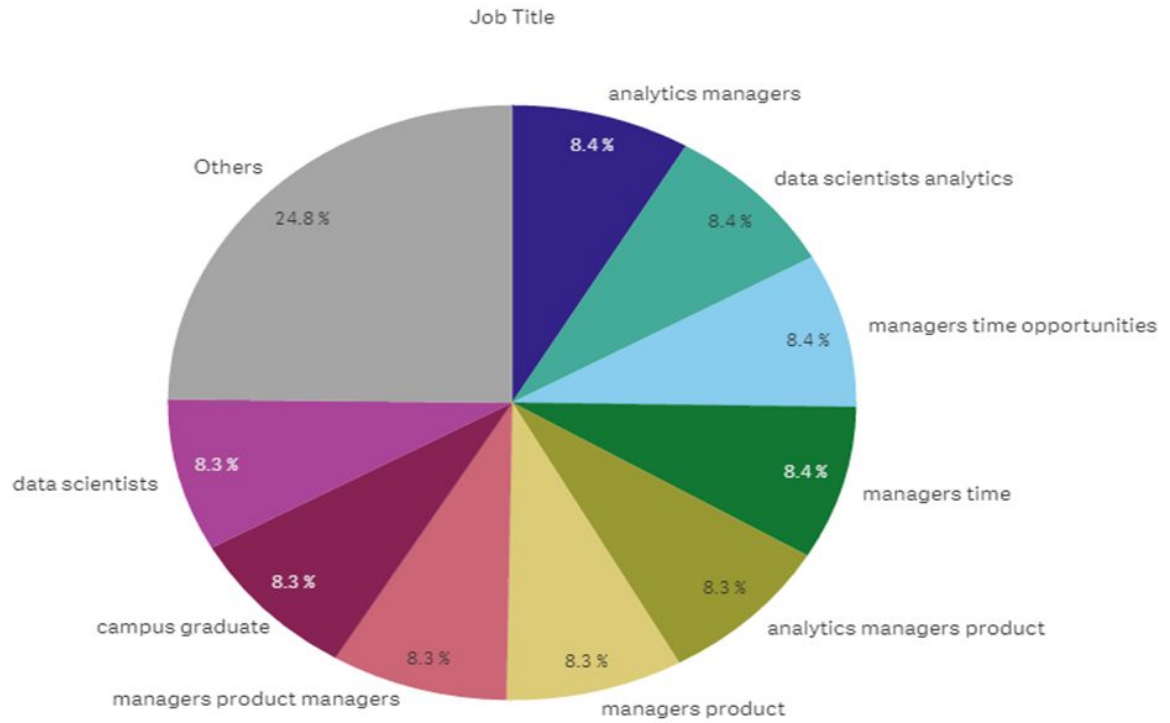


Text Rank analysis



Job Cat





Total Occurrences of each word in Fintech in Job Market

Total Occurrences of each word in Fintech in Job Market

