

## Project Ideas: Capstone Project

<b>Objective: Capstone</b>	<b>Understand the python codes &amp; packages to use and understand the whole A-Z of performing data analyses, including importing, cleaning, wrangling and visualization of the data.</b>
<b>Project Idea (I)</b>	<p><b>Text mining in Python:</b> More than 70% of potentially usable data exists in unstructured form. This data is stored in text format. Text mining or Natural language processing provides us with techniques to gain actionable insights from these data.</p> <p>In this project, I propose to address HR analytics problem of understanding what the employees are saying about their company and how HR head can develop their strategy based on the insights we provide them with text mining of the reviews.</p> <p><b>Data Set:</b> Amazon VS Google Reviews. (500+ and 500-ve reviews). Source (Data Camp)</p> <p><b>Questions:</b></p> <ul style="list-style-type: none"> <li>• Which company has a better work-life balance?</li> <li>• Which is better perceived to pay according to the online reviews?</li> </ul>
<b>Project Idea (II)</b>	<p><b>HR Analytics/ People Analytics / Workforce Analytics:</b> For companies it is becoming more important every day, to hire the best talent and maintain a low turnover rate. At the same time, they want their existing employees to keep their productivity and engagement levels high and accidents a lowest. In this project, I propose to analyze several data sets to address these issues through statistical</p> <p><b>Data Set:</b> a) HR data (2940, 4) b) Accident data (302,3), c) Survey data, d) Performance Data.</p> <p><b>Questions:</b></p> <ul style="list-style-type: none"> <li>• When and where is the highest accident rate?</li> <li>• Explaining the reasons for increase in accidents?</li> <li>• Regression to analyze drivers.</li> <li>• What is driving low employee engagement?</li> </ul>
<b>Project Idea (III)</b>	<p><b>Market Analytics / Statistical Modeling:</b> Companies have to make important decisions every day. Statistical models help companies make hypothesis-driven business decisions based on the market data. In this project for the capstone project, I propose to explore the sales data and help them in deciding which customers are important for their business and growth.</p> <p><b>Data Set:</b> Churn Data, Sales Data, Sales Data-months 2-4, survival data, Default data, News data.</p> <p><b>Questions:</b></p> <ul style="list-style-type: none"> <li>• Which customers are valuable to your business?</li> <li>• Which customers will leave your business?</li> <li>• Survival Analysis.</li> <li>• CRM data Analysis.</li> </ul>

## Text Mining for Data Science with Python

<b>Natural Language processing or Texting Mining</b>	<b>Introduction:</b> More than 70% of potentially usable data exists in unstructured form. This data is stored in text format. Text mining or Natural language processing provides us with techniques to gain actionable insights from these data.
<b>Problem Statement</b>	<b>Human Resources Analytics:</b> The HR department of Amazon wants to develop strategies for hiring appropriate talent and prospective employees who are more suited to their work culture. They also want to understand what distinguishes Amazon employees with Google employees, which is another technology giant. The Data Scientists from Springboard have been asked to explore online reviews and provide insights.
<b>Datasets</b>	<b>Amazon and Google Reviews.</b> (500+ and 500-ve reviews) Several employees from both companies submit online reviews. These are categorized into positive and negative reviews. Four data sets are available for analysis. Resource: Datacamp (Ted Kwartler)
<b>Questions to be addressed</b>	<ul style="list-style-type: none"><li>• What are employees saying about Amazon and Google?</li><li>• Which company has better work-life balance?</li><li>• Which company is perceived better in compensating according to the online reviews?</li><li>• Can HR use the analyses in strategizing their talent hunt?</li></ul>
<b>Methods</b>	Steps involved in Analyzing the Text Data (NLTK library) 1) Organization & Text processing (raw text -> sentence segmentation -> tokenization -> corpus -> matrix -> word count) 2) Feature extraction and Analysis (TF, DTM, TF-ID, TF-IDF, DM -> Top terms & Word associations) 3) Machine Learning (clustering, classification, association rules, predictive modeling) 4) Data visualization: Dendrograms, Word cloud. Pyramid plots.
<b>Significance</b>	NLP can provide key business intelligence insights, which are not just limited to human resource management but can also help other departments including marketing to improve sales, accounts and finances, purchasing, customer service support, distribution, research, and development.

# Text Mining, Tech Giant Reviews using Natural Language Processing

## Introduction

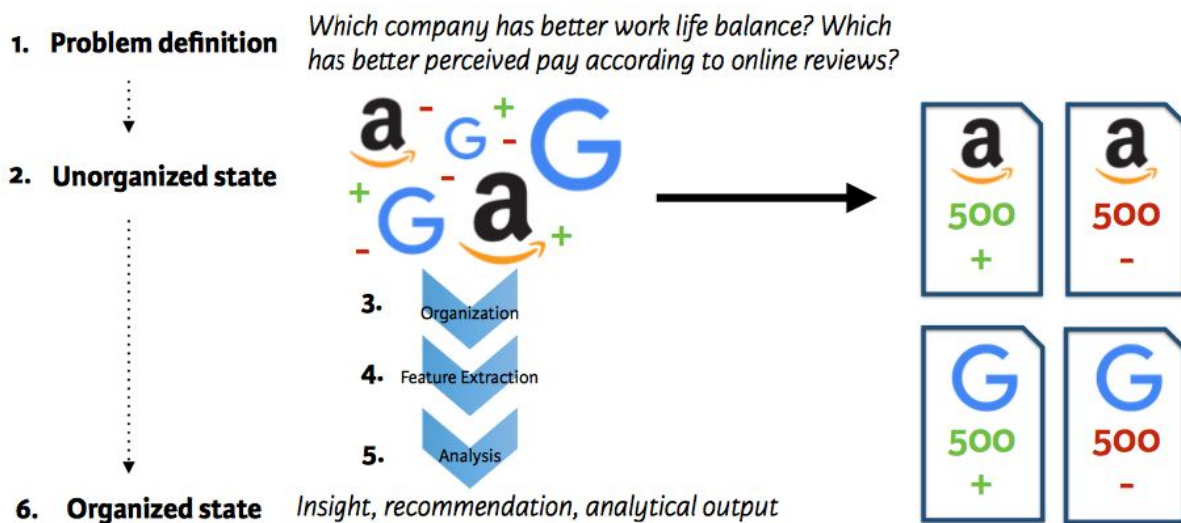
More than 70% of potentially usable data exists in unstructured form. This data is stored in text format. Text mining or Natural language processing provides us with techniques to gain actionable insights from these data.

## Problem Statement

The HR department of Amazon wants to develop strategies for hiring appropriate talent and prospective employees who are more suited to their work culture. They also want to understand what distinguishes Amazon employees with Google employees, which is another technology giant. The Data Scientists from Springboard have been asked to explore online reviews and provide insights.

## A case study in HR analytics

*Adapted from*



*Datacamp course*

## **Questions to be addressed through text mining**

- What are employees saying about Amazon and Google?
- Which company has better work-life balance?
- Which company is perceived better in compensating according to the online reviews?
- Can HR use the analyses in strategizing their talent hunt? Recommendations to the HR team?

## **Methods / Steps used to answer the above questions**

- Organization & Text pre-processing (raw text -> sentence segmentation -> tokenization -> corpus -> matrix -> word count)
- Feature extraction and Analysis (TF, DTM, TF-ID, TF-IDF, DM -> Top terms & Word associations)
- Machine Learning (clustering, classification, association rules, predictive modeling) to predict reviews.
- Data visualization: Dendrograms for word relationships, Word clouds. Pyramid plots for comparison of words used in reviews.

## **Observations through Analyses:**

### **Qs 1: What are employees saying about Amazon and Google?**

Analyses of top 15 words (bigrams) from the Amazon Reviews.

Amazon Pro		Amazon con	
good pay	25	long hours	29
great benefits	24	worklife balance	21
smart people	20	work life	21
place work	17	life balance	20
good benefits	16	not enough	9
fast paced	16	peak season	8
great place	12	management not	8
learn lot	12	not good	8
work environment	11	high turnover	7
great pay	11	work environment	6
good work	10	hard work	6
pay good	10	hours long	6
pay great	10	many people	6
people work	10	work hours	6
pay benefits	9	not really	6

Analyses of top 15 words (bigrams) from the Google Reviews.

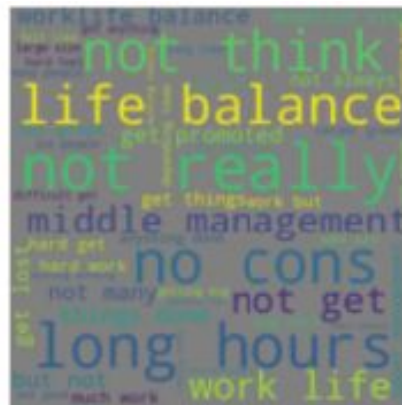
Google Pro		Google Con	
smart people	42	not really	11
free food	41	no cons	10
place work	26	long hours	10
great benefits	22	not think	9
great perks	20	not get	8
great work	18	middle management	8
people great	16	work life	8
work environment	16	life balance	8
great people	16	worklife balance	8
great place	15	get promoted	7
great culture	14	not many	7
worklife balance	12	but not	7
people work	12	things done	7
work life	12	get things	6
work great	11	get lost	6

We can also visualize these as word clouds.

### Amazon Pros and Cons



### Google Pros and Cons

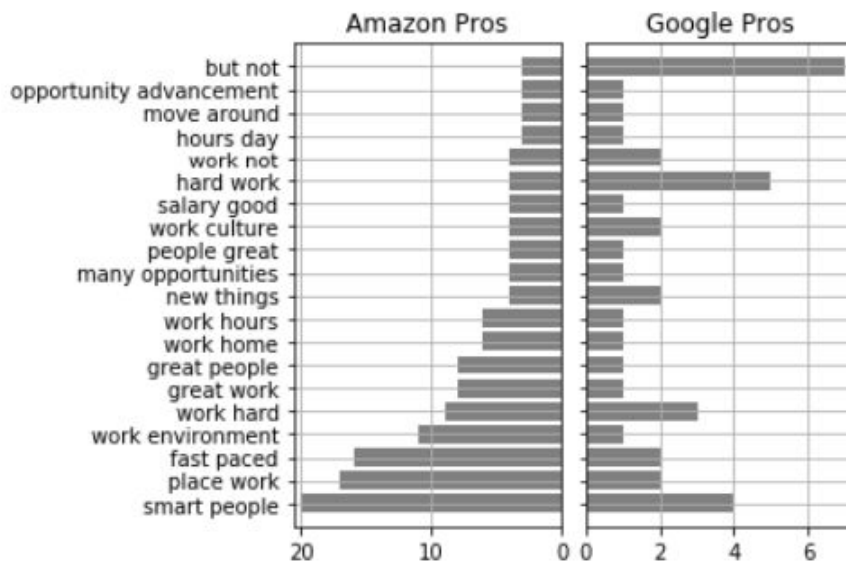


**Q2: Which company has better work-life balance?**

From the frequency table and the word cloud we see that work-life balance bigram appears in Amazon con reviews appears 21 times. Where as in Google reviews the word appears in both pros and cons 12 and 18 times respectively. This would suggest that Google has better work life balance ratings.

### Q3: Which company is perceived better in compensating according to the online reviews?

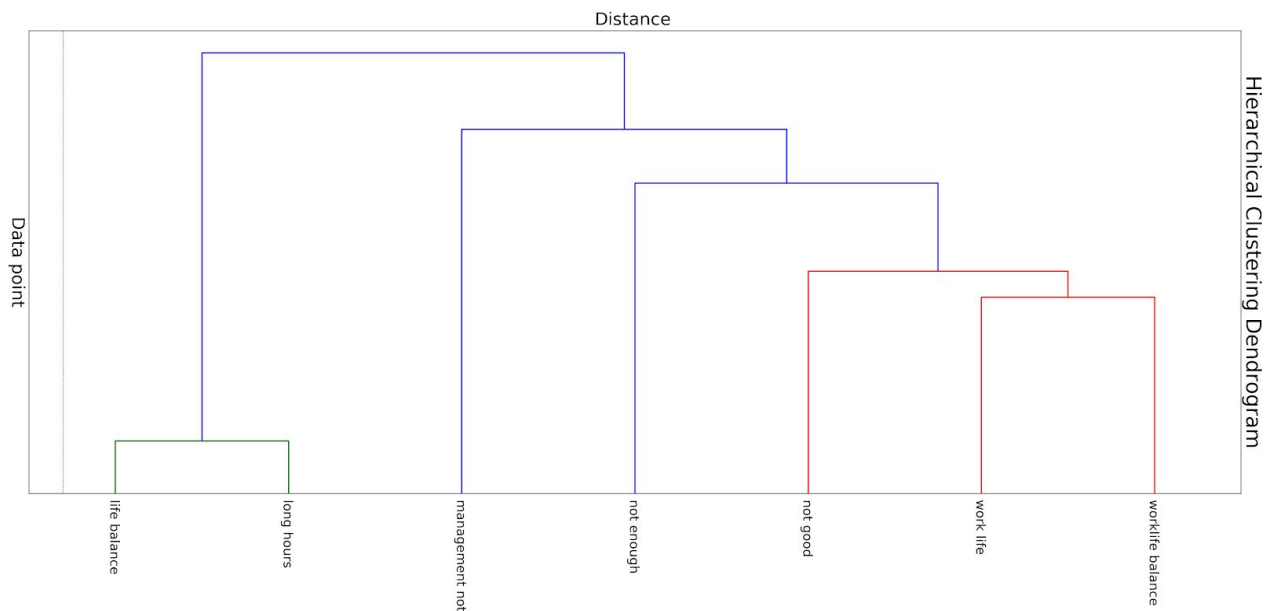
We can see that more employees in amazon are talking positive about the “good pay” and “benefits” as compared to google. This can be visualized by plotting pyramid plots.



#### Q4: Can HR use the analyses in strategizing their talent hunt?

##### Recommendations to the HR team?

For this we can do more detailed analysis including generating TF-IDF and word associations to observe which words are being used by same employees more consistently. It seems there is a strong indication of long working hours and poor work-life balance in the reviews.



From Hierarchical Clustering we also observe that employees like fast paced work at amazon. This provides an indication to HR to include this criteria in their hiring process, and to find prospective employees which would like fast paced environment. Please turn over for the dendrogram.

### **Machine Learning to predict if the a review is a pro or a con:**

We created a master dataframe containing all the pros and cons in a single column. We labeled the data from the four data sets we had with us. We split this data set into train and test data tests. We trained the model in using the train set and predicted using the test test.

We used logistic regression, random forest and multinomial naive bayes to train, test and predict whether the review is a pro or a con.

We also ran grid search to identify the best hyperparameters for the model.

### **Results:**

Multinomial naive bayes model worked the best for making the text predictions with better accuracy scores than the other two algorithms.

Following are the top words showing the highest and least probability for pros reviews.

### **Predicting top features / words of pros and cons.**



Pros	P(pros   word)
pro googl	0.94
multipl opportun	0.94
take easi	0.94
everyth veri	0.93
but long	0.92
amazon look	0.92
cultur transpar	0.92
cowork upward	0.92
well equal	0.92
good give	0.92
Cons	P(pros   word)
night work	0.16
day hr	0.15
big someth	0.15
balanc softwar	0.15
area live	0.15
process driven	0.15
due larg	0.14
burnout real	0.14
hour manag	0.12
great worker	0.09

## Conclusions:

We were able to process unstructured data and provide the following key insights from 2000 reviews on Glassdoor.

- **Work-life balance rating:** Based on the text analysis we observe that Google has better work life balance ratings than Amazon. Mid-management level needs an evaluation.
- **Compensation:** Between the two companies we find that Amazon provides better compensation than Google.
- **Recommendation for HR:**
  - We observe that Amazon employees prefer fast paced environment. This could be added to the job posting as an essential part of their work culture and set priority for applicants who like the same.
  - Amazon employees like good pay. This can be used by HR to attract more talent.
  - Google employees prefer work life balance. This can be used to attract more talented employees.

This NLP pipeline can be used to gain business Intelligence in areas of sales, marketing & commercialization.