



# Bagging Significant Reviews

# Presentation Highlights

## General Discussion Flow

- 
- About Our Project
  - Detailed Design Description
  - Scope Of Work
  - Findings
  - Things We Learnt

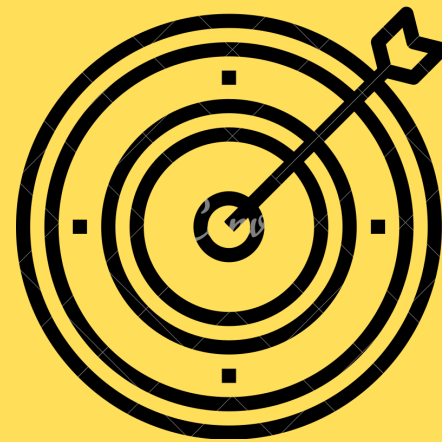
NEXT →

# OUR IDEA IN A MINUTE



NEXT →

# PROBLEM STATEMENT AND OBJECTIVE

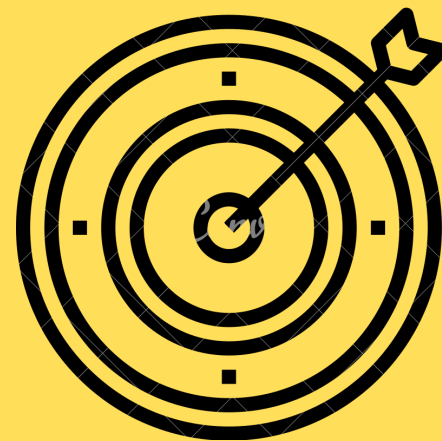


Looking at thousand of reviews can be a time-consuming task and choosing the right reviews for the prefect buy isn't that easy. So, we aim to help customers **choose the most significant reviews** before making an online purchase by extracting information from consumer data.

What we are trying to solve?

NEXT →

# PROBLEM STATEMENT AND OBJECTIVE



What we are trying to solve?

Looking at thousand of reviews can be a time-consuming task and choosing the right reviews for the prefect buy isn't that easy. So, we aim to help customers **choose the most significant reviews** before making an online purchase by extracting information from consumer data.

Human beings are emotional creatures and their needs are constantly evolving with time and **knowing their sentiments** about a product undoubtedly gives them an edge over their competitors. So, with the help of our project we'd like to help both buyers and sellers in making the most out of the reviews.

NEXT →

# ABOUT DATA EXTRACTION

Which product

Flipkart Perfect Homes Opus  
Engineered Wood Queen  
Box Bed from flipkart.com



NEXT →

# ABOUT DATA EXTRACTION

Which product

Flipkart Perfect Homes Opus  
Engineered Wood Queen  
Box Bed from flipkart.com

What we scraped

Reviews, Ratings, Likes,  
Dislikes, Date, Consumer Name



NEXT →

# ABOUT DATA EXTRACTION

Which product	Flipkart Perfect Homes Opus Engineered Wood Queen Box Bed from flipkart.com
What we scraped	Reviews, Ratings, Likes, Dislikes, Date, Consumer Name
Tools and Libraries Used	Python, Microsoft Excel Selenium, Request, Chrome Web Driver, Datetime etc.
No. of reviews scraped	2506

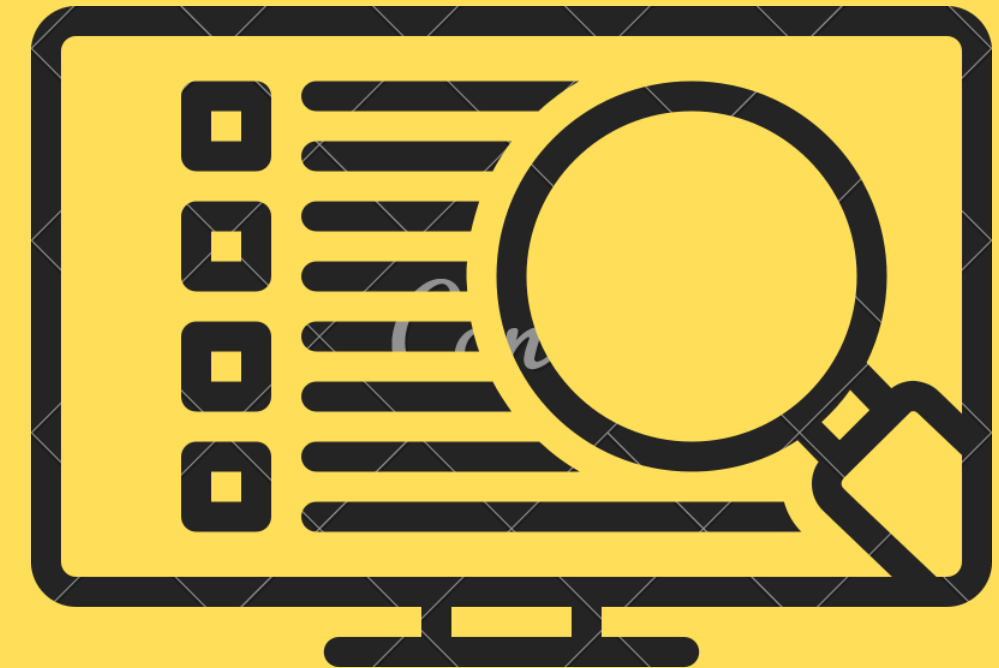


NEXT →



# Detailed Design Description

- Text Cleaning
- Information Extraction
- Noun Extraction and Pattern Mining
- Feature Selection
- Similar Words
- Polarity and Binary Data Preparation
- Calculation of Weights and Clustering



# TEXT CLEANING

## GRAMMAR CHECK

Using language tool python



NEXT →

# TEXT CLEANING

## GRAMMAR CHECK

Using language tool python

## RECTIFY SLANG

gud, nyc, gr8,



# TEXT CLEANING

## GRAMMAR CHECK

Using language tool python

## RECTIFY SLANG

gud, nyc, gr8

## SPELLING CORRECTION

Prodct --> Product

## PARSING HTML TAGS

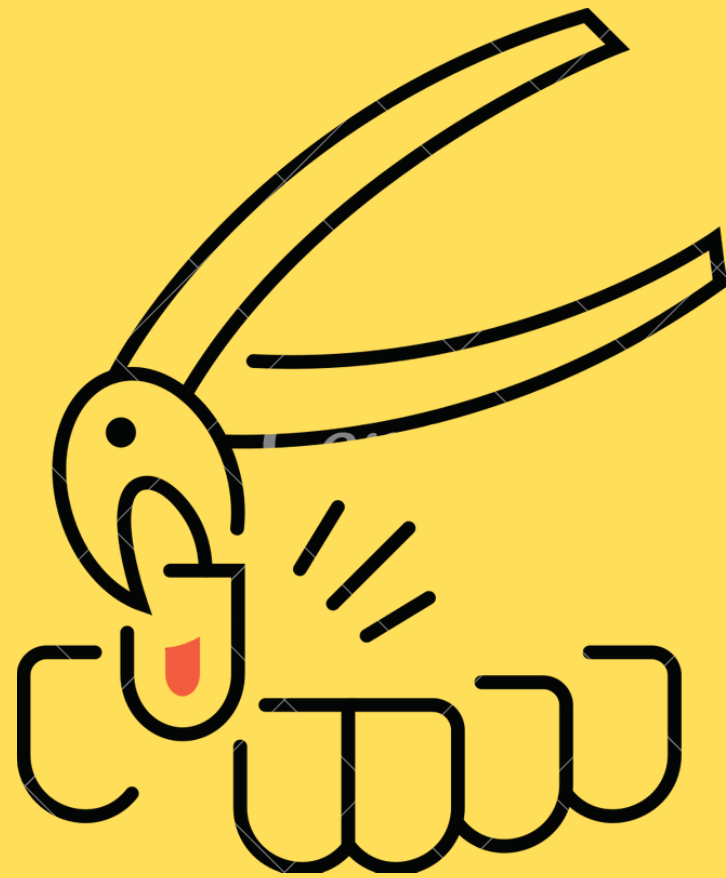
Using html.parser

## CONTRACTOR

I'm --> Im



# NOUN EXTRACTION AND PATTERN MINING



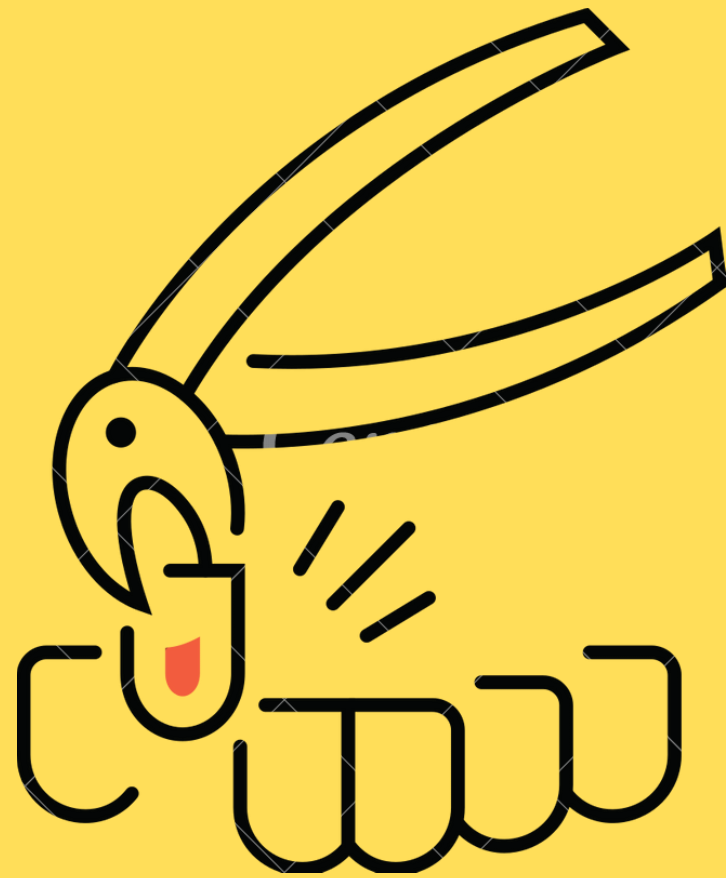
WHY NOUNS?

POS TAGGING

Bag of Nouns

NEXT →

# NOUN EXTRACTION AND PATTERN MINING



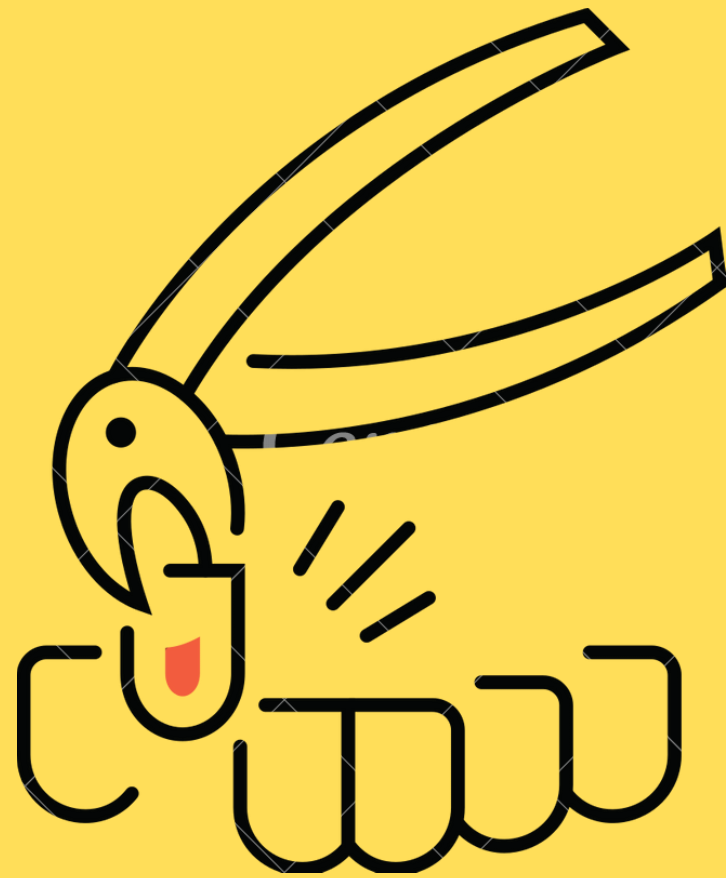
## POS TAGGING

Bag of Nouns

APRIORI ALGORITHM VS  
FP GROWTH ALGORITHM  
(EXPERIMENT)

NEXT →

# NOUN EXTRACTION AND PATTERN MINING



## POS TAGGING

Bag of Nouns

APRIORI ALGORITHM VS  
FP GROWTH ALGORITHM  
(EXPERIMENT)



FP GROWTH

NEXT →

# INFORMATION EXTRACTION

## USING NLP, SPACY AND POS TAGGING

### PREPOSITION LINKAGE

product within range  
installation of bed

### ADVERB ADJECTIVE LINKAGE

very good product

### ADJECTIVE-NOUN LINKAGE

good installation

### ASPECT KEYWORDS

good material  
extra screws

### NOUN-VERB LINKAGE

improve quality  
sharp edges





## SNAPSHOT OF INFORMATION EXTRACTION

	A	B	C	D	E
1	Noun_Verb_Noun_checked	adverb_adjective_noun	adjective_noun	adverb_adjective_noun	aspect_data
2	['well installation']	['very good person']	[]	['very good person']	['feeling well']
3	['sleep crack sound', 'we dismantle bed', 'you']	[]	['proper installation', 'bad thing', 'good']	[]	['bad person', 'even kneel', 'good storage facility', 'very bad person', 'kneel back', 'easily fit', 'good amount']
4	['product put inches', 'i have insecurity', 'it is']	['too sharp sometime']	['best product', 'good materials', 'little is']	['too sharp sometime']	['sometime hurts', 'great one', 'good ok product', 'such materials', 'best ok product', 'sleep now']
5	[]	[]	['perfect storage', 'medium size', 'white']	[]	['easily came', 'pressure immediately', 'actually installed', 'installed properly', 'even complain']
6	[]	[]	['build quality', 'extra screws', 'tough tie']	[]	['fits perfectly']
7	[]	[]	['exact time', 'orthopaedic mattress', 'to']	[]	['keep away', 'nicely suited', 'maintain properly', 'long sale time', 'exact sale time']
8	['improve quality']	['always good installation', 'also good br']	['friendly product', 'more customers', 'p']	['always good installation', 'also good']	['surely more customers', 'new house']
9	['i like bed']	['so much weight']	['moist area', 'only thing', 'light weight']	['so much weight']	['suffer separately', 'small inch plastic', 'light weight mattress']
10	[]	[]	[]	[]	[]
11	[]	[]	['little bit', 'right colour', 'same day']	[]	['little bit']
12	['durable within given price range']	[]	['few comments']	[]	['difficult to shift', 'well designed', 'easily shift']
13	['you buy it because of finishing']	[]	[]	[]	['thank much']
14	[]	[]	['first time']	[]	['thank much']
15	['product look value']	[]	['polite product']	[]	['nice n value']
16	['service guy installs bed', 'service guy take h']	['very good flipkart', 'very first time', 'all']	['premium finishing', 'yearsbroverall pr']	['very good flipkart', 'very first time', '']	['totally depends', 'depends cautiously', 'very first time']
17	[]	[]	['total bed', 'own everytime', 'good com']	[]	['other bed', 'direct sun light', 'whenever want']
18	[]	[]	['timely delivery', 'mess thanks']	[]	['very nice quality', 'nice quality']
19	[]	['very big storage']	['4th day', 'nice product', 'huge box', 'as']	['very big storage']	['very big box storage', 'big box storage', 'sounds all']
20	[]	[]	['next day']	[]	[]
21	[]	[]	['extra money', '3rd floor', 'next day', 'n']	[]	['extra money', 'widely used', 'good job flipkart', 'nodular furniture', 'only say']
22	[]	[]	['good products', 'good quality', 'super i']	[]	['very trained', 'good quality products', 'well trained']
23	['we use bedsheets', 'we shove ends', 'daugh']	['very sturdy amp', 'very sharp edges']	['joint amp', 'different way', 'next day']	['very sturdy amp', 'very sharp edges']	['different way', 'very strong joint', 'once installed', 'strong joint']
24	[]	[]	['good deal']	[]	['well behaved', 'br/>the experience']
25	[]	['also nice product']	['good product', 'huge storage', 'short re']	['also nice product']	['huge storage space', 'good va product', 'little firing job', 'shift then']
26	[]	[]	['worth rs', 'good product']	[]	['sleep comfortably', 'overall product']
27	[]	[]	['old reviews', 'much colour', 'same day']	[]	['old reviews', 'much colour selections', 'other service']
28	[]	['only such type']	['long term', 'ok product']	['only such type']	['long term use']
29	[]	[]	['sharp edges', 'plain surface', 'bottom b']	[]	['especially finishing', 'good design', 'borrowed easily']
30	[]	[]	['polite installation', 'same day', '2nd fl']	[]	['very good installation delivery person', 'good installation delivery person', 'quick installation']
31	['i buy part']	[]	[]	[]	[]
32	['corners hurt legs']	[]	['under 10k', 'wood fit']	[]	['once assembled', 'under price tag', 'cleverly designed', 'well designed', 'take apart']

# FEATURE SELECTION

## TERM FREQUENCY-INVERSE DOCUMENT FREQUENCY

TF-IDF is the multiplication of the TF and IDF

**10 Features** : money, service, installation, delivery, wood, storage, design, quality, bed, product



# FEATURE SELECTION

## TERM FREQUENCY-INVERSE DOCUMENT FREQUENCY

TF-IDF is the multiplication of the TF and IDF

**10 Features** : money, service, installation, delivery, wood, storage, design, quality, bed, product

## PHRASES FEATURES USING FP GROWTH ALGORITHM

price range, bed size, product installation, product design etc.



# FEATURE SELECTION

## TERM FREQUENCY-INVERSE DOCUMENT FREQUENCY

TF-IDF is the multiplication of the TF and IDF

**10 Features** : money, service, installation, delivery, wood, storage, design, quality, bed, product

## PHRASES FEATURES USING FP GROWTH ALGORITHM

price range, bed size, product installation, product design etc.

## CLUBBED PHRASES VS PHRASE FEATURES (EXPERIMENT)

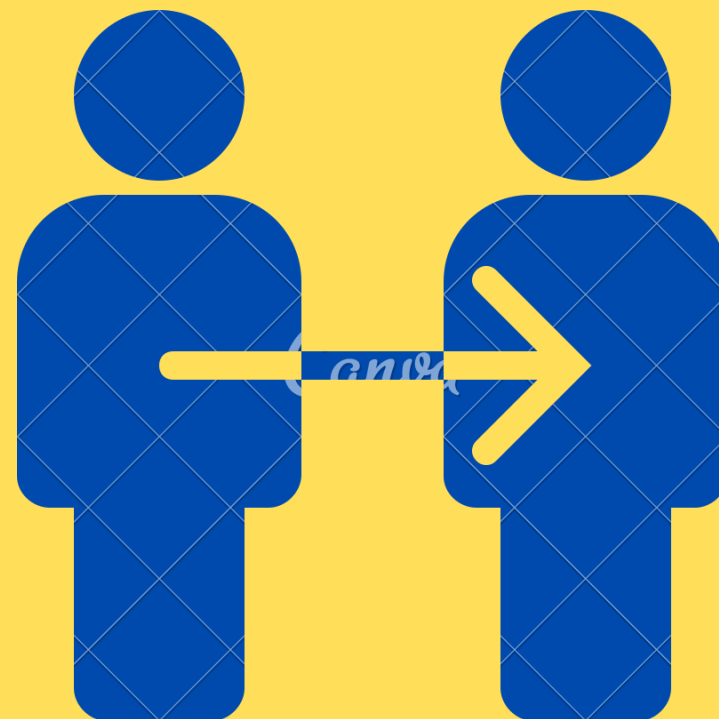


EXPERIMENT RESULT

A	B	C	D	E	F	G	H	I	J	K
features										
money product service  installation bed quality wood design storage delivery	value	product value	bed value	price range	product range	price bed	range	worth the money	worth the price	value for money
	product purchase	product bed	product delivery	product price	product wood	product design	product size	product		
	service installation	service time	service product	service quality						
				installation			installation			
	installation bed	installation quality	installation person	delivery	installation team	installation time	product			
	bed wood	bed height	bed delivery	bed size	bed quality	bed fit				
	quality delivery	quality wood								
	plywood									
	design bed									
	storage bed	storage product								
	time delivery									

42 PHRASED FEATURES CLUBBED INTO 10 MAIN FEATURES

## SIMILAR WORDS



BED --> COT  
BED --> FURNITURE

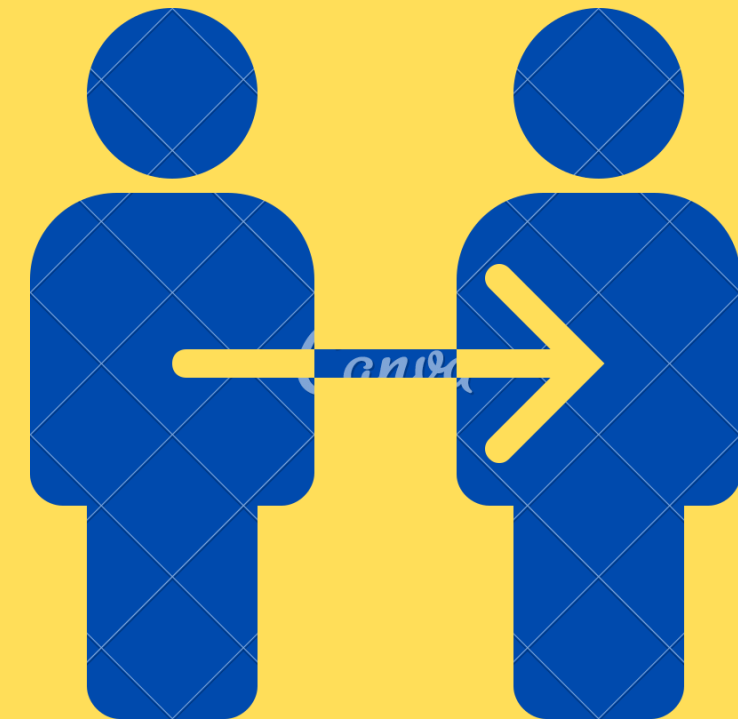
**WORD 2 VEC**  
**LEVENSHTEIN DISTANCE**

**NEXT →**

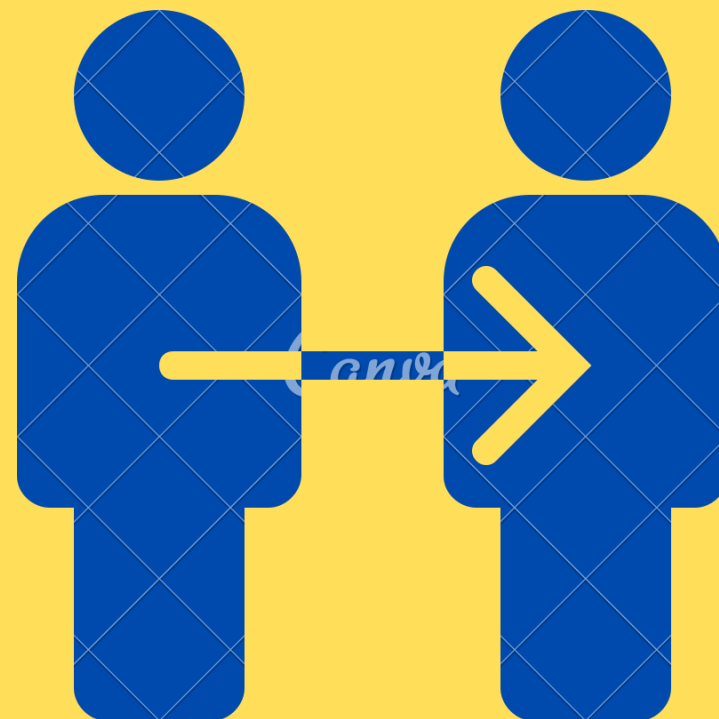
## LEVENSHTEIN DISTANCE

The Levenshtein distance is a number that tells you **how different two strings are**. The higher the number, the more different the two strings are.

For example, the Levenshtein distance between “kitten” and “sitting” is 3 since, at a minimum, 3 edits are required to change one into the other.



## SIMILAR WORDS



BED --> COT  
BED --> FURNITURE

**WORD 2 VEC**  
**LEVENSHTEIN DISTANCE**

**COSINE SIMILARITY**

**WEB SCRAPPING**

from synonyms.com

**NEXT** →



# POLARITY AND BINARY DATA PREPARATION

FUZZYWUZZY LIBRARY  
AND  
COSINE SIMILARITY

REVIEW ID										
	money	product	service	nstallation	bed	quality	wood	design	storage	delivery
0	0	0	0	1	0	0	1	0	0	0
1	0	0	0	0	1	1	1	0	1	0
2	1	1	1	0	0	1	1	0	0	0
3	1	0	0	0	1	0	1	0	1	0
4	1	1	1	0	1	0	0	0	0	0
5	0	0	0	0	1	0	0	0	0	0
6	1	1	1	1	0	1	1	0	0	0
7	0	0	0	1	1	0	0	0	0	0

## SENTIMENT EVALUATION TECHNIQUE

VADER SENTIMENT

	money	product	service	nstallation	bed	quality	wood	design	storage	delivery
0	0	0	0	0.2732	0	0	0.4927	0	0	0
1	0	0	0	0	-0.5849	0.5994	0.4404	0	0.4404	0
2	0.6249	0.7506	-0.3626	0	0	-0.4215	0.4404	0	0	0
3	0.34	0	0	0	-0.296	0	0.4404	0	0.5719	0
4	-0.128	0.6369	0.6369	0	0.0258	0	0	0	0	0
5	0	0	0	0	0.4404	0	0	0	0	0
6	0.4404	0.4939	0.4404	0.4404	0	0.4404	0.4404	0	0	0
7	0	0	0	0.6249	0.3612	0	0	0	0	0

NEXT →

# CALCULATION OF WEIGHTS AND CLUSTERING

feature	total	clusters	freq	weights
product	178.3828	2	541	3.032804
wood	97.4932	0	234	2.400167
bed	55.2017	3	267	2.520235
money	50.7408	3	231	2.180428
installatio	26.6609	1	115	1.212065
service	21.6191	1	139	1.465018
delivery	18.2396	1	73	0.769398
design	15.1776	1	70	0.737779
quality	10.4832	1	249	2.624384
storage	2.699	1	76	0.801017



BINARY DATA FILE

Weight Wix for the feature fi (Wix) = (frequency of fi in Cx) / (sum of the frequencies of features fi in Cx)



	money	product	service	nstallation	bed	quality	wood	design	storage	delivery
0	0	0	0	1.212065	0	0	2.400167	0	0	0
1	0	0	0	0	2.520235	2.624384	2.400167	0	0.801017	0

total
3.612232397
8.345803518

NEXT →

# CALCULATION OF WEIGHTS AND CLUSTERING

feature	total	clusters	freq	weights
product	178.3828	2	541	3.032804
wood	87.4033	0	334	2.400167
bed	55.2017	3	267	2.520235
money	50.7408	3	231	2.180428



BINARY DATA FILE

Weight  $W_{ix}$  for the feature  $f_i$  ( $W_{ix}$ ) = (frequency of  $f_i$  in  $C_x$ ) / (sum of the frequencies of features  $f_i$  in  $C_x$ )



	money	product	service	installation	bed	quality	wood	design	storage	delivery
0	0	0	0	1.212065	0	0	2.400167	0	0	0
1	0	0	0	0	2.520235	2.624384	2.400167	0	0.801017	0

total
3.612232397
8.345803518

NEXT →

# CALCULATION OF WEIGHTS AND CLUSTERING

feature	total	clusters	freq	weights
product	178.3828	2	541	3.032804
wood	97.4932	0	234	2.400167
bed	55.2017	3	267	2.520235
money	50.7408	3	231	2.180428
installatio	26.6609	1	115	1.212065
service	21.6191	1	139	1.465018
delivery	18.2396	1	73	0.769398
design	15.1776	1	70	0.737779
quality	10.4832	1	249	2.624384
storage	2.699	1	76	0.801017



BINARY DATA FILE

TOTAL EVALUATION OF  
CONSUMER SENTIMENT



Weight  $W_{ix}$  for the feature  $f_i$  ( $W_{ix}$ ) = (frequency of  $f_i$  in  $C_x$ ) / (sum of the frequencies of features  $f_i$  in  $C_x$ )



	money	product	service	installatio	bed	quality	wood	design	storage	delivery
0	0	0	0	1.212065	0	0	2.400167	0	0	0
1	0	0	0	0	2.520235	2.624384	2.400167	0	0.801017	0

total
3.612232397
8.345803518

NEXT →

# Results

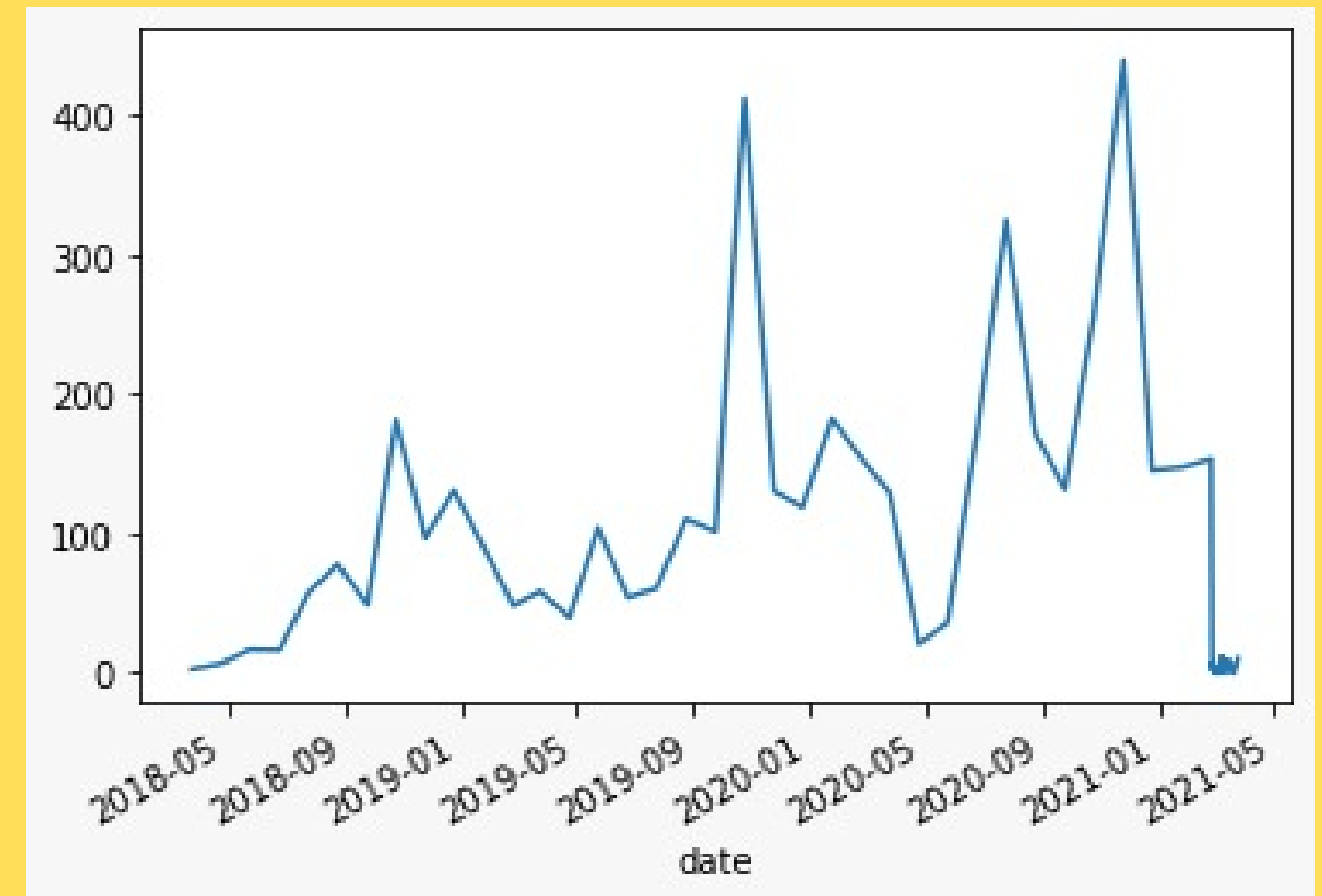


## OVERALL PRODUCT RATING



cluster	no. of reviews	mean	range	group
0	1243	0.074	1-1.21	insigni
1	910	2.686	1.46-3.86	signi
2	299	5.181	3.97-6.78	more signi
3	56	8.462	6.86-13.52	most signi

## MONEY



month	1	2	3	4	5	6	7	8	9	10	11	rating
year												
2018	NaN	NaN	4.000000	5.00	4.125000	3.125000	4.111111	3.621622	2.769231	3.988235	3.680556	3.490909
2019	3.625000	3.593750	3.538462	3.15	4.000000	4.034483	3.222222	3.444444	3.509091	3.943231	3.922078	4.012346
2020	4.101695	4.057692	4.075758	3.50	3.705882	4.103093	3.739884	3.525000	3.203125	4.033113	3.979592	3.825581
2021	3.812500	4.091743	4.017241	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN

This is the table with year and month wise ratings. It can be used to infer that in which month customers usually give higher ratings.

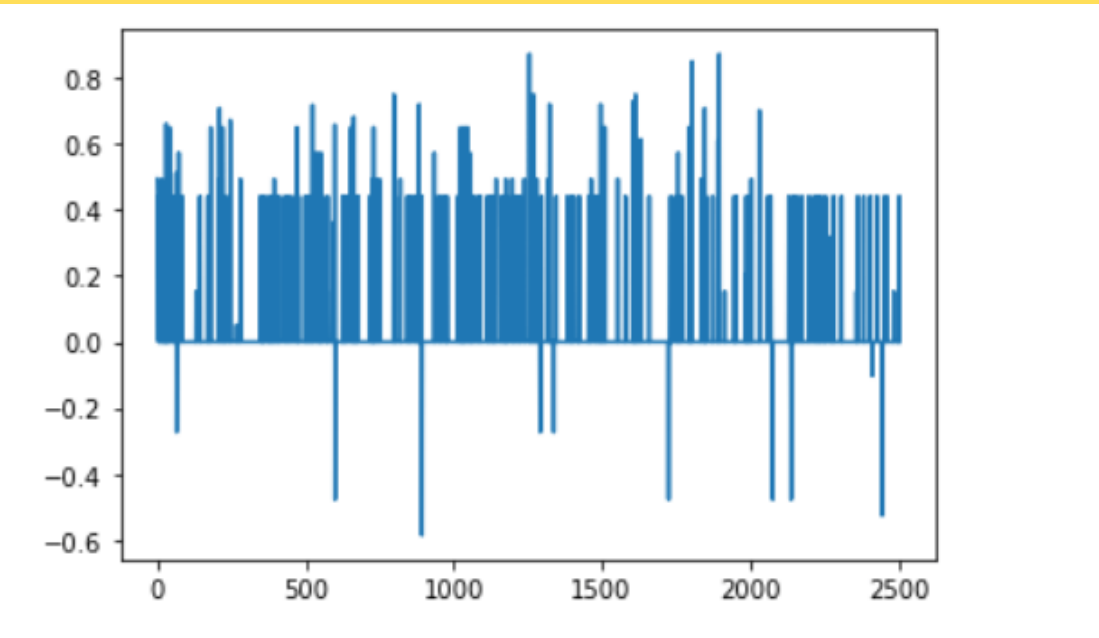
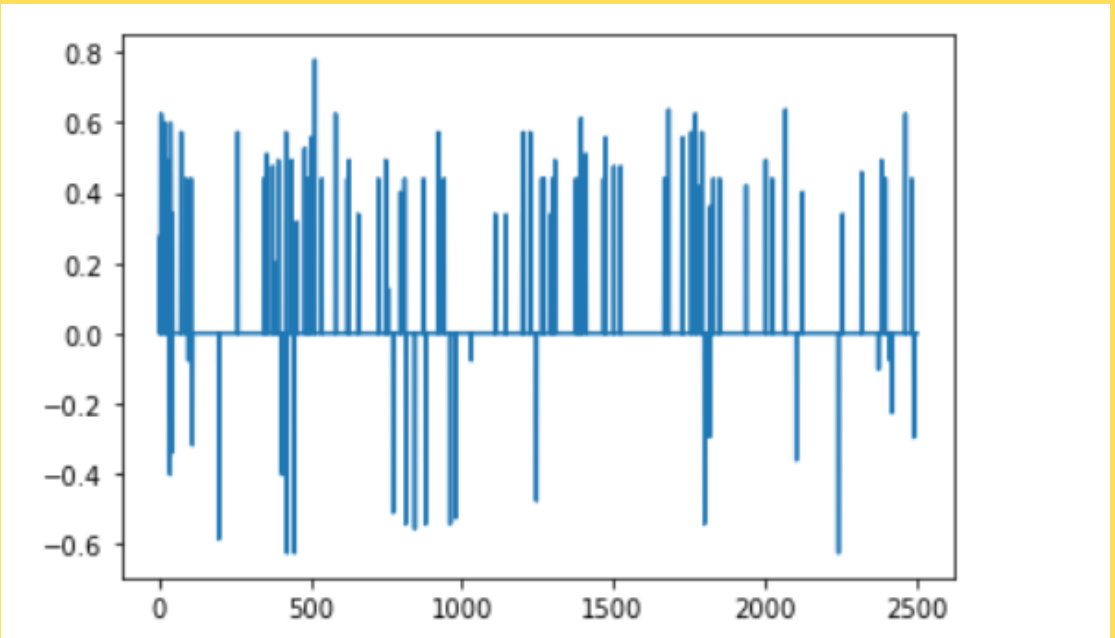
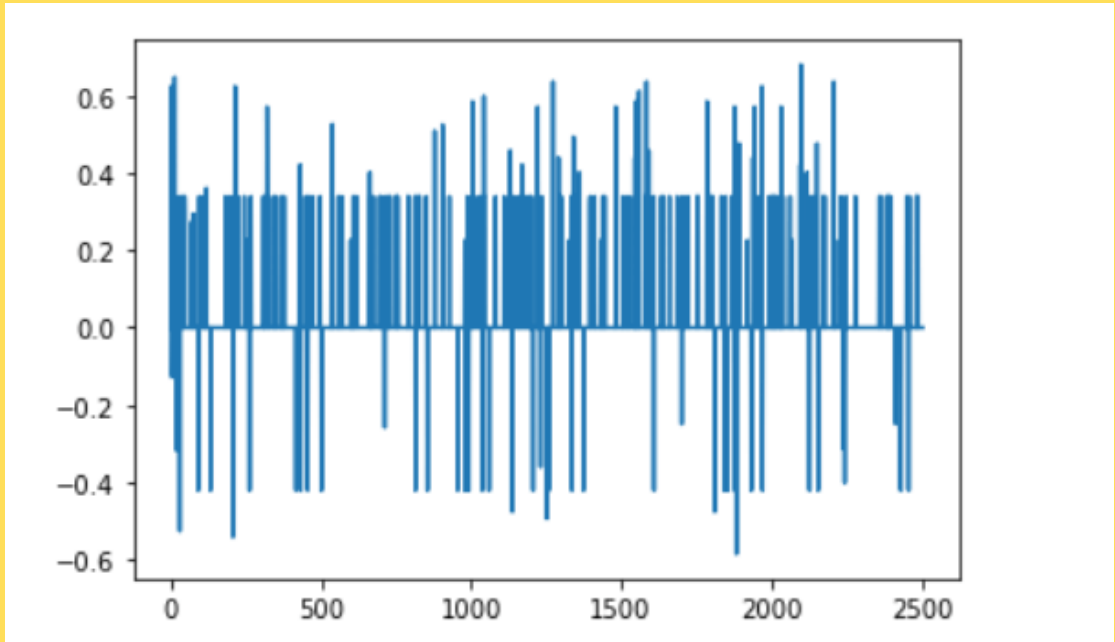
Usually in the months of April-May and Nov-Dec customers tend to give higher ratings. This maybe be because of heavy discounts given in Nov because of Diwali. On the other hand, events like big billion days and end of season sale could be reason for higher ratings in summers.

NEXT →

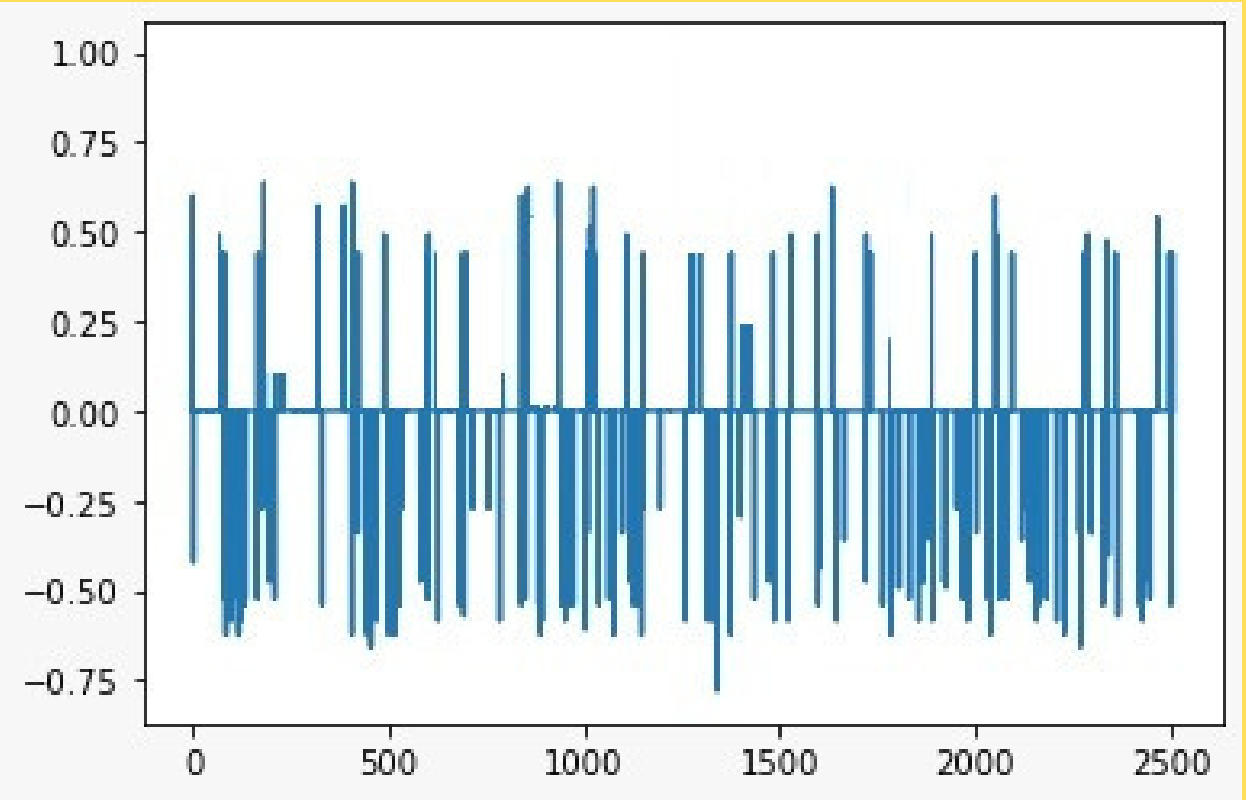
Features that have  
positive polarity

Features that have  
negative polarity

MONEY  
INSTALLATION  
WOOD

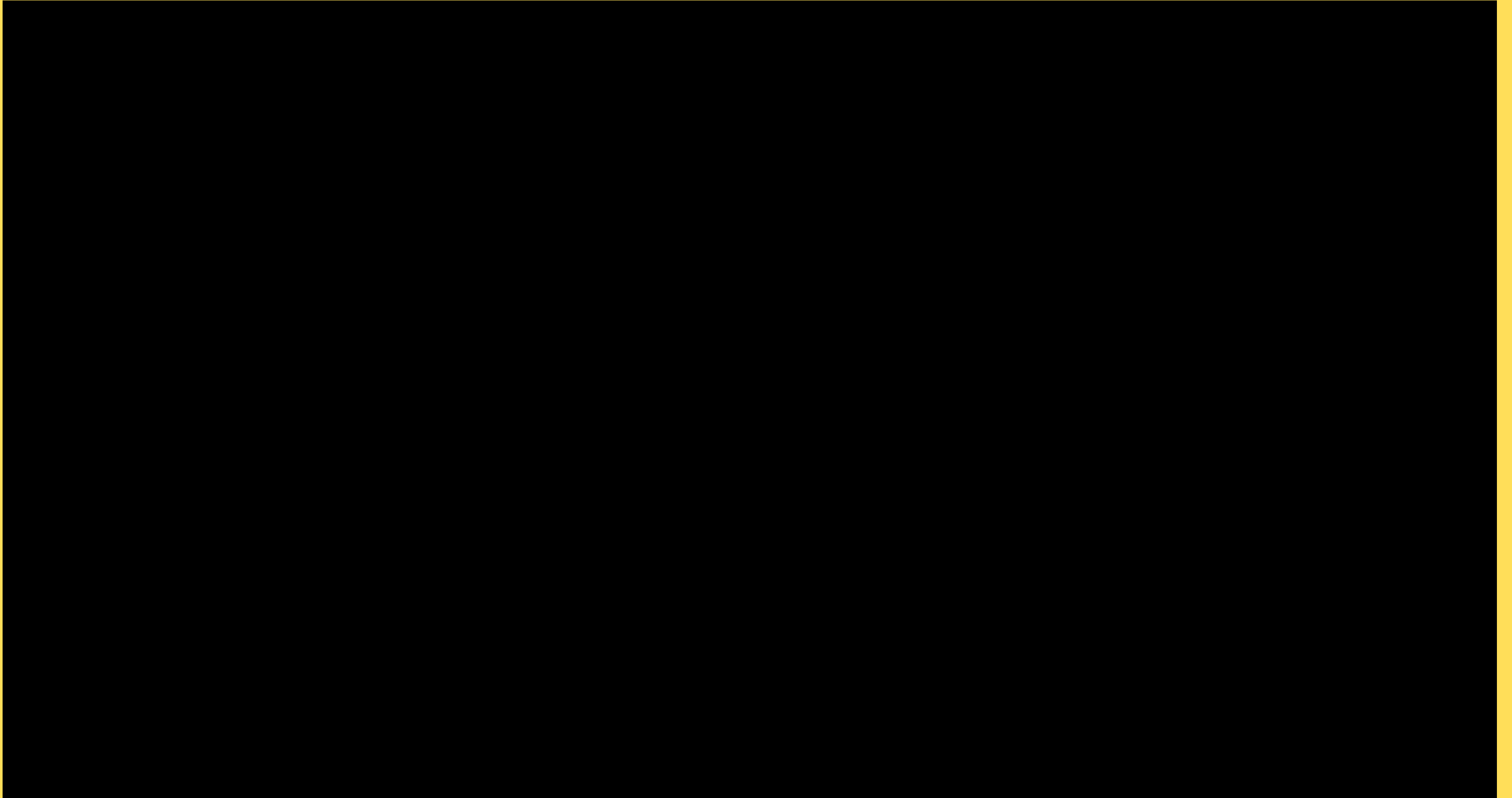


QUALITY



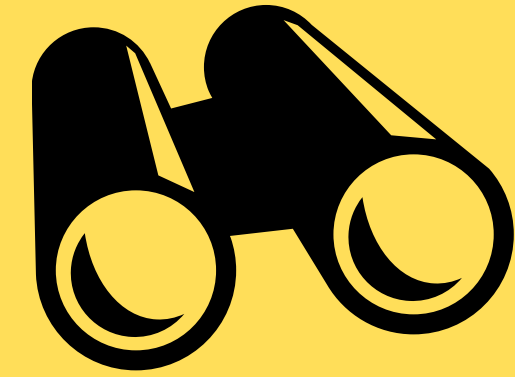
NEXT →

## QUICK DEMO - DEPLOYMENT USING HTML+FLASK





# Scope of Work



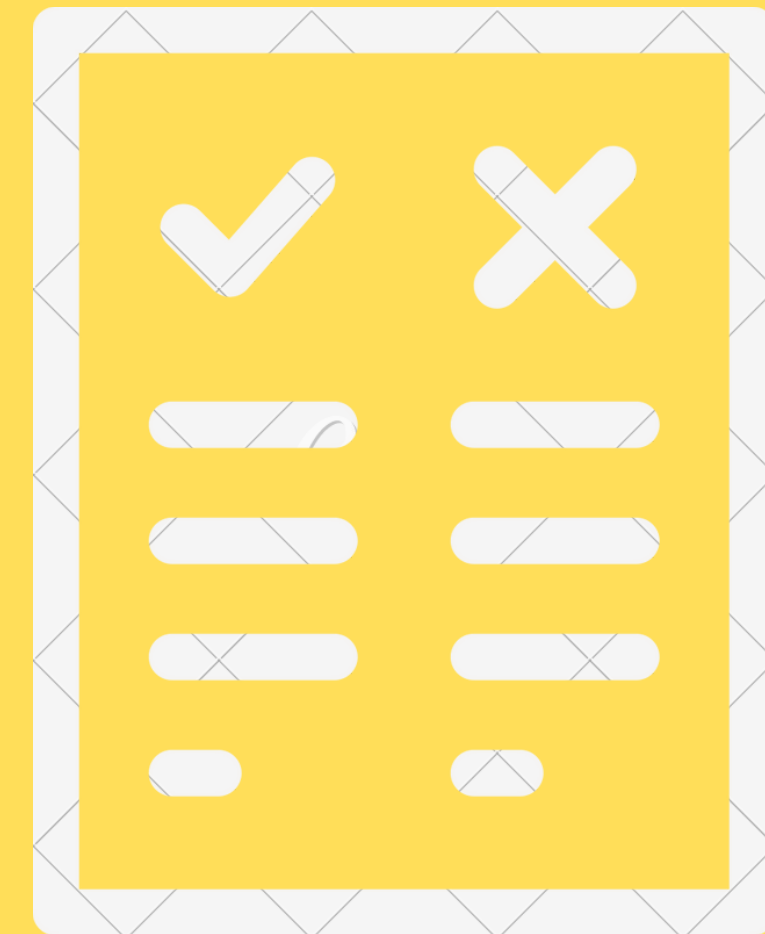
01

## Pros

- **easy** to understand and implement
- computationally **cheap**
- **dynamic** in nature
- **reduced manual intervention**

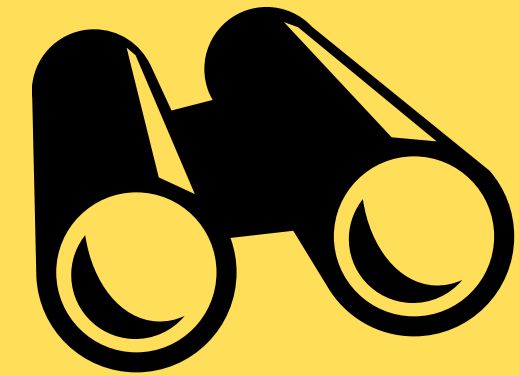
## Cons

- not feasible on newly launched products with **less reviews**
- not very effective on **poorly written reviews**
- ratings cannot be used since consumers **do not have a protocol to follow** while rating a product





# Scope of Work



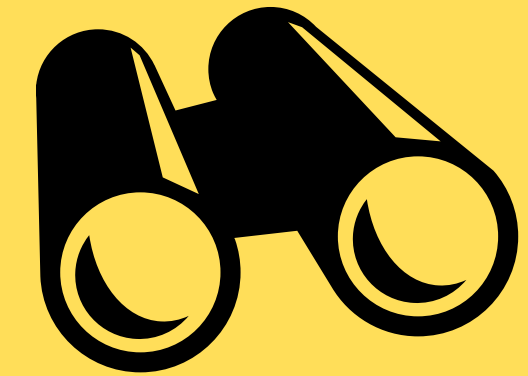
02



## Win-Win for everyone

Clustering reviews makes it **easier for buyers** to them. On the other hand, **extracting information** from these reviews via EDA, Time Series, etc. will help sellers to make the product better than before!

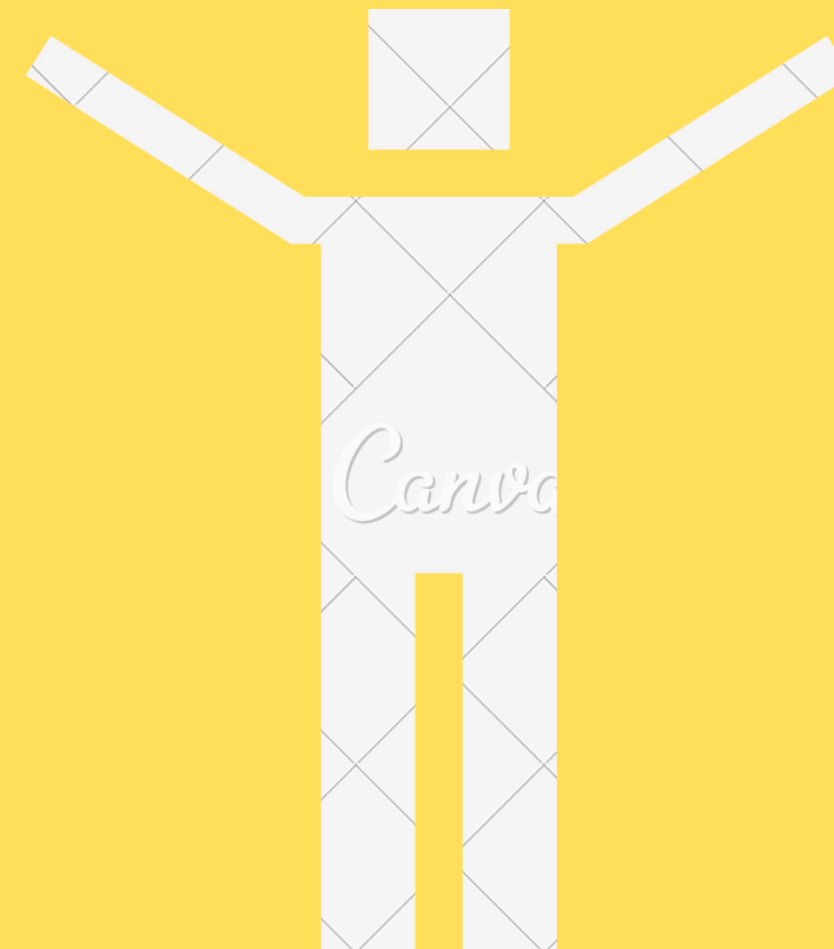
# Scope of Work



03

## Wider Adaptability

The project design is **flexible** to be implemented on any product. It is **adaptive** to capture all consumer experiences.





WHEN IN DOUBT, EXPERIMENT AND DECIDE

# Some Things That We Learnt



WHEN IN DOUBT, EXPERIMENT AND DECIDE



EVEN THE SIMPLEST DATA CAN HELP A LOT

B	C	D	E
rating	customer name	review title	reviews
5	Paritosh Pradhan	Wonderful	bed is broken with in 3 months... very poor quality
5	JAUNEET singh	Wonderful	EXCELLENT THIS PRICE 6999/-
5	Bhupender Pareek	Just wow!	good quality we liked the product
5	Lester fernandes	Excellent	Manjunath.S was very good and professional. Very fast installation.
5	Arif Siddiquie	Just wow!	The worst product ...quality is poor and received a damaged product ..installation was not done properly

# Some Things That We Learnt

NEXT →

# Some Things That We Learnt



**WHEN IN DOUBT, EXPERIMENT AND DECIDE**



**EVEN THE SIMPLEST DATA CAN HELP A LOT**



**THINGS MIGHT NOT GO ACCORDING TO THE  
PLAN SO ONE SHOULD LEARN TO IMPROVISE**



# Thank You

**GROUP 7**

**ABHIRUP SARKAR, NIPUN MOHINDRA, RUSHIKESH BADGUJAR,  
TANYA MANGATH AND VAMSITEJ GADIVEMUELA**