```
In [1]: from google.colab import drive
drive.mount('/content/MyDrive/')

Mounted at /content/MyDrive/
```

DESCRIPTION

Help a leading mobile brand understand the voice of the customer by analyzing the reviews of their product on Amazon and the topics that customers are talking about. You will perform topic modeling on specific parts of speech. You'll finally interpret the emerging topics.

Problem Statement:

A popular mobile phone brand, Lenovo has launched their budget smartphone in the Indian market. The client wants to understand the VOC (voice of the customer) on the product. This will be useful to not just evaluate the current product, but to also get some direction for developing the product pipeline. The client is particularly interested in the different aspects that customers care about. Product reviews by customers on a leading e-commerce site should provide a good view.

Domain: Amazon reviews for a leading phone brand

Analysis to be done: POS tagging, topic modeling usin

```
In [3]: import warnings
warnings.filterwarnings('ignore', category=DeprecationWarning)
```

In [29]: import pandas as pd
import nltk
nltk.download('all')

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In [9]: df = pd.read_csv('/content/MyDrive/MyDrive/NLP Simplilearn/Proj1/K8 Reviews v0.2.csv')
```

```
In [10]: df.head()
                             Good but need updates and improvements
                     0 Worst mobile i have bought ever, Battery is dr...
                      1 when I will get my 10% cash back.... its alrea...
                                                             Good
                     0 The worst phone everThey have changed the last...
In [14]: df = df.drop(['sentiment'],axis=1)
In [15]: df.head()
Out[15]:
                    Good but need updates and improvements
           1 Worst mobile i have bought ever, Battery is dr...
            2 when I will get my 10% cash back.... its alrea...
                                                   Good
            4 The worst phone everThey have changed the last...
In [16]: df.shape
Out[16]: (14675, 1)
```

Data Pre-Processing

Replacing/Dropping NULL values

Converting to LOWER case

```
In [18]: df['clean_review'] = df['review'].apply(lambda x: str(x).lower())
df.head()

Out[18]: review clean_review

0 Good but need updates and improvements good but need updates and improvements
1 Worst mobile i have bought ever, Battery is dr...
2 when I will get my 10% cash back.... its alrea...
3 Good good
4 The worst phone everThey have changed the last... the worst phone everthey have changed the last...
```

REMOVE NON-ALPHA DATA(DIGITS, PUNCTUATIONS, DIACRITICS)

```
In [19]: df['clean_review'] = df['clean_review'].str.replace(r'[^a-zA-Z\s]', '',regex=True)

Out[19]: review clean_review

O Good but need updates and improvements good but need updates and improvements

1 Worst mobile i have bought ever, Battery is dr... worst mobile i have bought ever battery is dr...

2 when I will get my 10% cash back... its alrea... when I will get my cash back its alrea...

3 Good good

4 The worst phone everThey have changed the last... the worst phone everthey have changed the last...
```

REMOVING WHITE SPACE

```
In [20]: df['clean_review'] = df['clean_review'].str.replace(r'\s{2,}', '',regex=True)

Out[20]:

review clean_review

O Good but need updates and improvements

Worst mobile i have bought ever, Battery is dr...

worst mobile i have bought ever, Battery is dr...

worst mobile i have bought ever, Battery is dr...

worst mobile i have bought ever, Battery is dr...

when I will get my 10% cash back... its alrea...

when I will get my cash back its already january

3 Good good

4 The worst phone everThey have changed the last... the worst phone everthey have changed the last...
```

WORD TOKENIZATION

```
In [22]: df['clean_review'] = df['clean_review'].apply(lambda x: word_tokenize(x))

Out[22]: review clean_review

O Good but need updates and improvements [good, but, need, updates, and, improvements]

Norst mobile i have bought ever, Battery is dr... [worst, mobile, i, have, bought, ever, battery...]

when I will get my 10% cash back... its alrea... [when, i, will, get, my, cash, back, its, alre...]

The worst phone everThey have changed the last... [the, worst, phone, everthey, have, changed, t...]
```

REMOVE UNNECESSARY WORDS

LEMMATIZATION

```
In [30]: 
df['clean_review'] = df['clean_review'].apply\
    (lambda x: [WordNetLemmatizer().lemmatize(word) for word in x])
    df.head()

Out[30]:
```

review clean_review

0 Good but need updates and improvements [good, need, update, improvement]

1 Worst mobile i have bought ever, Battery is dr... [worst, mobile, bought, ever, battery, drainin...

2 when I will get my 10% cash back.... its alrea... [cash, back, already, january]

3 The worst phone everThey have changed the last... [worst, phone, everthey, changed, last, phone,...

4 Only I'm telling don't buyl'm totally disappoi... [telling, buyi, totally, disappointedpoor, bat...

Extracting only NOUN

```
In [31]: dff'clean_review'] = dff'clean_review'].apply\
(lambda x: [word for word in x if nltk.pos_tag([word])[0][1] == 'NN'])

In [32]: df = df[df['clean_review'].map(lambda x: len(x)) > 1].reset_index(drop=True)

# Keeping records with more than single words

In [34]: df.head()

Out[34]: review clean_review

0 Good but need updates and improvements [need, update, improvement]

1 Worst mobile i have bought ever, Battery is dr... [mobile, bought, battery, hell, backup, hour, ...
```

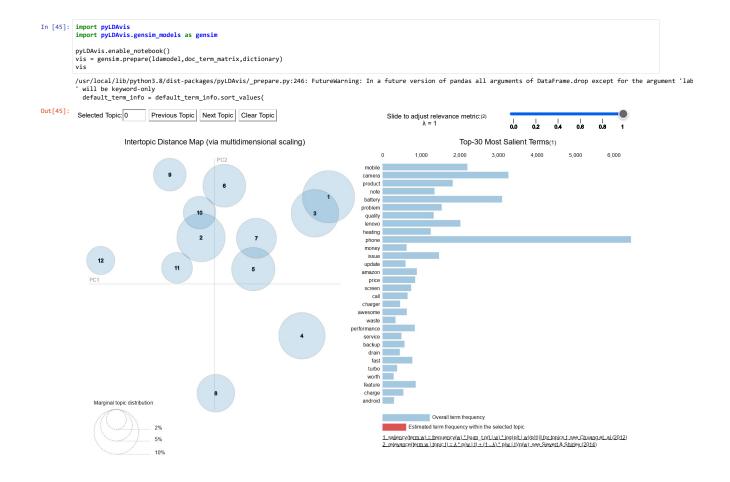
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2 when I will get my 10% cash back... its alrea...
3 The worst phone everThey have changed the last... [phone, everthey, phone, problem, amazon, phon...
4 Only I'm telling don't buyI'm totally disappoi... [buyi, disappointedpoor, batterypoor, camerawa...

Document Term Matrix

LDA

```
In [39]: from IPython.display import clear_output
In [40]: Lda = gensim.models.ldamodel.LdaModel
                  ldamodel = Lda(corpus=doc_term_matrix, num_topics=12, id2word=dictionary, passes=10,random_state=45)
                 clear_output()
                 # corpus requires document term matrix
# num topics is used to define number of topics to create from corpus
# id2word requires mapping of words
# passes is used to define number of iterations
In [41]: ldamodel.print_topics()
                  # We have printed all 12 topics and their keywords generated by LDA
Out[41]: [(0, '0.199*"camera" + 0.099*"quality" + 0.041*"phone" + 0.031*"sound" + 0.026*"front" + 0.025*"battery" + 0.022*"mode" + 0.019*"depth" + 0.017*"rear" + 0.016*"feature"'),
                   (1, "0.057*"android" + 0.042*"phone" + 0.034*"feature" + 0.031*"stock" + 0.028*"card" + 0.026*"contact" + 0.022*"user" + 0.021*"memory" + 0.020*"headphone" + 0.017*"slot"')
                   (2, "0.315*"mobile" + 0.162*"problem" + 0.091*"heating" + 0.031*"battery" + 0.022*"heat" + 0.014*"network" + 0.012*"game" + 0.008*"month" + 0.007*"class" + 0.007*"hang"'),
                   (3, "0.062*"phone" + 0.060*"screen" + 0.058*"charger" + 0.048*"turbo" + 0.039*"feature" + 0.027*"glass" + 0.018*"gorilla" + 0.017*"time" + 0.017*"charge" + 0.015*"core"'),
                   (4, '0.120*"update" + 0.053*"phone" + 0.049*"software" + 0.034*"need" + 0.034*"system" + 0.028*"oreo" + 0.026*"problem" + 0.019*"lenovo" + 0.013*"bill" + 0.012*"please"'),
                   (5, "0.196*"phone" + 0.011*"battery" + 0.057*"price" + 0.052*"camera" + 0.050*"awesome" + 0.047*"performance" + 0.044*"backup" + 0.027*"range" + 0.027*"life" + 0.020*"super
                   (6, "0.129*"battery" + 0.100*"issue" + 0.057*"heating" + 0.047*"fast" + 0.042*"phone" + 0.042*"drain" + 0.039*"hour" + 0.038*"charge" + 0.028*"time" + 0.021*"usage"'),
                   (7, "0.297*"product" + 0.036*"price" + 0.029*"excellent" + 0.018*"performance" + 0.013*"awesome" + 0.013*"till" + 0.013*"amazon" + 0.012*"expectation" + 0.012*"feature" + 0.013*"awesome" + 0.013*"awesome" + 0.013*"awesome" + 0.013*"awazon" + 0.012*"expectation" + 0.012*"feature" + 0.013*"awazon" + 0.012*"awazon" + 0.012*"awazon + 0.012*"awazon + 0.012*"awazon + 0.012**
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                      10,
'0.104*"phone" + 0.071*"amazon" + 0.043*"service" + 0.035*"lenovo" + 0.034*"return" + 0.023*"day" + 0.022*"please" + 0.020*"product" + 0.020*"problem" + 0.019*"custome
                   (11, '01.160*"money" + 0.086*"waste" + 0.074*"worth" + 0.061*"value" + 0.032*"delivery" + 0.017*"super" + 0.013*"buying" + 0.012*"facility" + 0.009*"dont" + 0.008*"iron"')]
```

Visualizing LDA model topics



Since, some topics in above graph are overlapping each other we will try to find optimal number of topics.

	Coherence_Score
Topic	
6	0.609116
10	0.605270
7	0.602581
8	0.597894
3	0.589486
2	0.582708
9	0.582576
5	0.579579
12	0.575871
13	0.571891
11	0.568505
4	0.568473
14	0.555884

```
In [55]: import matplotlib.pvplot as plt
              plt.plot(topic_num,cv_score,color='red', marker='o', linestyle='dashed')
plt.xticks(range(2,15))
plt.xlabel('Number of topics')
plt.ylabel('Coherence score')
              plt.show()
                 0.61
                  0.60
               e 0.59
                 0.58
                  0.57
                                                   9 10 11 12 13 14
we will be going with number of topic 6 as with 8 topics there will be many overlaps
   In [56]: # Creating LDA model with number of topics as 6
              Lda = gensim.models.ldamodel.LdaModel
               ldamodel = Lda(doc_term_matrix, num_topics=6, id2word=dictionary, passes=10,random_state=45)
              0.6091161154634883
   In [57]: ldamodel.print_topics()
  Out[57]: [(0, '0.156*"camera" + 0.078*"quality" + 0.039*"phone" + 0.022*"sound" + 0.021*"front" + 0.018*"mode" + 0.015*"depth" + 0.014*"performance" + 0.014*"display" + 0.014*"rear":
               (2, '0.154*"mobile" + 0.084*"problem" + 0.046*"heating" + 0.038*"product" + 0.038*"amazon" + 0.034*"issue" + 0.022*"return" + 0.018*"network" + 0.015*"lenovo" + 0.013*"tim
               (3, (3, "0.076*"phone" + 0.036*"money" + 0.034*"screen" + 0.025*"charger" + 0.022*"product" + 0.021*"lenovo" + 0.021*"turbo" + 0.019*"waste" + 0.014*"amazon" + 0.013*"value"'),
               (4, "0.086*"phone" + 0.040*"update" + 0.038*"issue" + 0.026*"problem" + 0.025*"service" + 0.022*"lenovo" + 0.021*"network" + 0.021*"software" + 0.013*"volta" + 0.012*"call"
                  .
0.135*"battery" + 0.126*"phone" + 0.035*"price" + 0.030*"awesome" + 0.030*"camera" + 0.030*"fast" + 0.029*"performance" + 0.027*"backup" + 0.023*"product" + 0.022*"hea
   In [60]: pyLDAvis.gensim_models.prepare(ldamodel,doc_term_matrix,dictionary)
              /usr/local/lib/python3.8/dist-packages/pytDAvis/_prepare.py:246: FutureWarning: In a future version of pandas all arguments of DataFrame.drop except for the argument 'lab
                will be keyword-only
default_term_info = default_term_info.sort_values(
   Out[60]: Selected Topic: 0
                                  Previous Topic Next Topic Clear Topic
                                                                                                                    Slide to adjust relevance metric:(2)
                              Intertopic Distance Map (via multidimensional scaling)
                                                                                                                                           Top-30 Most Salient Terms(1)
                                                                                                                                         2,000
                                                                                                                             1,000
                                                                                                                                                     3,000
                                                                                                                                                                  4,000
                                                                                                                                                                               5,000
                                                                                                                                                                                           6,000
                                                                                                            mobile
camera
battery
                                                                                                             quality
                                                                                                              note
                                                                                                            problem
                                                                                                            phone
lenovo
heating
                                                                                                             money
                                                                                                             update
                                                                                                             issue
                                                                                                           awesome
                                                                                                            amazon
screen
backup
                                                                                                             price
                                                                                                        performance
                                                                                                            charger
                                                                                                             turbo
                                                                                                             service
                                                                                                           excellent
                                                                                                             sound
                     Marginal topic distribution
                                                                                                                                Overall term frequency
                                        2%
                                                                                                                   1_{\text{saliency}(\text{lerm } w) = \text{frequency}(w)}^*_{\text{sum } \text{tp}(\text{L}|w)}^*_{\text{log}(p(\text{L}|w)p(\text{I}))}, \text{for topics.t. see. } \text{Chuang.et. al.}(2012) } \\ 2_{\text{relewance}(\text{lerm } w | \text{lopic.t}) = \lambda^*_{p(w), \text{L}}}^*_{\text{lopic.t}} + (1-\lambda)^*_{p(w), \text{L}}^*_{p(w), \text{L}}^*_{p(w), \text{L}}}^*_{\text{log}(w), \text{see. Sievert.\& Shirley.}}(2014) } 
                                        5%
   In [61]: df.head()
   Out[61]:
                                                                                            clean_review
                                                                                [need, update, improvement]
                         Good but need updates and improvements
                      Worst mobile i have bought ever, Battery is dr...
                                                                  [mobile, bought, battery, hell, backup, hour, ...
               2
                      when I will get my 10% cash back.... its alrea...
```

Creating a lookup table for topics

The worst phone everThey have changed the last... [phone, everthey, phone, problem, amazon, phon...

Only I'm telling don't buyl'm totally disappoi... [buyl, disappointedpoor, batterypoor, camerawa...

Topic_Number Top_ic_Name Top_Keywords

0 0 Camera, Sound [Camera', 'quality', 'phone', 'sound', 'front', 'mode', 'depth', 'performance', 'display', 'tear']

1 1 1 Mixed issues [Incle', 'lenovo', 'phone', 'call', 'feature', 'android', 'product', 'option', 'speaker', 'stock']

2 2 1 Heating issue [Incle', 'problem', 'heating', 'product', 'amazon', 'ssew', 'return', heatwork', 'lenovo', 'lenovo', 'lenovo', 'lenovo', 'lenovo', 'lenovo', 'lenovo', 'lenovo', 'lenovo', 'network', 'software', 'volta', 'call', 'lenovo', 'lenovo', 'network', 'software', 'volta', 'call', 'lenovo', 'lenovo',

Creating new columns and inserting topic numbers and names

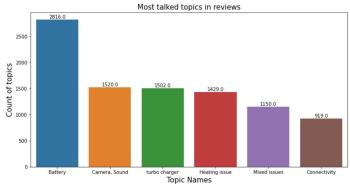
In [66]: df.head()

0	114	Гζ	د ء	
U	uс	Įυ	U.	

	review	clean_review	Topic_Number	Topic_Name
0	Good but need updates and improvements	[need, update, improvement]	0,4	Camera, Sound & Connectivity
1	Worst mobile i have bought ever, Battery is dr	[mobile, bought, battery, hell, backup, hour,	3	turbo charger
2	when I will get my 10% cash back its alrea	[cash, january]	0	Camera, Sound
3	The worst phone everThey have changed the last	[phone, everthey, phone, problem, amazon, phon	3	turbo charger
4	Only I'm telling don't buyI'm totally disappoi	[buyi, disappointedpoor, batterypoor, camerawa	4	Connectivity

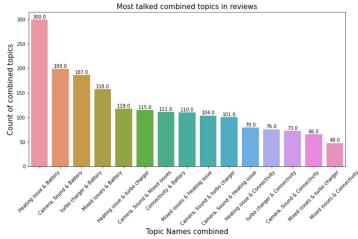
Visualization

```
In [68]: import seaborn as sns
In [68]: plt.figure(figsize=(12,6))
    ax = sns.barplot(x=df.Topic_Name.value_counts()[:6].index,y=df.Topic_Name.value_counts()[:6].values)
    for p in ax.patches:
        ax.annotate(p.get_height(), (p.get_x() + p.get_width() / 2., p.get_height()+50),ha = 'center', va = 'center')
    plt.xlabel('Topic Names',size=15)
    plt.ylabel('Count of topics',size=15)
    plt.title('Most talked topics in reviews',size=15)
    plt.show()
```



From above graph we can say that most of customers had issues with Battery of mobile

```
In [69]: plt.figure(figsize=(12,6))
    ax = sns.barplot(x=df.Topic_Name.value_counts()[6:].index,y=df.Topic_Name.value_counts()[6:].values)
    for p in ax.patches:
        ax.annotate(p.get_neight(), (p.get_x() + p.get_width() / 2., p.get_height()+5),ha = 'center', va = 'center')
    plt.xlabel('Topic Names combined', size=15)
    plt.ylabel('Count of combined topics', size=15)
    plt.title('Most talked combined topics in reviews', size=15)
    plt.xticks(rotation=45)
    plt.show()
```



From above graph we can say that most of customers had combined issues with,

- Heating issue & Battery
 Camera, Sound & Battery
- 3. turbo charger & Battery

```
In [70]: #Extracting reviews of 5 topic(review of battery)
df.loc[df.Topic_Number.str.contains('5'),['review','Topic_Name']].head(10)\
.style.set_properties(subset=['review'], **{'width': '300px'})
```

Out[70]:

Topic_Name	review	
Batter	Phone is awesome. But while charging, it heats up allotReally a genuine reason to hate Lenovo k8 note	5
Heating issue & Batter	Don't purchase this item, It is so much of heating &Battery life is very poor	10
Batter	Very good phone slim good battry backup good screen love it	12
Heating issue & Batter	Battery draining very rapidly I don't know whyTell me possible solutions for battery life	15
Batter	Excellent camera , excellent speed.excellent features.excelent battery.	17
Batter	It is not a very good product camera are very poorOs is not goodBattery draining very quicklyLike a odinary phoneIt was fully unexpected product from Lenovo	18
Heating issue & Batter	Awesome phone in this price and this is my second mobile from lenovo. It is fast and display has been improved.	21
Connectivity & Batter	Before the new update of 8.0 Oreo, it worked superbly, the battery back-up is also superb and there is not that much heating problem But After that update, my phone got heating up simply, battery is also draining unnecessarily really very much disappointed after that update of 8.0 Oreo	24
Batter	Good performance but the battery gets oveheated	26
Batter	Best camera and better backup is very bestIn this priceFull passa wasole phone	27

In []: