# Analysis on Public Perception Towards LSE on Reddit

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Github repository: link

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# Introduction

### **Motivation**

As undergraduate students at the London School of Economics and Political Science (LSE), we want to know what is the public's general perception about the university we are currently pursuing our degrees in. Through this project, we hope to understand of how reputable LSE is in the public's eyes, and how a degree from LSE could be beneficial in the workforce.

# Goal of the study

We hope to answer the following research questions:

- 1. What are some common traits that the public associate LSE with?
- 2. What do people on Reddit want to know more about LSE?

# **Data Acquisition**

# **Reddit posts**

To obtain the data required for our analysis, we used reddit API, which allows us to scrape up to 1000 posts from a subreddit. We collected our data on **15 April 2023**, where we managed to scrape 851 new posts from the subreddit "TheLse". For each post, we collected information about its title, content, and score (the number of upvotes minus the number of downvotes).

The code to acquire the data is stored in another notebook data\_acquisition.ipynb , and the data is stored in the file data.csv inside the github repository. The data we got is shown below:

```
In [1]: import pandas as pd

df = pd.read_csv('data.csv')
    df
```

Out[1]:

title content score 0 LSE accomodation for couples Hi! \n\nI will be moving with my partner to Lo... 3 which accommodation do rich international 1 0 where do rich students live during their studi... Im a general course student and can't choose 2 Bankside or Garden Hall 2 3 Is the reserve list separate from an ordinary ... I got placed on a reserve list for the Masters... 2 4 Postgraduate housing Hello! I was just accepted into LSE for my mas... 2 Update to waiting for decades + advice for Hi! I made [a post] 846 12 (https://www.reddit.com/r/T... Advice on Teachers Reference for Undergrad As part of my undergrad application to LSE 847 2 Арр... for... 848 international undergrad applicants??? a bunch of people i know still haven't receive... 3 So am I wrong or are the winter and 849 Winter/March holidays 1 March/Apri... 850 Different price for the same room? 2 Hi. Was checking room availability at Passfiel...

851 rows × 3 columns

# **Data Cleaning**

We performed the following steps to prepare our data for analysis:

- 1. Handle missing values
- 2. Convert dataframe column types
- 3. Lower all cases
- 4. Remove stop words and punctuations for the title and content of the post
- 5. Convert words into its dictionary form
- 6. Found 100 most common words in the title and content of the posts
- 7. Found 100 most common words in the title and content of the 50 top scoring posts

```
In [2]: # remove empty values
df.dropna(inplace = True)
df.shape
Out[2]: (703, 3)
```

After removing the empty values from the dataframe, there are only 703 posts left to be analyzed.

```
In [3]: #check dataframe type
df.dtypes
```

```
title
                   object
Out[3]:
                   object
        content
        score
                    int64
        dtype: object
In [4]: #make sure the title and content have the correct type
        df['title'] = df['title'].astype(str)
        df['content'] = df['content'].astype(str)
        #lower all cases
In [5]:
        df['title'] = df['title'].apply(str.lower)
        df['content'] = df['content'].apply(str.lower)
In [6]: # remove stopwords
        import nltk
        from nltk.corpus import stopwords
         sw = stopwords.words('English')
         stop words = set(stopwords.words('english'))
        df['title'] = df['title'].apply(lambda x: ''.join([word for word in x.split() if word
         df['content'] = df['content'].apply(lambda x: ' '.join([word for word in x.split() if
In [7]:
        # remove the puntuations
        import string
         def remove punctuations(text):
            if isinstance(text, str):
                 translator = str.maketrans('', '', string.punctuation)
                 return text.translate(translator)
            else:
                 return text
        df = df.fillna('')
        df = df.applymap(remove_punctuations)
        #convert words to its Lemma
In [8]:
        from nltk.tokenize import word tokenize
        from nltk.stem import WordNetLemmatizer
        lemmatizer = WordNetLemmatizer()
        def lemmatize_text(text):
            return ' '.join([lemmatizer.lemmatize(w) for w in word_tokenize(text)])
        df['title'] = df['title'].apply(lemmatize_text)
        df['content'] = df['content'].apply(lemmatize text)
        #Look at the cleaned data
In [9]:
        df = df.reset_index(drop=True)
        df
```

Out[9]: title content score

0	Ise accomodation couple	hi moving partner london looking couple accomm	
1	accommodation rich international student stay in	rich student live study arab chinese etc	
2	bankside garden hall	im general course student cant choose two advice	
3	reserve list separate ordinary waitlist	got placed reserve list master environmental p	
4	postgraduate housing hello accepted lse master wondering accommodat		2
•••			
698	update waiting decade advice future applicant lse	hi made a posthttpswwwredditcomrthelsecomments	
699	advice teacher reference undergrad application	part undergrad application lse bachelor histor	
700	international undergrad applicant	bunch people know still haven 't received res	3
701	wintermarch holiday	wrong winter marchapril holiday essentially st	1
702	differemt price room	hi checking room availability passfield two op	2

703 rows × 3 columns

### Prepare data for analysis

```
# Find 50 most common words for post titles
In [10]:
         import collections
          all_text_title = ' '.join(df['title'])
          title counts = collections.Counter(all text title.split())
          common titles = title counts.most common(50)
          print(common titles)
         [('lse', 193), ('msc', 82), ('student', 70), ('application', 44), ('accommodation', 4
         3), ('economics', 38), ('graduate', 36), ('course', 35), ('v', 35), ('bsc', 34), ('su
         mmer', 28), ('undergrad', 26), ('offer', 26), ('international', 25), ('admission', 2
         5), ('finance', 25), ('year', 25), ('hall', 24), ('question', 22), ('master', 22),
         ('help', 21), ('advice', 20), ('support', 19), ('school', 18), ('get', 17), ('undergr
         aduate', 17), ('political', 16), ('general', 16), ('anyone', 16), ('program', 16),
         ('chance', 15), ('online', 15), ('housing', 14), ('group', 14), ('history', 14), ('ap
         plying', 14), ('apply', 13), ('financial', 13), ('amp', 13), ('social', 13), ('lookin
         g', 13), ('bankside', 12), ('g', 12), ('science', 12), ('urbanest', 11), ('economy',
         11), ('economic', 11), ('late', 11), ('management', 11), ('degree', 11)]
In [11]: # Find 50 most common words for post contents
         all text_content = ' '.join(df['content'])
          content counts = collections.Counter(all text content.split())
          common contents = content counts.most common(50)
          print(common contents)
```

```
[('lse', 571), ('',', 344), ('student', 310), ('would', 300), ('im', 270), ('year', 24
         0), ('course', 219), ('get', 199), ('i', 191), ('anyone', 180), ('know', 179), ('offe
         r', 174), ('application', 173), ('hi', 170), ('msc', 145), ('like', 138), ('program',
         135), ('m', 122), ('economics', 120), ('thanks', 116), ('international', 116), ('als
         o', 113), ('school', 107), ('one', 106), ('got', 103), ('experience', 99), ('lookin
         g', 98), ('degree', 98), ('wondering', 94), ('master', 92), ('applied', 92), ('want',
         91), ('need', 90), ('finance', 90), ('time', 89), ('university', 89), ('apply', 89),
         ('study', 88), ('it', 84), ('hello', 84), ('help', 84), ('math', 83), ('first', 83),
         ('thank', 81), ('s', 80), ('currently', 79), ('really', 77), ('please', 77), ('good',
         76), ('people', 75)]
         # Find 50 most common words for the titles of the highest scoring 50 posts
In [12]:
         highest_scoring = df.sort_values(['score'], ascending = False).groupby('score').head(5
         top_text_title = ' '.join(highest_scoring['title'])
         top title counts = collections.Counter(top text title.split())
         top common titles = top title counts.most common(50)
         print(top_common_titles)
         [('lse', 113), ('msc', 45), ('student', 33), ('v', 26), ('application', 25), ('accomm
         odation', 22), ('graduate', 21), ('undergrad', 16), ('course', 16), ('offer', 15),
         ('economics', 15), ('help', 13), ('summer', 13), ('master', 13), ('advice', 12), ('in
         ternational', 12), ('political', 12), ('question', 12), ('school', 11), ('online', 1
         1), ('get', 11), ('undergraduate', 11), ('bsc', 11), ('support', 10), ('history', 1
         0), ('year', 10), ('applying', 10), ('admission', 10), ('group', 9), ('program', 9),
         ('got', 8), ('economy', 8), ('g', 8), ('looking', 8), ('hall', 8), ('economic', 8),
         ('ucl', 7), ('london', 7), ('entry', 7), ('accomodation', 7), ('science', 7), ('housi
         ng', 7), ('scheme', 7), ('urbanest', 7), ('uk', 7), ('social', 7), ('society', 7),
         ('degree', 7), ('apply', 7), ('anyone', 7)]
In [13]: # Find 50 most common words for the content of the highest scoring 50 posts
         top_text_content = ' '.join(highest_scoring['content'])
         top content counts = collections.Counter(top text content.split())
         top common contents = top content counts.most common(50)
          print(top_common_contents)
         [('lse', 330), ('', 204), ('student', 183), ('im', 157), ('would', 142), ('year', 13
         5), ('get', 118), ('course', 109), ('offer', 104), ('i', 103), ('know', 98), ('hi', 9
         0), ('application', 87), ('anyone', 86), ('like', 83), ('program', 79), ('msc', 78),
         ('m', 71), ('school', 71), ('thanks', 71), ('also', 66), ('international', 65), ('eco
         nomics', 64), ('looking', 61), ('one', 59), ('got', 57), ('s', 55), ('need', 54), ('s
         tudy', 53), ('help', 52), ('degree', 51), ('university', 51), ('want', 50), ('financ
         e', 50), ('experience', 50), ('wondering', 47), ('it', 47), ('math', 47), ('master',
         46), ('good', 46), ('really', 45), ('received', 45), ('2', 43), ('ive', 43), ('firs
         t', 43), ('applied', 42), ('hello', 42), ('thank', 42), ('please', 42), ('grade', 4
         2)]
```

```
In [14]: import pickle
         # Save the list to a file for data analysis
         with open('words.pkl', 'wb') as file:
             pickle.dump(common titles, file)
              pickle.dump(common_contents, file)
              pickle.dump(top common titles, file)
              pickle.dump(top common contents, file)
```

# **Data Analysis**

# **Exploratory Data Analysis (EDA)**

To perform the exploratory data analysis, we load the list obtained from the previous section.

```
In [15]: import pickle

# Load the list from the file
with open('words.pkl', 'rb') as file:
    titles = pickle.load(file)
    contents = pickle.load(file)
    top_titles = pickle.load(file)
    top_contents = pickle.load(file)

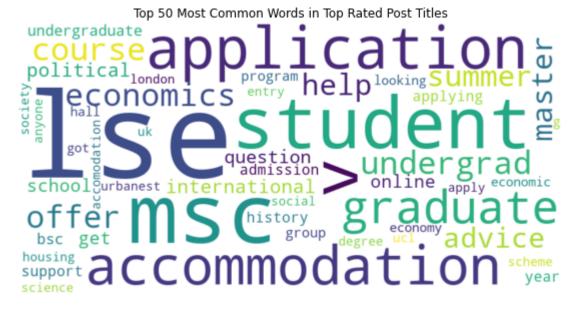
titles_d = dict(titles)
    contents_d = dict(contents)
    top_titles_d = dict(top_titles)
    top_contents_d = dict(top_contents)
```

Firstly, we loaded the top 50 common words in top rated post titles and plot a word cloud to visualize the size of the words in proportion with its frequency.

```
import matplotlib.pyplot as plt
from wordcloud import WordCloud

# Create a word cloud for 50 most common words in the title of the highest scoring 50
wordcloud = WordCloud(width=800, height=400, background_color="white").generate_from_f

# Plot the word cloud
plt.figure(figsize=(10, 5))
plt.imshow(wordcloud, interpolation='bilinear')
plt.title('Top 50 Most Common Words in Top Rated Post Titles')
plt.axis('off')
plt.show()
```



• The plot above shows that 'lse', 'application', 'student', 'msc', and 'accomodation' are some of the most frequently appearing words in titles of 50 highest scoring posts.

• Most of the words that appear in the word cloud have neutral meanings and are associated with general information about LSE.

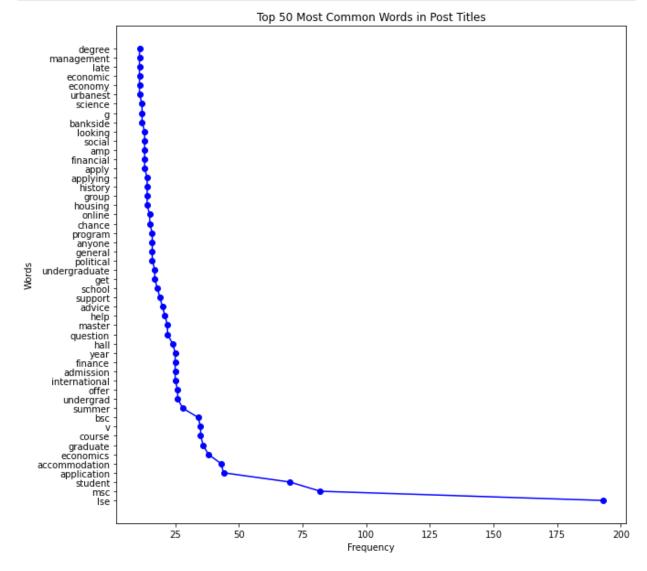
Secondly, we plot a line graph to visualize the top 50 words in all post titles.

```
In [37]: # Extracting keys and values from dictionary
    names = list(titles_d.keys())
    nums_t = list(titles_d.values())

# Creating a Line graph
    plt.figure(figsize=(10, 10))
    plt.plot(nums, names, marker='o', color='blue')

# Adding LabeLs and titLe
    plt.ylabel('Words')
    plt.xlabel('Frequency')
    plt.title('Top 50 Most Common Words in Post Titles')

# Displaying the graph
    plt.show()
```

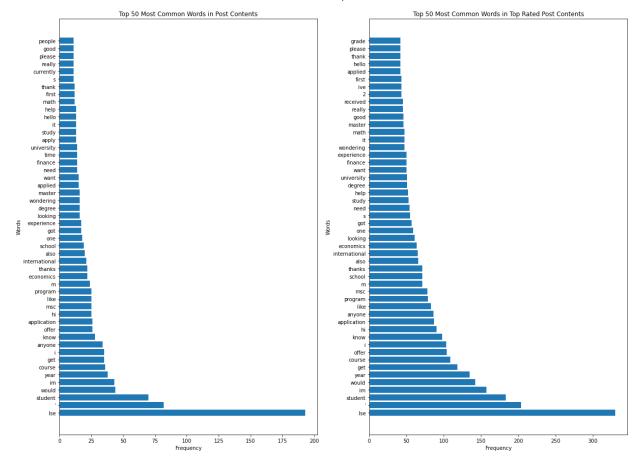


• By visualizing the frequency of the word appearance, we can see that the most frequently appearing word is 'lse', followed by 'msc', 'student', and 'application'.

- There is a large gap between the first and second most common word, and the appearance of words after the top ten common words does not vary a lot.
- Other than providing numerical count to each word's appearance, the result from this plot is actually similar to the previous plot.

Thirdly, we plot two bar charts to show the top 50 words in post contents and in top rated post contents respectively.

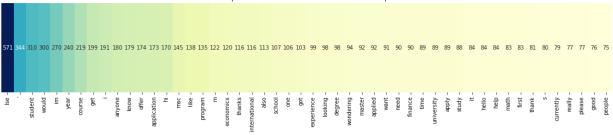
```
In [38]:
         # Extracting keys and values from dictionary
         words = list(contents d.keys())
         nums c = list(contents d.values())
         top words = list(top contents d.keys())
          top_nums = list(top_contents_d.values())
         fig, ax = plt.subplots(1,2, figsize = (20,15))
          # Create bar chart for the top 50 words in post contents.
          ax[0].barh(words, nums)
          ax[0].set xlabel('Frequency')
          ax[0].set ylabel('Words')
          ax[0].set title('Top 50 Most Common Words in Post Contents')
          # Create bar chart for the top 50 words in top rated post contents.
          ax[1].barh(top_words,top_nums)
          ax[1].set_xlabel('Frequency')
          ax[1].set ylabel('Words')
         ax[1].set_title('Top 50 Most Common Words in Top Rated Post Contents')
         Text(0.5, 1.0, 'Top 50 Most Common Words in Top Rated Post Contents')
Out[38]:
```



- Comparing the two graphs above, we can see that the most common words in all posts and top rated posts are very similar although the number of appearances differ.
- The most frequently appearing words are 'lse', 'student', 'im', 'would', and 'year', where the pronouns and auxiliary verbs presented here are not really helpful in analyzing public perception of LSE.
- Other words that are associated with LSE include 'economics', 'international', 'finance', 'good', and 'experience', which are some features that LSE is known for.

Finally, we plot a heat map to visualize the top 50 words in all post contents.

Top 50 Common Words in Post Contents Heatmap



• The heatmap here shows similar results to the previous graphs.

# Q1: What are some common traits that the public associate LSE with?

It can be shown that the common traits that the public associates LSE with are being international and strong in economics. This is because these two traits are represented by the most common words we extracted. There are also many keywords associated with MSc and graduates, which can be inferred that the postgraduate education of the LSE may be either famous or controversial.

### Q2: What do people on Reddit want to know more about LSE?

People may want to know more about the application process, the academic reputation and the accommodation of the LSE because there are many common keywords (offer, application, applied, accommodation, program, course, degree, personal, etc) related to these topics.

In [50]:	highest_scoring.head()			
Out[50]:		title	content	score
	607	got in 🥞	global medium culture msc applied nov 17 got h	24
	60	got in 🤩	global medium culture msc applied nov 17 got h	24
	370	whats job graduating	might help current prospective student underst	18
	570	lse v ucl v warwick ppe	hi all im fortunate position offer warwick ucl	16
	395	2024 undergrad — wait excruciating	completely understand circumstance excess appl	16

Moreover, if we look at the top scoring posts, we can see that the top rated post are mostly about people who received offers from LSE, indicating that people on Reddit are mostly interested about whether they can get accepted into the university.

# Conclusion

# **Findings**

In conclusion, our report illustrates that many comments on Reddit about the LSE are focused on its academic reputation, its application process and its accommodation. The common traits that the LSE shows to the public are its strength in diversity and economic research. There are also many keywords related to the postgraduate education of the LSE, which indicates that this topic has drawn many heated discussions as well.

### Limitations

While we were trying to understand public perception of LSE, we only gathered data from Reddit, which is a relatively small sample of the population to draw general conclusions from our findings.

Furthermore, most of the words we found have neutral meanings and failed to give us an insight of the public's like or dislike toward the university. The tokenized words also contains single words and pronouns that after the data cleaning process and are not insightful for our analysis.

Last but not least, we could only obtain fragmented idea about what the post creator is trying to convey by analyzing the words separately.

### **Future Improvements**

In the future, we could use additional features of the NLTK module that are not currently included to remove auxiliary verbs and pronouns. We could also make use of other Python libraries like PushShift or PRAW to analyse and collect more than 1000 posts. If possible, we would also like to collect different data sources to extend our study to a larger population.

To address the issue of collecting neutral words that are less insightful, we could also look at whether there are more positive or more negative words and analyse the top scored posts in regards to this.

Testing and debugging could also be a potential future improvement. We could invest more time in creating thorough tests and debugging tools to make it easier to catch and debug errors to make sure the code runs smoothly, reliably, and achieve the intended goal.

In [ ]: