# Meeting Minutes - 22 Feb 2022



## Meeting Minutes for 22 February 2022

Date:	22 Feb 2022
Time:	5:00PM
Location:	MS Teams
Present:	<ul> <li>@ Ow Ling Jia</li> <li>@ Tian Le Cheow</li> <li>@ Chen Jian Yu</li> <li>@ Joshua Wong</li> <li>@ Sarah Chin</li> </ul>
Absent:	@ ONG JHIN YEE _

Goals

Updates for Prof Jisun on what's happening with the project.

## 🖺 Agenda

No.	Agenda Item	Remarks
	Updates on setting up scrapers on AWS	@ Chen Jian Yu
	a. Testing of the scrapers on the server	
	b. Confirmed methods for scraping	
2.	Machine Learning Methodology	@ Ow Ling Jia @ Joshua Wong
	a. Trend Analysis	
	b. Keyword Analysis	
	c. Sentiment Analysis	
3.	Frontend Progress	@ Tian Le Cheow
	a. Showcase current progress for the webpage	

**P**E Discussion topics

Agenda	Action By
Updates on setting up scrapers on the server	

<ul> <li>Scraper - Backend <ul> <li>created a few servers (3 in total)</li> </ul> </li> <li>1st one: Reddit and twitter data - Reddit can only get the most recent 7 days <ul> <li>Reddit Historical: Working but most recent 3 months cannot get comments; no error, just blank</li> <li>Twitter Historical: Cannot get anything past 7 days <ul> <li>tried using snscrape; plans to try again some time this well</li> </ul> </li> <li>Instagram Historical: the data cannot be scraped and data cannot be retrieved, might not want to use IG anymore</li> <li>Facebook: For now we will use Facepager, since it is the only one that can manually scrape data</li> <li>YouTube: We are using Selenium, might have to try the code again (daily) <ul> <li>Historical: might have to give up as we run into the same issue</li> </ul> </li> <li>Prof Jisun: Twitter might be okay to forgo since Singaporeans don't use this social media a lot</li> <li>Instagram: If not possible, don't need to pursue</li> <li>First, use historical data to visualise how it will appear on the dashboard</li> <li>YouTube: Selenium method is slow so might not be scalable <ul> <li>Prof suggested using this YouTube API: https://developers.google.com/youtube/v3</li> </ul> </li> <li>Action 1: Data collection process to be sent to Jisun</li> </ul> </li> <li>Issues with Facepager on different social media sites: YouTube have to use the playlist to get the videos but not every channel have multiple playlists</li> <li>On top of that, not every video is included in the playlists</li> </ul> <li>AWS Server: Can close the server and transfer all on the local server</li> <li>Will the local server crash, seeing that the AWS one crashed?</li> <li>Prof: local server should be powerful enough to handle it; loop back to the server either this week or next week since most people wont be running it during this period</li> <li>Action 2: Transfer all the data to the local server</li>	Jian Yu
Reddit, Twitter and Youtube daily scrapers are more or less fixed and there are no pressing issues	
<ul> <li>Action 3: Let Prof Jisun know about the size of the historical data</li> </ul>	
Cost of the server: roughly US\$12	
No updates from Amazon's side	
Machine Learning Methodology	
Graphs	Action 1, 2 & 3: Ling Jia
<ul> <li>Graphs</li> <li>The team tested the trend analysis module to see how it would look like over time</li> <li>Engagement metrics will change according to the social media platform <ul> <li>e.g. Twitter - number of tweets, Facebook - number of posts</li> </ul> </li> </ul>	Action 1, 2 & 3: Ling Jia
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- But the results aren't very good; doesn't seem to get the results that we want
- The same word can also reflect different entity
- '#' represents tokenizing; LinkedIn might not be in the library so it breaks it down into different attributes to be analysed
  - Action 2: Find out what's happening with the code
  - · Prof Jisun: Library might have different ways of storing the word
    - · Look at aggregated level and compare the methods and choose which one would be the best
    - Some models are used for research instead of real world applications and therefore may not be the best model
    - · Split data by month or year and try the different models to see which one works best
    - Action 3: Try the proposed method and update Prof Jisun

#### Sentiment Analysis

Polarity and emotions were looked into

Emotion classifier - captures 7 different emotions (Hugging Face)

- There's no neutral emotion, so the team manually inserted a threshold to get a neutral emotion (7.5)
  - Calculated the average number of Facebook comments per post and the length of the comments • Limitations to long and short comments: Unable to tell whether the entirety of the comment is
- relevant to the topic at handTested the accuracy of the code
  - Results were not good
  - Data was trained on the twitter dataset but doubt that it is the training data's fault
  - Could be the model's fault
  - Prof Jisun: more first person based data and might be affecting the model
    - Hugging face zero shot classification: pretrained data where you give the label and the model will just find the best label; in classroom setting it works well so can try it out
    - Have backup plan: choose the ones that works the best at this moment
    - Problem with the hugging face: Works relatively slow
    - Prof Jisun shared some codes: https://github.com/anjisun221/css\_codes/blob/main/ay21t1 /Lab05\_text\_classification/Lab05\_text\_classification%20-%20Students.ipynb
  - Problem to address: the more labels there are the worse it works
  - Positive and negative might work better as compared to searching for emotion
  - Prof Jisun: Reasonable accuracy should be 70%
    - For emotions the accuracy might not be as high

Looking into Singlish words

- A lot of the data is using Singlish but a lot of models are not trained to classify Singlish words
- Still need to label the data even if the tokenizer is working

Jian Yu: Core features better or more features?

- Prof Jisun: focus on core first, but from professors' side, project might look a little too barren and basic
   Don't have to focus on everything on secondary features but can include some that makes us stand out
  - Create at least one special feature that makes us stand out to justify to the stakeholders
- Prof Jisun to update us abt linking the GPU server with the local server

Jian Yu: The app itself or the machine learning models?

- If the app can work on historical data it should be fine as well
- Even if we show until last November on the app itself it should be okay
- Visualisation would be very important to tie everything together

**Frontend Progress** 

Build using Vue2 instead of Vue3 because some of the applications still do not work, but it can still be migrated over in the future	Tian Le
Search Feature	
<ul> <li>In the search box, entering a query would then have a dropdown showing the related topics</li> <li>Date period can also be selected or customised</li> <li>Sentiment and platform filter has also been included</li> </ul>	
Tooltip - shows the user how the component can be used	
<ul> <li>Intend of showing the sentiment related to the topics in the Trending Topics segment</li> </ul>	
Trending Analysis	
The graph sown will be affected by the main filters that the user can customise	
Top Keywords	
<ul> <li>Intend on inserting more data</li> <li>Will insert the legend also</li> <li>Action 1: Work on the attributes above</li> </ul>	
Graphs are using 2 different packages - since some of the features are only supported in Vue3, therefore need to use 2 different packages	

#### :note: Meeting Notes - Updates

Update	Risk Level	Mitigation Plan
School server has been restarted and can be used		Transfer all the data to the local server

**J** Decisions

Administrative Matters

Date of Next Meeting:	24 Feb 2022
Time of Next Meeting:	😂 5РМ