## STAT GR5243/GU4243 Fall 2023 Project 1: What made you happy today

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This report explores on what makes people happy everyday. The whole dataset: HappyDB is a corpus of 100,000 crowd-sourced happy moments. The external link of this dataset is <a href="https://megagon.ai/happydb-a-happiness-database-of-100000-happy-moments">here (https://megagon.ai/happydb-a-happiness-database-of-100000-happy-moments</a>); it also contains a detailed paper explaining relevant research. This dataset collect happy moments that happened either within a 24-hour or a 3-month period for each participant.

As for the happy-moment data, I will use the "cleaned\_hm.csv" dataset since it cleans misspelled words and incomplete sentences. This dataset serves as my main dataset.

#### Stage 1: Initial Exploration

1.1 First I will perform some initial exploration to understand this dataset. I will read in the "cleaned\_hm.csv" dataset and apply a "data\_info\_present" function to get an overview of the dataset. Throughout this project, I will consistently call functions from my "lib" folder for the seek of clean coding.

The dataset looks clean and the "happy moment" descriptions are all quite complete. There is about half of the happy moments from 24h and the other half is recalled within 3 months, indicating that this is a balanced dataset.

There are some other datasets from the "Happy DB" author, yet I think the only useful one is the "demographics.csv". However, this analysis and data story are not related to the background information of the participants, so I will solely focus on the "cleaned\_hm.csv".

In [2]: # read in dataset
df = pd.read\_csv("../data/cleaned\_hm.csv")
df.head(10)

Out[2]:		hmid wid reflection_period		reflection_period	original_hm cleaned_hm		modified	num_sentence	ground_truth_category	predicted_category
	0	27673	2053	24h	I went on a successful date with someone I fel	I went on a successful date with someone I fel	True	1	NaN	affection
	1	27674	2	24h	I was happy when my son got 90% marks in his e	I was happy when my son got 90% marks in his e	True	1	NaN	affection
	2	27675	1936	24h	I went to the gym this morning and did yoga.	I went to the gym this morning and did yoga.	True	1	NaN	exercise
	3	27676	206	24h	We had a serious talk with some friends of our	We had a serious talk with some friends of our	True	2	bonding	bonding
	4	27677	6227	24h	I went with grandchildren to butterfly display	I went with grandchildren to butterfly display	True	1	NaN	affection
	5	27678	45	24h	I meditated last night.	I meditated last night.	True	1	leisure	leisure
	6	27679	195	24h	I made a new recipe for peasant bread, and it	I made a new recipe for peasant bread, and it	True	1	NaN	achievement
	7	27680	740	24h	I got gift from my elder brother which was rea	I got gift from my elder brother which was rea	True	1	NaN	affection
	8	27681	3	24h	YESTERDAY MY MOMS BIRTHDAY SO I ENJOYED	YESTERDAY MY MOMS BIRTHDAY SO I ENJOYED	True	1	NaN	enjoy_the_moment
	9	27682	4833	24h	Watching cupcake wars with my three teen children	Watching cupcake wars with my three teen children	True	1	NaN	affection

# In [4]: # This function is from ../lib/data\_exploration.py. It should serve as a good overview of the dataset data\_info\_present(df)

There are 100535 rows and 9 features Check what features are there and the data type hmid int64 int64 reflection\_period object original hm object cleaned\_hm object modified bool num sentence int64 ground\_truth\_category object predicted\_category object dtype: object Check how many uniques value for each column hmid: 100535 unique values wid: 10841 unique values reflection\_period: 2 unique values original\_hm: 96557 unique values cleaned\_hm: 96486 unique values modified: 2 unique values num\_sentence: 47 unique values
ground\_truth\_category: 7 unique values predicted\_category: 7 unique values Check for missing values: wid reflection\_period 0 original\_hm cleaned\_hm 0 modified 0 num sentence ground\_truth\_category predicted\_category 0 dtype: int64

1.2 After exploring the overall structure and statistics of this dataset, I will further reduce features by the "feature reduction" function in lib.

- Drop the "original\_hm". I assume a cleaned version is better (or at least not worse).
- Drop the "modified" column since the "original\_hm" column is dropped
- Remove "ground\_truth\_category". There is too much missing value; plus, "happiness" is an abstract concept, it is not realistic to have a correct and defined category for each situation, so "predicted\_category" is sufficient for analysis. In this analysis, the "predicted\_category" is more or less equivalent to ground truth.

## In [3]: df\_processed = feature\_reduction(df) df\_processed.head()

Out[3]:

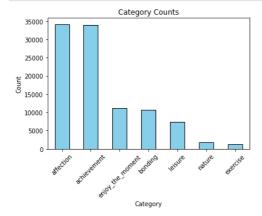
Check if there is any duplicated row Number of duplicate rows: 0

	hmid	wid	reflection_period	cleaned_hm	num_sentence	predicted_category
0	27673	2053	24h	I went on a successful date with someone I fel	1	affection
1	27674	2	24h	I was happy when my son got 90% marks in his e	1	affection
2	27675	1936	24h	I went to the gym this morning and did yoga.	1	exercise
3	27676	206	24h	We had a serious talk with some friends of our	2	bonding
4	27677	6227	24h	I went with grandchildren to butterfly display	1	affection

## 1.3 Visualize the count of each category of the happy moments using the "bar\_chart\_visualization" function in lib

I am mostly interested in what type of happiness is included in this dataset. Affection and achievement are the top 2 category counts, which are also significantly greater than the rest. This is reasonable since most of the time, people are happy about seeing things they like or improving themselves. These 2 scenarios are associated with long-term happiness, and people are likely to maintain a better impression on "long-term" positive changes, while other categories like "enjoy\_the\_moment", "leisure", or "nature" are likely to make you happy for a shorter time; you probably won't remember it for long after leaving the scenarios.

### In [6]: bar\_chart\_visualization(df\_processed, "predicted\_category")



#### Stage 2: Happy Moment in 24h VS Happy Moment in 3m

Happy moments shared by the participants could be divided into 2 categories according to the "reflection\_period", either recalled from within 24h or 3 months. Are the happy words different between recent and old happy moments?

There might be some differences. A hypothesis can be if people regard something happened about 3 month ago as their response, it is more likely to be a "happier" moment than something happened within 24h. However, this is not necessarily correct.

I will use my "word\_cloud" method in lib, which draws word cloud after filtering out some distracting words, including the default stopwords used in the wordcloud package and extra non-informative words like "happy", "day", "went", "got", "able".

In fact, from just the 2 word clouds, it is hard to tell if there is any source difference of happiness since both include overlapped words like "day", "friend", "work", "family", "new", etc.

My conclusion would be the happy words is indifferent from the time perspective. However, the time difference is truly reflected in the wordcloud, the 24h wordcloud has more frequent time-related words like "morning", "yesterday", "last night".

```
In [7]: df_3m = df_processed[df_processed["reflection_period"]=="3m"]
df_24h = df_processed[df_processed["reflection_period"]=="24h"]
```

In [8]: word\_cloud(df\_3m)



In [9]: word\_cloud(df\_24h)



### Stage 3: Sentiment Analysis, which type of positive emotion? what are the happy words?

I want to follow the sentiment analysis that is similar to what Professor shared. However, since this is a happy moment dataset, I would assume most participant response will be positive and probably be classified as joy. Thus, instead of performing a thorough sentiment analysis, I will first mainly focus on classifying into 4 emotions: positive, joy. trust. anticipation.

The main technique is my "sentiment\_analysis" file in the lib. I will tokenize each response sentence through the NLTK package and apply the NRCLex python library to extract the emotion scores for my target 4 emotions.

Then my question is what are the "happy words" that lead a sentence to be a happy sentence. Each word in the tokenized list would be explored, and only those lead to the 4 target emotions will be recorded.

To speed up the process, I stored this new generated csv file, which contains a few new columns, into the output folder. Heatmap is applied to view the correlation between the emotion scores. Joy scores are strongly related to positive and trust scores with 0.17 and 0.15, while anticipation might not be a "strict" happiness emotion.

Building upon on this detailed analysis on the emotions happened behind happy moments, I am also interested in what are the exact words that contribute to happiness. This is different from the word cloud which just collects high frequent words; these new happy words are extracted through separate score extraction in the above process, which eventually becomes the "happy\_words" column in the output csv file.

The too 10 happy words are ['happy', 'time', 'good', 'friend', 'found', 'dinner', 'lond', 'daughter', 'iob', 'finally'], which sort of makes sense. "Time" is likely to occur in a

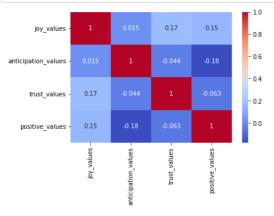
```
In [2]: # Load the intermediate csv output if it is already stored
    if not os.path.exists("../output/hm_word_analzyed.csv"):
        df_word_analzyed = process_text(df_processed)
        df_word_analzyed.to_csv("../output/hm_word_analzyed.csv", index=False)
    else:
        df_word_analzyed = pd.read_csv("../output/hm_word_analzyed.csv")
```

#### In [3]: df\_word\_analyzed

Out[3]:		hmid	wid	reflection_period	cleaned_hm	num_sentence	predicted_category	happy_words	joy_values	positive_values	trust_values	anticipation_values	en
	0	27673	2053	24h	I went on a successful date with someone I fel	1	affection	['successful', 'sympathy']	0.000000	0.333333	0.000000	0.000000	positive_
	1	27674	2	24h	I was happy when my son got 90% marks in his e	1	affection	['happy']	0.142857	0.142857	0.142857	0.142857	joy_
	2	27675	1936	24h	I went to the gym this morning and did yoga.	1	exercise	0	0.000000	0.000000	0.000000	0.000000	joy_
	3	27676	206	24h	We had a serious talk with some friends of our	2	bonding	['talk', 'good']	0.000000	0.181818	0.000000	0.000000	positive_
	4	27677	6227	24h	I went with grandchildren to butterfly display	1	affection	['grandchildren']	0.250000	0.250000	0.250000	0.250000	joy_
	100530	128762	112	24h	My husband announced he is getting a decent bo	1	affection	['decent']	0.000000	0.400000	0.000000	0.000000	positive_
	100531	128763	714	24h	Had a can of Pepsi to drink.	1	enjoy_the_moment	0	0.000000	0.000000	0.000000	0.000000	joy_
	100532	128764	3934	24h	Cuddling with my girlfriend last night.	1	affection	0	0.000000	0.000000	0.000000	0.000000	joy_
	100533	128765	1629	24h	I had a great meeting yesterday at work with m	1	bonding	['excited', 'working', 'efficient', 'team']	0.000000	0.300000	0.300000	0.000000	positive_
	100534	128766	141	24h	I had a great workout last night.	1	exercise	0	0.000000	0.000000	0.000000	0.000000	joy_

#### 100535 rows × 12 columns

#### In [4]: heatmap(df word analyzed)



```
In [4]: top_10_happy_words = extract_keywords(df_word_analyzed)
print(top_10_happy_words)
```

```
['happy', 'time', 'good', 'friend', 'found', 'dinner', 'long', 'daughter', 'job', 'finally']
```

```
In [3]: %%capture
! pip install nbconvert
```

```
In [5]: | jupyter nbconvert --to html Clarence_Jiang_yj2737_proj1_blog.ipynb
```

[NbConvertApp] Converting notebook Clarence\_Jiang\_yj2737\_proj1\_blog.ipynb to html [NbConvertApp] Writing 1274648 bytes to Clarence\_Jiang\_yj2737\_proj1\_blog.html

In [ ]: