

Codebook: Twitter Depression Dataset

```
In [63]:
          #importing libraries
          import numpy as np
          import matplotlib as plt
          import pandas as pd
          import seaborn as sns
In [53]:
         #importing the data
          url = 'https://raw.githubusercontent.com/zeineb-ouerghi/DS4SG-Project-/main/Data/Mental-He
          raw = pd.read csv(url)
          # drop the original column for index since df has index now
          raw.drop("Unnamed: 0", axis=1, inplace=True)
          # drop post id column since we won't need it for our analysis
          raw.drop('post id', axis=1, inplace=True)
          # rearrange variable order
          post created = raw['post created']
          raw = raw.drop(columns=['post created'])
          raw.insert(loc=1, column='post_created', value=post_created)
          raw.head()
```

Out[53]:		post_text	post_created	user_id	followers	friends	favourites	statuses	retweets	label	
	0	It's just over 2 years since I was diagnosed w	Sun Aug 30 07:48:37 +0000 2015	1013187241	84	211	251	837	0	1	
	1	It's Sunday, I need a break, so I'm planning t	Sun Aug 30 07:31:33 +0000 2015	1013187241	84	211	251	837	1	1	
	2	Awake but tired. I need to sleep but my brain	Sat Aug 29 22:11:07 +0000 2015	1013187241	84	211	251	837	0	1	
	3	RT @SewHQ: #Retro bears make perfect gifts and	Sat Aug 29 18:40:49 +0000 2015	1013187241	84	211	251	837	2	1	
	4	It's hard to say whether packing lists are mak	Sat Aug 29 18:40:26 +0000 2015	1013187241	84	211	251	837	1	1	

Variable information:

```
In [54]: # checking the number of rows and variables
  raw.shape

Out[54]: (20000, 9)
```

post_text : the post contents, stored as a string, gathered from the original tweets.

post_created: the date and time when the tweet was posted, stored as a string (we will convert the entire column to be in datetime format), taken from the tweets.

user_id: the ID of the user, stored as an integer, taken from the tweet information.

followers: the number of followers of the user, stored as an integer, taken from their profile.

friends: the number of the friends of the user, also stored as an integer, taken from their profiles.

favourites: the number of "likes" the post has, stored as an integer, taken from the tweet itself.

statuses: the number of total tweets they have, stored as an integer, taken from the users' profiles.

retweets: the number of retweets the post had, stored as an integer, also taken from the tweet data itself.

label: whether the user has depression or not, stored as '0' or '1', where 0 is not depressed and 1 is depressed.

The number of null values will be explored in the next code cell. We have no null values in this dataset.

```
In [55]:
```

```
# checking the proportion of null-values in each column
percent_null = [(i, raw[i].isnull().mean()) for i in raw]
percents = pd.DataFrame(percent_null, columns=("Column", "% null values"))
display(percents)
```

	Column	% null values
0	post_text	0.0
1	post_created	0.0
2	user_id	0.0
3	followers	0.0
4	friends	0.0
5	favourites	0.0
6	statuses	0.0
7	retweets	0.0
8	label	0.0

```
In [56]:
```

```
# data information
raw.info()
```

```
RangeIndex: 20000 entries, 0 to 19999
Data columns (total 9 columns):
 # Column Non-Null Count Dtype
0 post text 20000 non-null object
 1 post created 20000 non-null object
 2 user id 20000 non-null int64
 3 followers
               20000 non-null int64
               20000 non-null int64
 4 friends
 5 favourites 20000 non-null int64
 6 statuses 20000 non-null int64
               20000 non-null int64
 7
  retweets
    label
                20000 non-null int64
dtypes: int64(7), object(2)
memory usage: 1.4+ MB
```

<class 'pandas.core.frame.DataFrame'>

All variables except post_created and post_text are in integer format. In the following code cell, we will convert the post_created variable to datatime format and only keep the date (excluding timestamp).

```
In [57]: raw['post_created'] = pd.to_datetime(raw['post_created']).dt.strftime('%m/%d/%Y')
```

In [59]: #Description of the data
raw.describe().T

Out[59]:		count	mean	std	min	25%	50%	75%	
	user_id	20000.0	3.548623e+16	1.606083e+17	14724376.0	324294391.0	1.052122e+09	2.285923e+09	7
	followers	20000.0	9.004840e+02	1.899914e+03	0.0	177.0	4.760000e+02	1.197000e+03	2.
	friends	20000.0	7.824288e+02	1.834818e+03	0.0	211.0	5.610000e+02	7.010000e+02	2.
	favourites	20000.0	6.398236e+03	8.393073e+03	0.0	243.0	2.752000e+03	8.229000e+03	3.9
	statuses	20000.0	4.439442e+04	1.407785e+05	3.0	5129.0	1.325100e+04	5.289200e+04	1.
	retweets	20000.0	1.437927e+03	1.511967e+04	0.0	0.0	0.000000e+00	1.000000e+00	8.
	label	20000.0	5.000000e-01	5.000125e-01	0.0	0.0	5.000000e-01	1.000000e+00	1.0

In [61]: # correlation
 raw.corr()

Out[61]:		user_id	followers	friends	favourites	statuses	retweets	label
	user_id	1.000000	0.098022	-0.073787	-0.050215	-0.050672	-0.020921	-0.220954
	followers	0.098022	1.000000	0.891919	-0.016242	-0.000215	0.001500	-0.011181
	friends	-0.073787	0.891919	1.000000	-0.081756	-0.045420	0.009946	0.119818
	favourites	-0.050215	-0.016242	-0.081756	1.000000	-0.053930	0.069551	-0.252289
	statuses	-0.050672	-0.000215	-0.045420	-0.053930	1.000000	-0.005193	-0.246473
	retweets	-0.020921	0.001500	0.009946	0.069551	-0.005193	1.000000	-0.025198
	label	-0.220954	-0.011181	0.119818	-0.252289	-0.246473	-0.025198	1.000000

The classes in our dataset are balanced

In [64]: sns.countplot(data=raw,x="label")

Out[64]: <AxesSubplot:xlabel='label', ylabel='count'>

