Text Mining_Q1_Elon Musk

1) Perform sentimental analysis on the Elon-musk tweets (Elon-musk.csv)

1. Import Libs

```
In [14]:
```

```
pip install -U spacy ...
```

In [2]:

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import string
import spacy
from wordcloud import WordCloud
import re
import warnings
warnings.filterwarnings('ignore')

from nltk.stem import WordNetLemmatizer
from nltk.corpus import stopwords
from nltk.tokenize import word_tokenize
from nltk.tokenize import sent_tokenize
from sklearn.feature_extraction.text import TfidfVectorizer
```

2. Import Data

In [3]:

```
tweets=pd.read_csv('Elon_musk.csv',encoding='Latin-1')
tweets
```

Out[3]:

	Unnamed: 0	Text
0	1	@kunalb11 I m an alien
1	2	@ID_AA_Carmack Ray tracing on Cyberpunk with H
2	3	@joerogan @Spotify Great interview!
3	4	@gtera27 Doge is underestimated
4	5	@teslacn Congratulations Tesla China for amazi
1994	1995	@flcnhvy True, it sounds so surreal, but the n
1995	1996	@PPathole Make sure to read ur terms & Dr
1996	1997	@TeslaGong @PPathole Samwise Gamgee
1997	1998	@PPathole Altho Dumb and Dumber is <u+0001f525< th=""></u+0001f525<>
1998	1999	Progress update August 28

1999 rows × 2 columns

In [4]:

```
tweets.drop(['Unnamed: 0'],inplace=True,axis=1)
tweets
```

Out[4]:

	Text					
0	@kunalb11 I m an alien					
1	@ID_AA_Carmack Ray tracing on Cyberpunk with H					
2	@joerogan @Spotify Great interview!					
3	@gtera27 Doge is underestimated					
4	@teslacn Congratulations Tesla China for amazi					
1994	@flcnhvy True, it sounds so surreal, but the n					
1995	@PPathole Make sure to read ur terms & amp; con					
1996	@TeslaGong @PPathole Samwise Gamgee					
1997	@PPathole Altho Dumb and Dumber is <u+0001f525< th=""></u+0001f525<>					
1998	Progress update August 28					

1999 rows × 1 columns

3. Data Preprocessing

remove both the leading and the trailing characters

```
In [5]:
```

```
tweets=[Text.strip() for Text in tweets.Text]
tweets=[Text for Text in tweets if Text] # removes empty strings.
tweets[0:10]
```

Out[5]:

```
['@kunalb11 I\x92m an alien',
    '@ID_AA_Carmack Ray tracing on Cyberpunk with HDR is next-level. Have you tried it?',
    '@joerogan @Spotify Great interview!',
    '@gtera27 Doge is underestimated',
    '@teslacn Congratulations Tesla China for amazing execution last year. Now on to the next for even more!!',
    'Happy New Year of the Ox! https://t.co/9WFKMYu2oj', (https://t.co/9WFKMYu2oj',)
    'Frodo was the underdoge,\nAll thought he would fail,\nHimself most of all.
https://t.co/zGxJFDzzrM', (https://t.co/zGxJFDzzrM',)
    '@OwenSparks_ @flcnhvy @anonyx10 Haha thanks:)',
    '@flcnhvy @anonyx10 Indeed! Tweets definitely do not represent real-world time allocation.',
    'The most entertaining outcome is the most likely']
```

Joining the list into one string/text

In [6]:

```
tweets_text=' '.join(tweets)
tweets_text
```

RationalEtienne @Adamklotz_ Once we can predict cash flow reasonably well, Starlink will IPO @RationalEtienne @Adamklotz_ Starlink is a staggeringly difficult technical & amp; economic endeavor. However, if we don\x85 http s://t.co/9Z8Ac6skqx (https://t.co/9Z8Ac6skqx) @RationalEtienne @Adamklotz SpaceX needs to pass through a deep chasm of negative cash flow over the n ext year or\x85 https://t.co/7J1c92hdjl (https://t.co/7J1c92hdjl) @ID AA C armack Lowest cost per ton of carbon sequestered, net of value of any prod uct made. \n\nMust be scalable to g\x85 https://t.co/XMyI7qWSgw (https:// t.co/XMyI7qWSgw) @Adamklotz_ It\x92s meant to be the same price in all cou ntries. Only difference should be taxes & amp; shipping. @tobyliiiiiiiii T his is intended for Earth, but there may be some ideas that apply to Mars too @CryptoShrikar @CoinDesk @Tesla @Dan_Z_Palmer <U+0001F923><U+0001F923 > XPrize team will manage the \$100M carbon capture prize https://t.co/fSw5 IanLOr (https://t.co/fSw5IanLOr) @SRuhle Everyone at Tesla receives stock. \n\nMy comp is all stock/options, which I do not take off the table. \n\nT hat\x92s what you\x92re missing. Back to work I go @CapybaraSurfer @MattWa llace888 Does seem a bit high @itsALLrisky Doge appears to be inflationar y, but is not meaningfully so (fixed # of coins per unit time), whereas\x8 5 https://t.co/4UH4rARC6T (https://t.co/4UH4rARC6T) @MichaelGalanin @kevin

In [7]:

```
tweets_text = re.sub("[^A-Za-z" "]+"," ",tweets_text).lower() #remove special character
tweets_text = re.sub("[0-9" "]+"," ",tweets_text).lower() #remove numbers
tweets_text = re.sub(r'^https?:\/\/.*[\r\n]*', '', tweets_text).lower() #remove hyperlink
tweets_text
```

tarlink satellites u https t co bjfjlczwdk https t co ln reesbw off twitte r for a while it will be filled with graffiti art eiraum u c u b u u u b u u u c giga berlin progress https t co ekpg qcbui hamoon neuralink neural ink is working super hard to ensure implant safety amp is in close communi cation with the https t co yky llpumd rt spacex this mission enables acces s for everyday people who dream of going to space rt spacex announcing the first commercial astronaut mission to orbit earth aboard dragon u https t co mbesvnakad https t co uklsjffrjk if you ve worked on advanced wearables phones or robots those skills are needed neuralink feels weird helping mak e hopefully good version of cyberpunk come true please consider working at neuralink short term solve brain spine injuries long term human ai symbios is latte https t co qrc alads https t co xccokojylt on clubhouse tonight a t pm la time erdayastronaut tj cooney lrocket spacex felixschlang marcusho usegame tom has some great stories about experim https t co teezkiwcmk rt commercial crew launch alert u f nasa and spacex are targeting no earlier than april for the launch of the second crew rotation mis erdayastronaut t j cooney lrocket spacex felixschlang marcushousegame he became a big fan o f methane but was https t co mdv vdefyc tj cooney lrocket spacex felixschl ang marcushousegame that sounds correct tom certainly deserves a lot of cr e https t co hdri hfa b tj cooney lrocket spacex felixschlang marcushouseg

In [8]:

```
wordnet = WordNetLemmatizer()
```

In [9]:

```
text_tokens = word_tokenize(tweets_text)
```

In [10]:

```
tokens_without_sw = [word for word in text_tokens if not word in stopwords.words()]
```

In [11]:

```
tf = TfidfVectorizer()
```

In [12]:

```
text_tf = tf.fit_transform(tokens_without_sw)
```

Create the DataFrame

In [13]:

```
feature_names = tf.get_feature_names()
dense = text_tf.todense()
denselist = dense.tolist()
df =pd.DataFrame(denselist, columns=feature_names)
```

In [14]:

df

Out[14]:

	aa	аауурсрр	abo	aboard	abort	absence	absolute	absolutely	absorb	absorption
0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
1	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
2	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
3	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
4	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
14512	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
14513	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
14514	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
14515	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
14516	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0

14517 rows × 5099 columns

4

In [15]:

```
word_list = ' '.join(df)
```

In [47]:

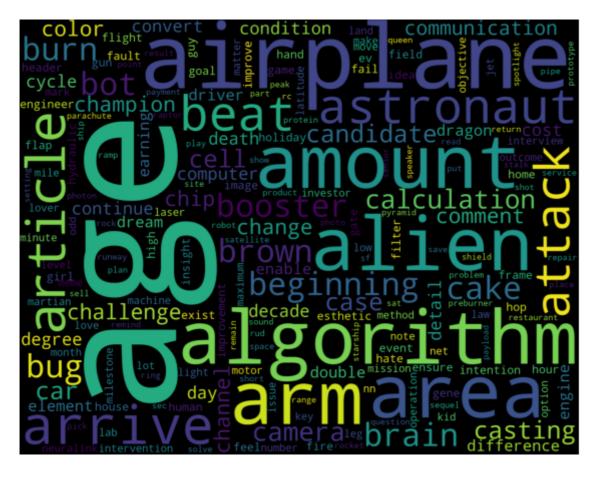
wordcloud = WordCloud(background_color='black',width=1800,height=1400).generate(word_list)

In [49]:

```
plt.figure(figsize=(10,10))
plt.imshow(wordcloud)
plt.axis('off')
```

Out[49]:

(-0.5, 1799.5, 1399.5, -0.5)



4. Sentimental Analysis

```
In [43]:
```

```
with open('positive-words.txt') as pw:
    positive_words = pw.read().split("\n")
```

In [44]:

```
positive_words = positive_words[35:]
positive words
 'adroit',
 'adroitly',
 'adulate',
 'adulation',
 'adulatory',
 'advanced',
 'advantage',
 'advantageous',
 'advantageously',
 'advantages',
 'adventuresome',
 'adventurous',
 'advocate',
 'advocated',
 'advocates',
 'affability',
 'affable',
 'affably',
 'affectation',
 'affection'
In [45]:
```

```
with open('negative-words.txt') as nw:
    negative_words = nw.read().split("\n")
```

In [46]:

```
negative_words = negative_words[35:]
negative_words
 'ail',
 'ailing',
 'ailment',
 'aimless',
 'alarm',
 'alarmed',
 'alarming',
 'alarmingly',
 'alienate',
 'alienated'
 'alienation',
 'allegation',
 'allegations',
 'allege',
 'allergic',
 'allergies',
 'allergy',
 'aloof',
 'altercation',
 'ambiquity'
In [22]:
txt_pos_in_pw = ' '.join([word for word in df if word in positive_words])
```

In [23]:

wordcloud_pos = WordCloud(background_color='black',width=1800,height=1400).generate(txt_pos

In [57]:

```
plt.figure(figsize=(7,15))
plt.imshow(wordcloud_pos)
plt.axis('off')
```

Out[57]:

(-0.5, 1799.5, 1399.5, -0.5)



In [25]:

```
txt_neg_in_nw = ' '.join([word for word in df if word in negative_words])
```

In [26]:

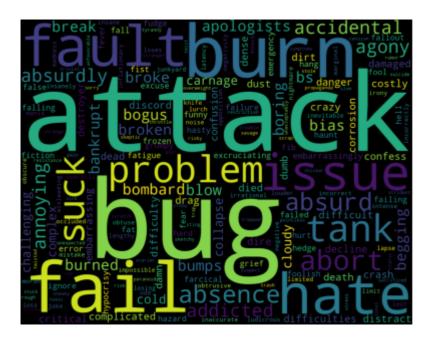
wordcloud_neg = WordCloud(background_color='black',width=1800,height=1400).generate(txt_neg

In [58]:

```
plt.figure(figsize=(7,15))
plt.imshow(wordcloud_neg)
plt.axis('off')
```

Out[58]:

(-0.5, 1799.5, 1399.5, -0.5)



END

In []: