# Text analysis on Data from Twitter using API Calling

Healthcare in US

## **Project Definition**

### 1. Target

3 stocks in sector of Healthcare in US

#### 2. Period

2 week of December (from 02 Dec 2019 to 16 Dec 2019)

#### 3. Data collection

Using API calling to collect data from Twitter

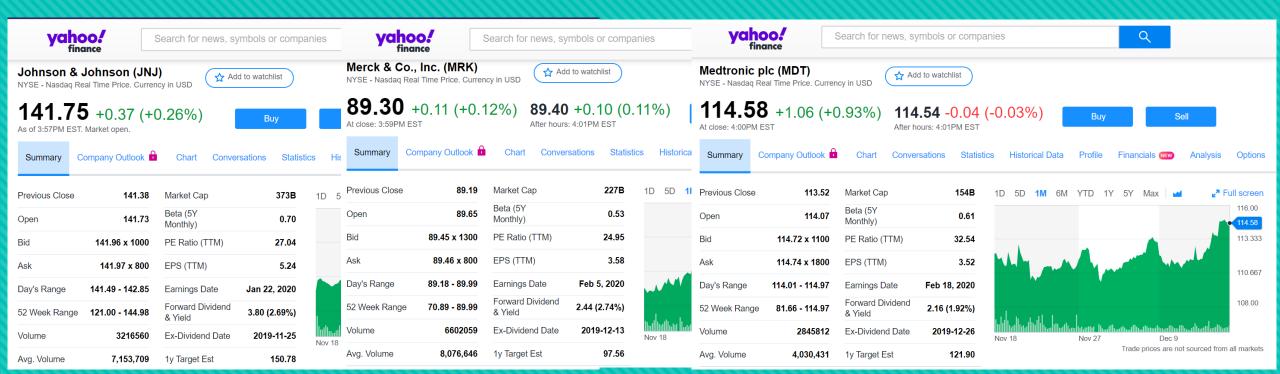
## 4. Text analysis

most common words wordcloud sentiment detection

#### Johnson Johnson





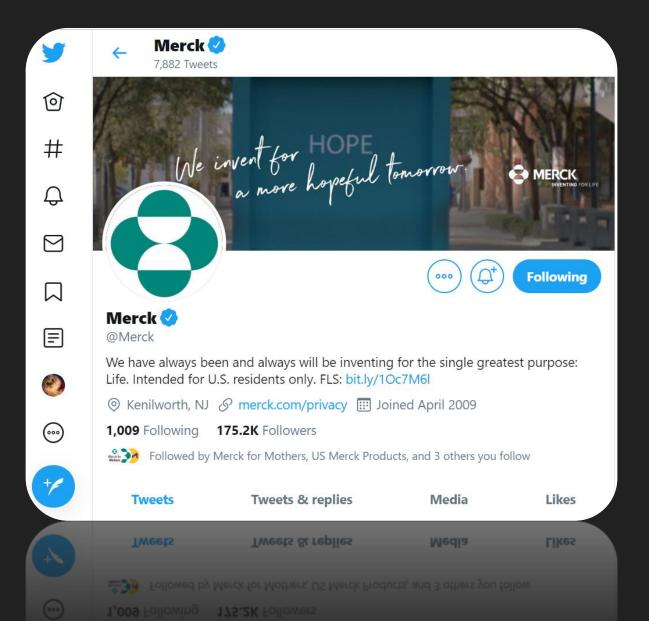


# Target

## JNJ twitter profile

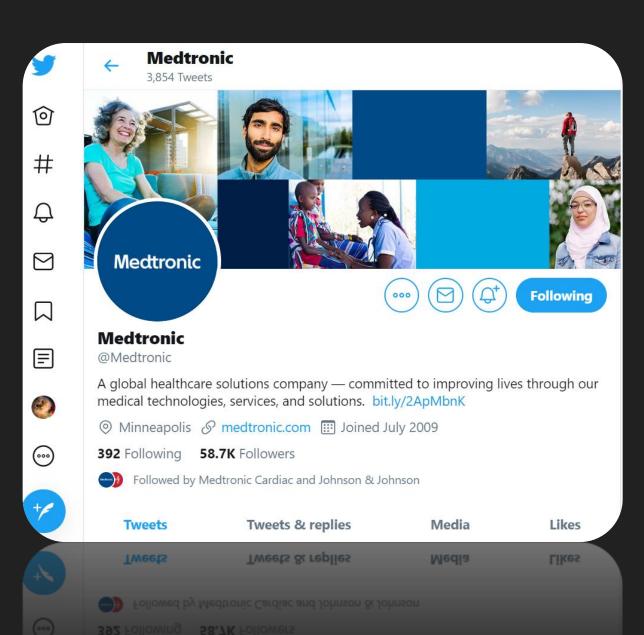


## MRK twitter profile





## MDT twitter profile





Likes

Likes

## 1. API Calling: tweepy consumer\_key, consumer\_secret, access\_token, access\_token\_secret

2. Get User information: get\_user()

Id
name
Screen\_name
Location
url
Description...

3. Exact tweets: user\_timeline()

## Data Collection

## Tweets collected

0 On #WorldAIDSDay, #JNJ is proud to share the powerful story of employee Fernando Salinas, who shortly after adoptin… https://t.co/oXRGNrI06e
1 "I did what many wouldn'tâ€"I removed my gloves and touched the patients.†â€" @5BFilm's Alison Paolercio, a nurse who… https://t.co/zefy5jBu23
2 "The way to deal with the shame and trauma, in part, is to talk about it and find ways to heal."â€"@Lavernecox, the f… https://t.co/3X179mO4i7
3 Meet Dr. @SchuitemakerH, the trailblazing HIV researcher who leads viral vaccine programs at @JanssenGlobal. Ahead… https://t.co/aNBAKCWRhG
4 #DYK before BAND-AID® Brand adhesive bandages, home cooks treated cuts & nicks with strips of gauze they had to tie… https://t.co/RmhGGi6OOB
5 Ahead of #WorldAIDSDay, #JNJ is proud to share the story of @5BFilm's Cliff Morrison & Damp; champion nurses on the front… https://t.co/VQy7Mxr0IW
6 RT @Ethicon: Strategic acquisitions & partnerships are crucial to our mission to advance lung surgery. Through our acquisition of Auris Hea…
7 The Lung Cancer Initiative at #JNJ is working to develop solutions that aim to prevent, intercept, and cure… https://t.co/m3JEB4Mvk1
8 #JNJ is proud to join Emmy-nominated actress & amp; activist @LaverneCox in the crusade to #makeHIVhistory. In honor of… https://t.co/UtmY3DUWkU
9 When four millennial #JNJ scientists noticed they were experiencing common skin dilemmas like fine lines & dark spo… https://t.co/Rab13bH9tq
10 As families across the U.S. prepare for #Thanksgiving, go back in @JNJHistory and learn about the Laurel Club, an a… https://t.co/h54C3IR28p
11 @HeartRiteHealth Hi there. Thanks for your interest in doing business with Johnson & Donson. We ask that you visit… https://t.co/SxI0R4HVzd
12 RT @JNJHistory: #JNJ has always inspired its employees to volunteerâ€"from our Laurel Club's local efforts in the early 1900s, to today's Glo…
13 Empowering nurses to develop & move their innovative ideas to propel patient care is just one way #JNJ champions nu… https://t.co/6QnfKzFBJq
14 From generous work-family time to subsidized childcare expenses, #JNJ is helping ensure families of all shapes… https://t.co/poCM49glTF
15 #DYK trillions of microorganisms called the microbiome co-exist in and on your bodyâ€"and could transform skin in sur… https://t.co/WEZ2l97kRD
16 RT @USGLC: Thrilled to have @JNJCares as the sponsor for tonight's #USGLCtribute dinner musical guests, the @ACChoir! These dynamic musicia…
17 @JanssenGlobal Learn more about the many innovative ways #JNJ is working to build a diverse workforce: https://t.co/csT5mn1dVi #BelongatJNJ
18 "We can only achieve gender parity by having men & women partnering in the journey. We need to take bolder actionsâ€"… https://t.co/XEQy1FGAQL
19 RT @JanssenEMEA: "Be intentional about your decisions - every time you make one ask yourself, is this a chance to advance diversity & inclu…

#### **NLTK**

Provides a set of diverse natural languages algorithms. Consists of the most common algorithms such as tokenizing, part-of-speech tagging, stemming, sentiment analysis, topic segmentation, and named entity recognition.

#### 1. Tokenization

- Sentence Tokenization
- Word Tokenization

#### 2. Frequency Distribution

most\_common()

#### 3. Remove 'noise' words

- Set stopwords
- Remove Stopwords

# 3 Most Common Words

# JNJ 3 Most Common Words

Market
 Research
 Global

10

```
#Most common words in JNJ tweets
JNJ_fdist.most_common(5)
[('market', 60),
 ('research', 26),
 ('global', 15),
 ('lenses', 14),
('20192025', 14)]
#Plot frequency distribution of tweets from JNJ
JNJ_fdist.plot(20,cumulative=False)
plt.show()
Sounts
30
   20
```

# MRK 3 Most Common Words

- 1. Diabetes
- 2. Multiplesclerosis
- 3. HIV

```
#Most common words in MRK tweets
MRK fdist.most common(7)
[('diabetes', 7),
 ('multiplesclerosis', 7),
('hiv', 5),
 ('lungcancer', 4),
 ('dyk', 4),
 ('healthcare', 4),
 ('health', 4)]
#Plot frequency distribution of tweets from JNJ
MRK_fdist.plot(20,cumulative=False)
plt.show()
  5.5
  4.0
  3.5
```

## MDT 3 Most Common Words

- 1. AHA2019\*
- 2. MedtronicAcademy
- 3. Heart failure

```
#Most common words in MDT tweets
MDT fdist.most common(5)
[('aha19', 8),
 ('medtronicacademy', 5),
 ('heart', 4),
 ('failure', 4),
 ('vadtherapy', 4)]
#Plot frequency distribution of tweets from JNJ
MDT_fdist.plot(20,cumulative=False)
plt.show()
```

#### Wordcloud

Functionality to create pretty word clouds, visualize differences and similarity between documents, and avoid over-plotting in scatter plots with text.

#### 1. Prepare data

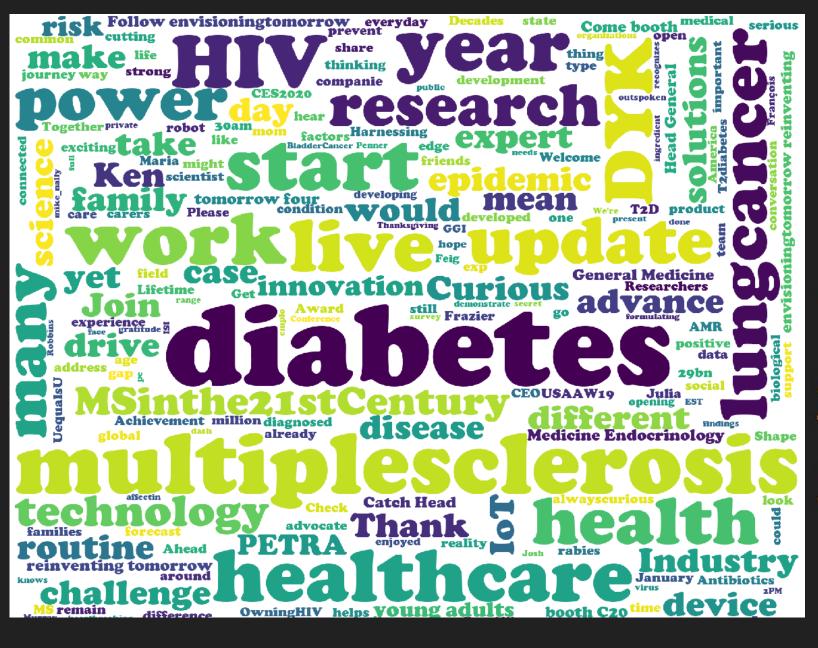
- Tokenize
- Remove URLs, RTs, and twitter handles
- 2. Set stopwords
- 3. Plot wordcloud

## Wordcloud



JNJ 3 Most Common Words

- 1. Market
- 2. Research
- 3. Global



## MRK 3 Most Common Words

- 1. Diabetes
- 2. Multiplesclerosis
- 3. HIV



MDT
3 Most
Common Words

- 1. AHA2019\*
- 2. MedtronicAcademy
- 3. Heart failure

Vader Sentiment Analysis (Valence Aware Dictionary and sEntiment Reasoner)
A lexicon and rule-based sentiment analysis tool
Specifically attuned to sentiments expressed in social media. VADER uses a combination of A sentiment lexicon is a list of lexical features (e.g., words) which are generally labelled according to their semantic orientation as either positive or negative.

#### 1. Prepare data

- Tokenize
- Remove URLs, RTs, and twitter handles
- 2. Print polarity\_scores
- 3. Summarize the scores

# Sentiment Analysis

```
#Detect sentiment using Vader
sid = SentimentIntensityAnalyzer()
for sentence in MDT[1]:
    print(sentence)
    ss = sid.polarity_scores(sentence)
    for k in sorted(ss):
        print('{0}: {1}, '.format(k, ss[k]), end='')
        print()
```

```
We're committed to working with public and private organizations to help address the full range of factors affectin... compound: 0.5859, neg: 0.0, neu: 0.769, pos: 0.231, For more than 30 years, we have been committed to addressing the global challenges of the HIV epidemic but our work... compound: 0.1779, neg: 0.0, neu: 0.876, pos: 0.124, We have the secret ingredient to formulating a great #Thanksgiving: a dash of gratitude for all our patients, emplo... compound: 0.8126, neg: 0.0, neu: 0.67, pos: 0.33,
```

```
#summarize the score
MDT_summary = {"positive":0,"neutral":0,"negative":0}
for x in MDT[1]:
    ss = sid.polarity_scores(x)
    if ss["compound"] == 0.0:
        MDT_summary["neutral"] +=1
    elif ss["compound"] > 0.0:
        MDT_summary["positive"] +=1
    else:
        MDT_summary["negative"] +=1
print(MDT_summary)

{'positive': 40, 'neutral': 20, 'negative': 6}
```

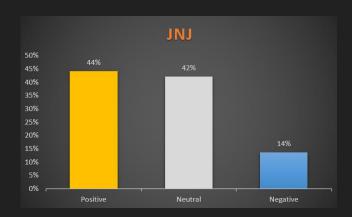
## **Sentiment Analysis**

#### O JNJ

Positive: 42

Neutral: 40

• Negative: 13

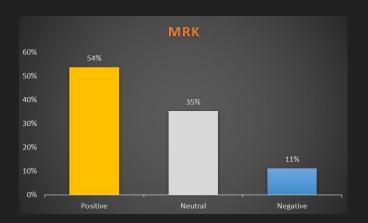


#### O MRK

• Positive: 58

Neutral: 38

• Negative: 12

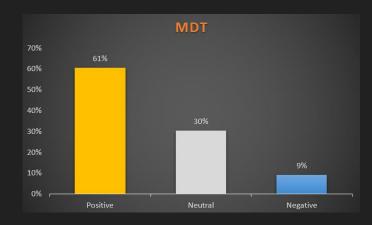


#### O MDT

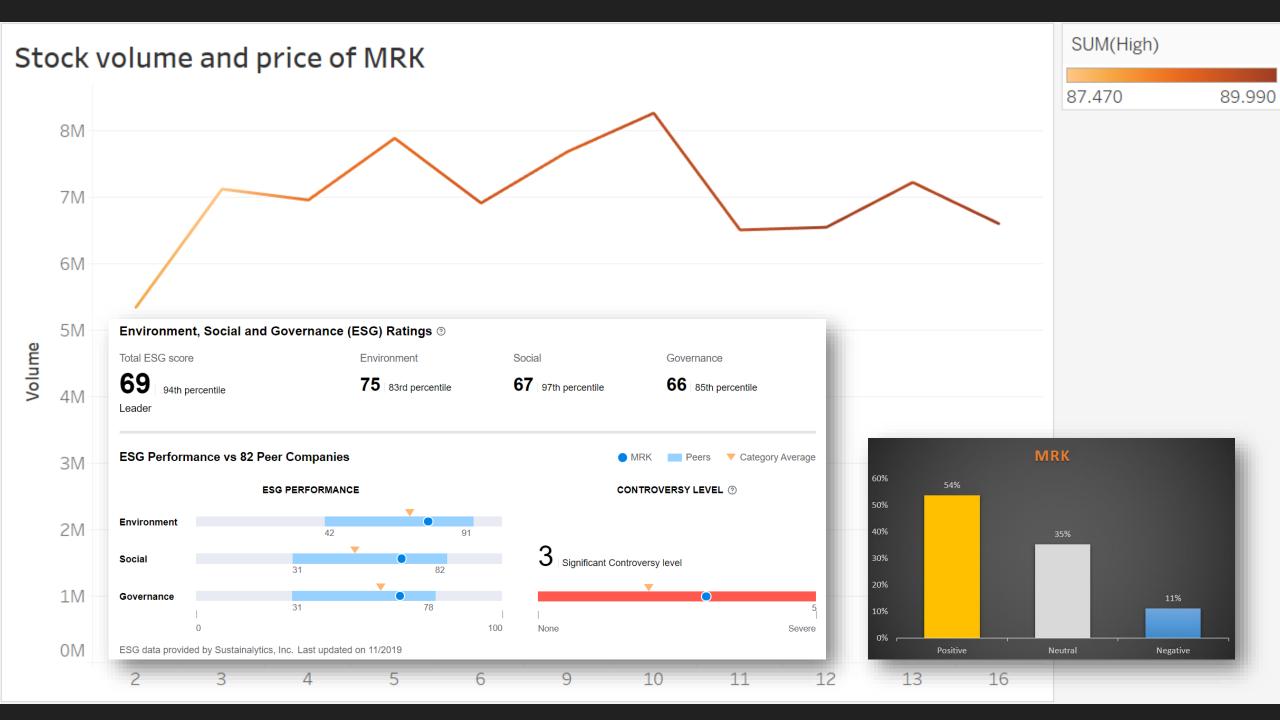
• Positive: 40

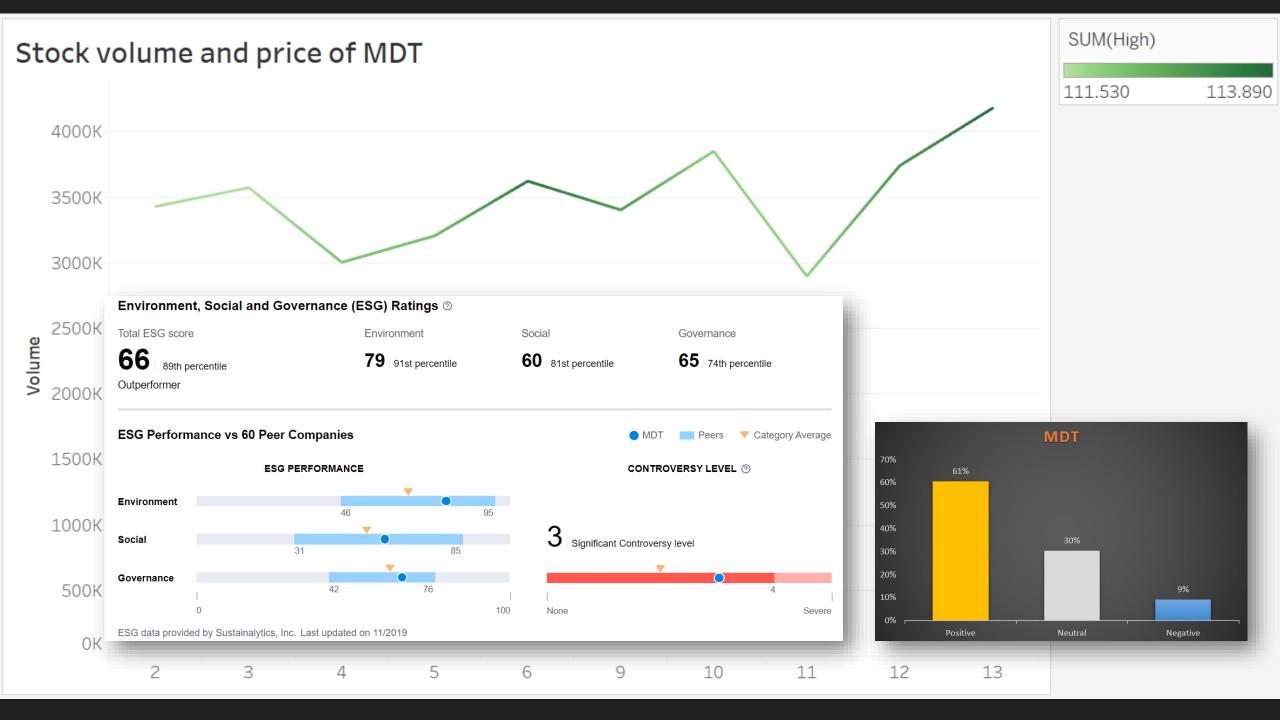
Neutral: 20

• Negative: 6











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