

#### 1. Importing Necessary Packages and setting the Work Directory:

```
In [ ]: import re
        import string
        import pandas as p
        import numpy as np
        import datetime
        import os
        import nltk
        from nltk.tokenize import RegexpTokenizer
        from nltk.tokenize import word_tokenize
        from nltk.corpus import stopwords
        from string import punctuation
        from nltk.stem import PorterStemmer
        from nltk.stem import WordNetLemmatizer
        os.path.abspath(os.getcwd())
        os.chdir('K:/RajK/Final')
        input path = 'K:/RajK/Final'
        output path = 'K:/RajK/Final/cleaned'
        import warnings
        warnings.filterwarnings("ignore")
```

## 2. Assigning stop words, root words, punctuations and Personal Pronoun list:

```
In []: lemmatizer = WordNetLemmatizer()
    my_stopwords = nltk.corpus.stopwords.words('english')
    word_rooter = nltk.stem.snowball.PorterStemmer(ignore_stopwords=False).stem
    my_punctuation = '!"$%&\'()*+,-./:;<=>?[\\]^_`{|}~~@â'
    Semantics=['linked to','threw me','induces','gave me','give me','resulted into',
    'resulted to','as a result','as a result of', 'because of', 'owing to', 'as cons
    equence of', 'by reason of', 'on account of', 'in consequence of', 'on the basis
    of','because','due to','consequence of','because of','causes into', 'caused', 't
    riggered','led to','leads']
    IMeMyself=['i','im', 'me', 'my', 'mine', 'myself', 'she', 'he', 'her', 'him', 'y
    ou', 'thou', 'your', 'have', 'has', 'we']
```

#### 3. Creating a Master Clean Function:

#### 3.1 Importing All Side-Effects List:

```
In []: All_Side_Effects = p.DataFrame(p.read_excel("ALL_SideEffects.xlsx"))
    List_SideEffects = []
    for i in range(len(All_Side_Effects)):
        SE = All_Side_Effects.SE[i]
        List_SideEffects.append(SE)
```

#### 3.2 Drug Specific Listed Sede-Effect Importing Function:

#### 3.3 Cleaning Function that returns Cleaned Tweet Tockens and Cleaned Tweets:

```
In [ ]: def remove_links(tweet):
            tweet = re.sub(r'http\S+', '', tweet)
            tweet = re.sub(r'bit.ly/\S+', '', tweet)
            tweet = tweet.strip('[link]')
            return tweet
        def remove_users(tweet):
            tweet = re.sub('(@[A-Za-z]+[A-Za-z0-9-]+)^{-}, '', tweet)
           return tweet
        def reduce lengthening(word):
            pattern = re.compile(r''(.) \setminus 1\{2,\}'')
            return pattern.sub(r"\1\1", word)
        from autocorrect import Speller
        from autocorrect import Speller
        spell = Speller(lang='en')
        def clean tweets (tweet, bigrams=False):
           tweet = remove users(tweet)
           tweet = remove links(tweet)
           tweet = tweet.lower() # lower case
           tweet = re.sub(r'#([^\s]+)', r'\1', tweet)
           tweet = re.sub('['+my punctuation + ']+', ' ', tweet)
            tweet = re.sub('\s+', '', tweet)
            tweet = re.sub('([0-9]+)', '', tweet)
            tweet token list = [word for word in tweet.split(' ') if word not in my stop
            tweet token list = [lemmatizer.lemmatize(word) if '#' not in word else word
                               for word in tweet token list]
            if bigrams:
               tweet token list = tweet token list+[tweet token list[i]+' '+tweet token
        list[i+1]
                                                   for i in range (len (tweet token lis
        t) - 1)]
            tweet = ' '.join(tweet token list)
            return tweet
        def clean tokens (tweet):
            tweet token list = [word for word in tweet.split(' ') if word not in my stop
        wordsl
           while ('') in tweet_token_list: tweet_token_list.remove('')
            return tweet_token_list
```

### 3.4 Sorting and Filtering Algorith, that counts the presence of Listed as well as Un-listed Side-Effects in a tweet:

```
In [ ]: def count_prescribed (cleaned_tweet):
            j=0
            prescribed_token = []
            for i in (Drug1 SideEffects):
                if ((cleaned tweet.find(i) != -1) and (i not in prescribed token)):
                    j += 1
                    prescribed token.append(i)
            Dict = {j : sorted(prescribed token)}
            return Dict
        def count UNprescribed (cleaned tweet):
            j=0
            UNprescribed token = []
            for i in (List SideEffects):
                if ((cleaned tweet.find(i) != -1) and (i not in Drug1 SideEffects) and
        (i not in UNprescribed_token)):
                    j+=1
                    UNprescribed token.append(i)
            Dict = {j : sorted(UNprescribed token)}
            return Dict
```

# 3.5 Checking for Organizational Tweeter handle based on I ME Myself Words, and Assigning 1 to Organizatinal Handles. Also, writing a function for Sementic Word Check in the Tweet and assign Sementic Score of 1 for Potential Tweets:

```
In [ ]: | def org check (tweet):
              tweet = tweet.lower() # lower case
              tweet = re.sub(r'#([^\s]+)', r'\1', tweet) # remove the # in #hashtag
              tweet = re.sub('['+my punctuation + ']+', ' ', tweet) # strip punctuation
              \#tweet = re.sub('\s+', '', tweet) \#remove double spacing \#tweet = re.sub('([0-9]+)', '', tweet) \# remove numbers
              tweet token list = [word for word in tweet.split(' ')]
              for word in tweet token list:
                  if (word in IMeMyself):
                      a=0
                      break
                  else: a=1
              return a
         def SementicScore (cleaned_tweet):
              j=0
              for i in Semantics:
                  if (cleaned tweet.find(i) != -1):
                      j+=1
                  else: continue
              return i
```

#### 3.6 Combining All the previously made functions in to a common Master Function:

```
In [ ]: def clean master (df, Org 1, Clean Tweets, Clean Tweet Tokens, Prescribed Key Va
        lue, UNPrescribed Key Value, Sementic score):
                                     Step 1 of 7 >>> Cleaning Started")
            print("
            df[Org_1]=df.Tweet.apply(org_check)
                                      Step 2 of 7 >>> Organization Checked")
            print ("
            df[Clean Tweets] = df. Tweet.apply(clean tweets)
                                      Step 3 of 7 >>> Tweets Cleaned")
            df[Clean Tweet Tokens]=df.Clean Tweets.apply(clean tokens)
                                     Step 4 of 7 >>> Cleaned Tweets Tocanised")
            df[Prescribed_Key_Value] = df.Clean_Tweets.apply(count_prescribed)
                                     Step 5 of 7 >>> Key Value pair for Prescribed Prep
        ared")
            df[UNPrescribed_Key_Value] = df.Clean_Tweets.apply(count_UNprescribed)
                                      Step 6 of 7 >>> Key Value pair for UN Prescribed P
            df[Sementic score] = df.Clean Tweets.apply(SementicScore)
            print ("
                                      Step 7 of 7 >>> Semantic Scoring Done")
            return df
        def drugClean(UFfile, SEfile, OPname):
            Data=p.DataFrame(p.read_csv(UFfile))
            Drug1 SideEffects = DrugSEClean(SEfile)
            Clean_Data = clean_master(Data,"Org1","Clean_Tweets", "Clean_Tweet_Tokens",
        "Prescribed_Key_Value", "UNPrescribed_Key_Value", "Senamtic_Scores")
            return (Clean_Data.to_excel(r'K:/RajK/Final/Semantic/'+str(OPname)))
```

#### 4. Deployment of Master Function for All the 7 Drugs:

```
----- Drug Cleaning Master Function Init
Cleaning Started for Drug No. 0 Out of 7 -> 1Albuterol
Cleaning Started
Organization Checked
Tweets Cleaned
Cleaned Tweets Tocanised
Key Value pair for Prescribed Prepared
Key Value pair for UN Prescribed Prepared
Semantic Scoring Done
----- Drug Cleaning Master Function Init
_____
Cleaning Started for Drug No. 1 Out of 7 -> 2Amoxicillin
Cleaning Started
Organization Checked
Tweets Cleaned
Cleaned Tweets Tocanised
Key Value pair for Prescribed Prepared
Key Value pair for UN Prescribed Prepared
Semantic Scoring Done
----- Drug Cleaning Master Function Init
_____
Cleaning Started for Drug No. 2 Out of 7 -> 3Levothyroxine
Cleaning Started
Organization Checked
Tweets Cleaned
Cleaned Tweets Tocanised
Key Value pair for Prescribed Prepared
Key Value pair for UN Prescribed Prepared
Semantic Scoring Done
----- Drug Cleaning Master Function Init
_____
Cleaning Started for Drug No. 3 Out of 7 -> 4Lipitor
Cleaning Started
Organization Checked
Tweets Cleaned
Cleaned Tweets Tocanised
Key Value pair for Prescribed Prepared
Key Value pair for UN Prescribed Prepared
Semantic Scoring Done
----- Drug Cleaning Master Function Init
-----
Cleaning Started for Drug No. 4 Out of 7 -> 5Sertraline
Cleaning Started
Organization Checked
Tweets Cleaned
Cleaned Tweets Tocanised
Key Value pair for Prescribed Prepared
Key Value pair for UN Prescribed Prepared
Semantic Scoring Done
----- Drug Cleaning Master Function Init
_____
Cleaning Started for Drug No. 5 Out of 7 -> _6Warfarin
Cleaning Started
Organization Checked
Tweets Cleaned
Cleaned Tweets Tocanised
```

7 of 7