DOKUMENTASI TUGAS AKHIR DATA MINING A11.4408

Nama : Alfebrian Ivo Kurnia Adi

NIM: A11.2021.13920

```
import numpy as np
import pandas as pd
import os
```

```
df = pd.read_csv('Tweets.csv')
df.head()
```

Output:

0 570366135677786513 neutral 1,0000 NaN NaN NaN NaN NaN Caridin NaN 0 6Virgin-herica What NaN 115552-8000 NaN		444 04 0	2.40 N. J. 546		7.00		ration that it say		200	2 2 2	304	No. 10	W 14 TEVE	S 000 Na	
1 570301130888122388 positive 0.3486 NaN 0.0000 Mirgin NaN practice NaN 0.0000 America NaN practice NaN 0.0000 Mirgin NaN practice NaN 0.0000 Mirgin NaN practice NaN 0.0000 Mirgin NaN 0.00000 Mirgin NaN 0.0000 Mirgin NaN 0.00000 Mirgin NaN 0.000000 Mirgin NaN 0.00000 Mirgin NaN 0.000000 Mirgin NaN 0.00000 Mirgin NaN	tweet_id	airline_sentiment	airline_sentiment_confidence	negativereason n	egativereason_confidence	airline a	irline_sentiment_gold	nane	negativereason_gold	retweet_count	text	tweet_coord	tweet_created	tweet_location	user_timezone
1 57039/138888122388 positive 0 3496 NaN 0,0000 America NaN paratro NaN 0 you've added commercials NaN 11559-84000 NaN 29704244 11559-84000 NaN 2970424 115904 1	0 570306133677760513	neutral	1.0000	NaN	NaN	Virgin America	NaN	cairdin	NaN			NaN		NaN	Eastern Time (US & Canada)
America Nam Promonent Vision Nam	1 570301130888122368	positive	0.3496	NaN	0.0000	Virgin America	NaN	jnardino	NaN					NaN	Pacific Time (US & Canada)
7 STUDY TO SHOULD HELD TO THE TOUR SHOULD HELD TOUR SHOULD HE HELD TOUR SHOULD HE HEL	2 570301083672813571	neutral	0.6837	NaN	NaN	Virgin America	NaN	yvonnalynn	NaN			NaN		Lets Play	Central Time (US & Canada)
4 STROMENTULASTYYY nestrine 10001 Cart Tel 10000 Wight NaN partin NaN 0 (Wighthests and Es NaN 2015-12-24 NaN Pacific Time (US	3 570301031407624196	negative	1,0000	Bad Flight	0.7033	Virgin America	NaN	jnardino	NaN			NaN		NaN	Pacific Time (US & Canada)
America really big bad filing. 11.14.45-4000 Cario	4 570300817074462722	negative	1,0000	Can't Tell	1,0000	Virgin America	NaN	jnardno	NaN		@WirginAmerica and it's a really big bad thing	NaN	2015-02-24 11:14:45-0800	NaN	Pacific Time (US & Canada)

```
import sqlite3
conn = sqlite3.connect('database.sqlite')
```

```
cur = conn.cursor()
cur.execute("Select * From Tweets Limit 5")
```

Output:

<sqlite3.Cursor at 0x7b276869c2c0>

```
rows = cur.fetchall()
for row in rows:
    print(row)
```

Output:

```
CONTENTIONING, which is a "Silly", "Sil
```

df.describe()

Output:

tweet_ id	airline_sentiment_con fidence	negativereason_conf idence	retweet_co unt	
count	1.464000e+04	14640.000000	10522.0000 00	14640.000 000
mean	5.692184e+17	0.900169	0.638298	0.082650
std	7.791112e+14	0.162830	0.330440	0.745778
min	5.675883e+17	0.335000	0.000000	0.000000
25%	5.685592e+17	0.692300	0.360600	0.000000
50%	5.694779e+17	1.000000	0.670600	0.000000
75%	5.698905e+17	1.000000	1.000000	0.000000
max	5.703106e+17	1.000000	1.000000	44.000000

df.info()

Output:

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 14640 entries, 0 to 14639
Data columns (total 15 columns):
                                       Non-Null Count Dtype
   tweet id
                                       14640 non-null int64
                                       14640 non-null object
     airline_sentiment_confidence 14640 non-null float64 negativereason 9178 non-null object negativereason_confidence 10522 non-null float64
    negativereason
    negativereason confidence
                                       14640 non-null object
                                       40 non-null object
                                      14640 non-null object
    negativereason gold
                                      32 non-null object
                                      14640 non-null int64
                                      14640 non-null object
                                      1019 non-null object
14640 non-null object
                                      9907 non-null object
                                       9820 non-null object
dtypes: float64(2), int64(2), object(11)
memory usage: 1.7+ MB
```

Output:

```
df['airline_sentiment'].unique()
```

Output:

```
array(['neutral', 'positive', 'negative'], dtype=object)
```

```
df['airline_sentiment'].value_counts()
```

Output:

```
negative 9178
neutral 3099
positive 2363
```

Name: airline_sentiment, dtype: int64

```
def execute_sql(command):
    # Function that execute command from the database and print the
    results
        cur = conn.cursor()
        cur.execute(command)
        rows = cur.fetchall()
        for row in rows:
            print(row)
```

```
execute_sql("Select DISTINCT(airline_sentiment) From Tweets")
('neutral',)
('negative',)
('positive',)
```

Output:

```
('neutral',)
('negative',)
('positive',)
('positive',)
```

```
execute_sql("Select airline_sentiment, COUNT(airline_sentiment) From
Tweets GROUP BY airline_sentiment")
('negative', 9082)
('neutral', 3069)
('positive', 2334)
```

Output:

```
('negative', 9082)
('neutral', 3069)
('positive', 2334)
('positive', 2334)
```

```
pd.read_sql("Select DISTINCT(airline_sentiment) From Tweets", conn)
```

Output:

airline_sentiment

0 neutral

1 negative

2 positive

```
df_neg = df.loc[(df.airline_sentiment == 'negative') &
  (df.negativereason != "Can't Tell")].reset_index(drop=True)
df neg.shape
```

Output:

(7988, 15)

Output:

(7906, 15)

```
df neg.groupby('negativereason').negativereason confidence.mean()
```

Output:

```
negativereason
```

Bad Flight 0.631731

Cancelled Flight 0.783096

Customer Service Issue 0.780054

Damaged Luggage 0.733432

Flight Attendant Complaints 0.659639

Flight Booking Problems 0.606797

Late Flight 0.768907

Lost Luggage 0.813019

longlines 0.594076

Name: negativereason confidence, dtype: float64

Output:

negativereason	average_confidence	
0	Bad Flight	0.630785
1	Cancelled Flight	0.783200
2	Customer Service Issue	0.779946
3	Damaged Luggage	0.734204
4	Flight Attendant Complaints	0.658255
5	Flight Booking Problems	0.607153
6	Late Flight	0.768978
7	Lost Luggage	0.812209

negativereason

average_confidence

8

longlines 0.593856