

In [1]:

```
import tensorflow as tf
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
from sklearn.preprocessing import OneHotEncoder, MinMaxScaler
from sklearn.compose import make_column_transformer
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score
from sklearn.tree import DecisionTreeClassifier
from sklearn.svm import SVC
```

In [2]:

```
tweet = pd.read_csv("Tweets.csv")
len(tweet)
```

Out[2]:

14640

In [3]:

```
tweet.head()
```

Out[3]:

|   | tweet_id           | airline_sentiment | airline_sentiment_confidence | negativereason | nega |
|---|--------------------|-------------------|------------------------------|----------------|------|
| 0 | 570306133677760513 | neutral           | 1.0000                       | NaN            |      |
| 1 | 570301130888122368 | positive          | 0.3486                       | NaN            |      |
| 2 | 570301083672813571 | neutral           | 0.6837                       | NaN            |      |
| 3 | 570301031407624196 | negative          | 1.0000                       | Bad Flight     |      |
| 4 | 570300817074462722 | negative          | 1.0000                       | Can't Tell     |      |



In [4]:

```
tweet.describe()
```

Out[4]:

|       | tweet_id     | airline_sentiment_confidence | negativereason_confidence | retweet_count |
|-------|--------------|------------------------------|---------------------------|---------------|
| count | 1.464000e+04 | 14640.000000                 | 10522.000000              | 14640.000000  |
| 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     |



In [5]:

```
tweet.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 14640 entries, 0 to 14639
Data columns (total 15 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   tweet_id                             14640 non-null  int64
1   airline_sentiment                    14640 non-null  object
2   airline_sentiment_confidence         14640 non-null  float64
3   negativereason                       9178 non-null   object
4   negativereason_confidence            10522 non-null  float64
5   airline                              14640 non-null  object
6   airline_sentiment_gold                40 non-null     object
7   name                                 14640 non-null  object
8   negativereason_gold                   32 non-null     object
9   retweet_count                        14640 non-null  int64
10  text                                 14640 non-null  object
11  tweet_coord                           1019 non-null   object
12  tweet_created                         14640 non-null  object
13  tweet_location                        9907 non-null   object
14  user_timezone                         9820 non-null   object
dtypes: float64(2), int64(2), object(11)
memory usage: 1.7+ MB
```

In [7]:

```
def deal_missing_values(X_full):
    #drop col where data is very less
    X_full = X_full.drop('airline_sentiment_gold', axis=1)
    X_full = X_full.drop('negativereason_gold', axis=1)
    X_full = X_full.drop('tweet_coord', axis=1)
    # replace null values with mean
    X_full['negativereason_confidence'] = X_full['negativereason_confidence'].fillna(X_f
    return X_full
```

```
tweet = deal_missing_values(tweet)
tweet.info()
```

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

RangeIndex: 14640 entries, 0 to 14639

Data columns (total 12 columns):

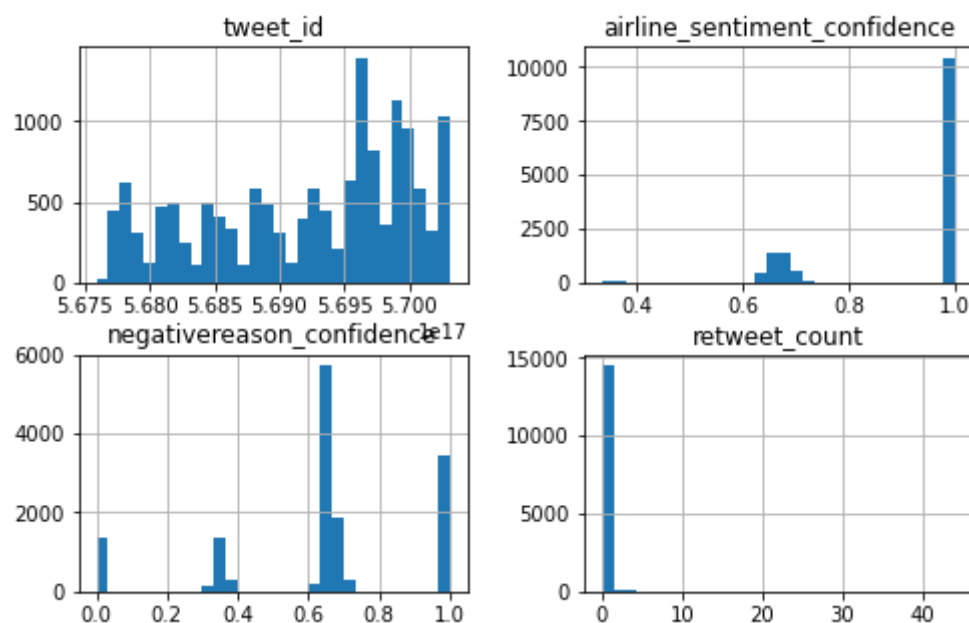
| #  | Column                       | Non-Null Count | Dtype   |
|----|------------------------------|----------------|---------|
| 0  | tweet_id                     | 14640 non-null | int64   |
| 1  | airline_sentiment            | 14640 non-null | object  |
| 2  | airline_sentiment_confidence | 14640 non-null | float64 |
| 3  | negativereason               | 9178 non-null  | object  |
| 4  | negativereason_confidence    | 14640 non-null | float64 |
| 5  | airline                      | 14640 non-null | object  |
| 6  | name                         | 14640 non-null | object  |
| 7  | retweet_count                | 14640 non-null | int64   |
| 8  | text                         | 14640 non-null | object  |
| 9  | tweet_created                | 14640 non-null | object  |
| 10 | tweet_location               | 9907 non-null  | object  |
| 11 | user_timezone                | 9820 non-null  | object  |

dtypes: float64(2), int64(2), object(8)

memory usage: 1.3+ MB

In [8]:

```
tweet.hist(bins = 30, figsize = (8,5))
plt.show()
```



In [9]:

```
(tweet['airline'].unique())
```

Out[9]:

```
array(['Virgin America', 'United', 'Southwest', 'Delta', 'US Airways',  
      'American'], dtype=object)
```

In [10]:

```
(tweet['negativereason'].unique())
```

Out[10]:

```
array([nan, 'Bad Flight', "Can't Tell", 'Late Flight',  
      'Customer Service Issue', 'Flight Booking Problems',  
      'Lost Luggage', 'Flight Attendant Complaints', 'Cancelled Flight',  
      'Damaged Luggage', 'longlines'], dtype=object)
```

In [11]:

```
tweet.tail()
```

Out[11]:

|       | tweet_id           | airline_sentiment | airline_sentiment_confidence | negativereason         |
|-------|--------------------|-------------------|------------------------------|------------------------|
| 14635 | 569587686496825344 | positive          | 0.3487                       | NaN                    |
| 14636 | 569587371693355008 | negative          | 1.0000                       | Customer Service Issue |
| 14637 | 569587242672398336 | neutral           | 1.0000                       | NaN                    |
| 14638 | 569587188687634433 | negative          | 1.0000                       | Customer Service Issue |
| 14639 | 569587140490866689 | neutral           | 0.6771                       | NaN                    |



In [19]:

```
X = tweet.drop('airline_sentiment', axis = 1)  
y = tweet['airline_sentiment']
```

In [20]:

```
from sklearn.compose import make_column_transformer
from sklearn.preprocessing import MinMaxScaler, OneHotEncoder
from sklearn.model_selection import train_test_split

ct = make_column_transformer(
    (MinMaxScaler(), ["tweet_id"]),
    (OneHotEncoder(handle_unknown="ignore"), ["airline", "retweet_count"])
)

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)

ct.fit(X_train)
X_train_normal = ct.transform(X_train)
X_test_normal = ct.transform(X_test)
```

In [22]:

```
lr_model = LogisticRegression(max_iter = 1000)
lr_model.fit(X_train_normal, y_train)
tree_model = SVC()
tree_model.fit(X_train_normal, y_train)
```

Out[22]:

```
▼ SVC
SVC()
```

In [23]:

```
y_pred = lr_model.predict(X_test_normal)
accuracy = accuracy_score(y_test, y_pred)
y_pred_tree = tree_model.predict(X_test_normal)
accuracy_tree = accuracy_score(y_test, y_pred_tree)
print("Accuracy: ", accuracy_tree)
```

Accuracy: 0.6454918032786885