Goals

- sentiment analysis on twitter thread interactions.
- train a model to be able to predict conversation sentiment.
 - o based on the first tweet?
 - o proactive measures for dealing with customers.

```
!pip install -q kaggle
from google.colab import files
files.upload()
!mkdir ~/.kaggle
!cp kaggle.json ~/.kaggle/
!chmod 600 ~/.kaggle/kaggle.json
!kaggle datasets download -d thoughtvector/customer-support-on-twitter
```

#in drive https://drive.google.com/drive/u/1/folders/1Sh4w-8e1p2Yl_9QrAvpgRf4nmeY_KW-z

```
Choose Files kaggle.json
```

• kaggle.json(application/json) - 67 bytes, last modified: 4/26/2021 - 100% done
Saving kaggle.json to kaggle (1).json
mkdir: cannot create directory '/root/.kaggle': File exists
customer-support-on-twitter.zip: Skipping, found more recently modified local copy (use

```
!pip install cloudmesh-installer
!pip install cloudmesh-common
```

Requirement already satisfied: pytest-cov in /usr/local/lib/python3.7/dist-packages Requirement already satisfied: cloudmesh-common in /usr/local/lib/python3.7/dist-pa Requirement already satisfied: oyaml in /usr/local/lib/python3.7/dist-packages (fro Requirement already satisfied: humanize in /usr/local/lib/python3.7/dist-packages (Requirement already satisfied: pip-tools in /usr/local/lib/python3.7/dist-packages Requirement already satisfied: psutil in /usr/local/lib/python3.7/dist-packages (fr Requirement already satisfied: pipdeptree in /usr/local/lib/python3.7/dist-packages Requirement already satisfied: wheel in /usr/local/lib/python3.7/dist-packages (fro Requirement already satisfied: python-hostlist in /usr/local/lib/python3.7/dist-pac Requirement already satisfied: pytest in /usr/local/lib/python3.7/dist-packages (fr Requirement already satisfied: bump2version==1.0.0 in /usr/local/lib/python3.7/dist Requirement already satisfied: idna<3,>=2.5 in /usr/local/lib/python3.7/dist-packag Requirement already satisfied: chardet<4,>=3.0.2 in /usr/local/lib/python3.7/dist-r Requirement already satisfied: urllib3!=1.25.0,!=1.25.1,<1.26,>=1.21.1 in /usr/loca Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.7/dist-Requirement already satisfied: mccabe<0.7.0,>=0.6.0 in /usr/local/lib/python3.7/dis Requirement already satisfied: pycodestyle<2.8.0,>=2.7.0 in /usr/local/lib/python3. Requirement already satisfied: pyflakes<2.4.0,>=2.3.0 in /usr/local/lib/python3.7/c Requirement already satisfied: importlib-metadata; python_version < "3.8" in /usr/] Requirement already satisfied: pyfiglet in /usr/local/lib/python3.7/dist-packages (Requirement already satisfied: simplejson in /usr/local/lib/python3.7/dist-packages Requirement already satisfied: todm in /usr/local/lib/nython3.7/dist-nackages (from

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Import libraries

```
import time
from cloudmesh.common.StopWatch import StopWatch
from cloudmesh.common.Benchmark import Benchmark
from cloudmesh.common.Shell import Shell

import zipfile

#dealing with data
import matplotlib.pyplot as plt
import pandas as pd
import numpy as np
import tensorflow as tf
```

```
import keras
from keras.preprocessing.text import Tokenizer
from keras.preprocessing import sequence
import sklearn
from sklearn.model selection import train test split
#nl libraries
import string
import nltk #natural language tool kit
from nltk.sentiment.vader import SentimentIntensityAnalyzer
nltk.download("vader lexicon")
nltk.download('punkt')
     [nltk data] Downloading package vader lexicon to /root/nltk data...
                   Package vader lexicon is already up-to-date!
     [nltk data]
     [nltk data] Downloading package punkt to /root/nltk data...
                 Package punkt is already up-to-date!
     [nltk data]
     True
```

Download and reduce the data working with

```
StopWatch.start("get data")
zfile=zipfile.ZipFile('/content/customer-support-on-twitter.zip')
data=pd.read csv(zfile.open('twcs/twcs.csv'))
StopWatch.stop("get_data")
print(data.shape)
data.head()
StopWatch.start("manageability")
df=data.drop(['inbound','created_at'],axis=1)
df["text"]=df["text"].astype(str)
df.sample(frac=1)
df=df.head(650)
StopWatch.stop("manageability")
     (2811774, 7)
StopWatch.start("removing companies")
#companies are not anon so they have actual letters
def companies(df):
  idx=[]
```

```
tor id in range(len(dt["author_id"])):
    #assuming at least one vowel in each company name
      vowels=['a','e','i','o', 'u']
      for vowel in vowels:
        if vowel in df["author id"][id]:
          idx.append(id)
          break
  return df.drop(idx)
#we only want the consumer side
customers=companies(df)
StopWatch.stop("removing_companies")
```

Analyze senitment for individual tweets

```
StopWatch.start("sentiment score")
sent analyzer=SentimentIntensityAnalyzer()
#analyze the raw tweets
customers["sentiment"]=customers["text"].apply(lambda x: sent_analyzer.polarity_scores(x)["co
StopWatch.stop("sentiment score")
```

→ Build a new dataframe

```
# #show distribution of overall tweets and by customers
cust average=customers.groupby("author id")
cust sent average=cust average.sentiment.mean()
cust sent average=pd.DataFrame({'author id':cust sent average.index, 'sentiment average': cus
#getting the first tweet for each author
customers = customers[pd.isnull(df.in_response_to_tweet_id)]
#new dataframe with the essentials
customer_first=pd.DataFrame({'author_id':customers['author_id'],
                             'tweet':customers['text'],
                             'first_sentiment':customers['sentiment'],
                             'classification': 'Na'})
#add the overall sentiment binary for the entire thread
for author in customer first['author id']:
 sentiment=cust_sent_average.loc[cust_sent_average.author_id==author, 'sentiment_average']
 if sentiment.values < 0:
    customer first.loc[customer first.author id==author,'classification'] = 0
```

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:7: UserWarning: Boolean Ser import sys

author_id		tweet	first_sentiment	classificat
6	115712	@sprintcare is the worst customer service	0.0	
12	115713	@115714 y'all lie about your "great" connectio	0.0	
14	115715	@115714 whenever I contact customer support, t	1.0	
23	115716	actually that's a broken link you sent me and		
25	115717	Yo @Ask_Spectrum, your customer service reps a	1.0	

show sentiment distribution

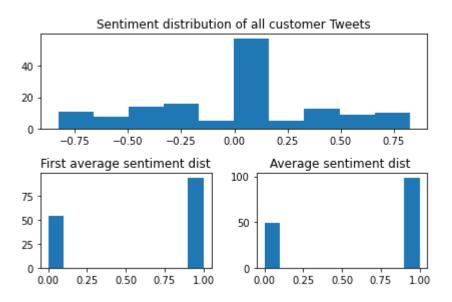
```
fig=plt.figure()

axs0=plt.subplot(211)
axs0.hist(customers["sentiment"])
axs0.set_title("Sentiment distribution of all customer Tweets")

axs1=plt.subplot(223)
axs1.hist(customer_first["first_sentiment"])
axs1.set_title("First average sentiment dist")

axs2=plt.subplot(224)
axs2.hist(customer_first["classification"])
axs2.set_title("Average sentiment dist ")
```

```
plt.tight_layout()
fig.savefig("customer_dist.png")
fig.show()
```



nltk.download('averaged perceptron tagger') def encode_tweets(df): count=0 encoded=[] embedding={} for scentence in df.tweet: encode=[] token=nltk.word_tokenize(scentence.lower()) token=[ele for word_tuple in nltk.pos_tag(token) for ele in word_tuple] for word in token: if word not in embedding: embedding[word]=count count+=1 encode.append(embedding[word]) encoded.append(encode) return (encoded) #endoce the natural text into something the network will be able to read encoded=encode tweets(customer first) #replace the raw text with the encoded text customer first['tweet']=encoded customer_first

```
[nltk_data] Downloading package averaged_perceptron_tagger to
[nltk_data] /root/nltk_data...
[nltk_data] Package averaged_perceptron_tagger is already up-to-
[nltk_data] date!
```

	author_id	tweet	first_sentiment	classification
6	115712	[0, 1, 2, 1, 3, 4, 5, 6, 7, 8, 9, 1, 10, 1]	0.0	0
12	115713	[0, 11, 12, 13, 14, 15, 16, 17, 18, 6, 19, 17,	0.0	0
14	115715	[0, 11, 12, 13, 42, 43, 44, 1, 45, 17, 9, 1, 4	1.0	1
23	115716	[69, 11, 70, 6, 71, 4, 72, 6, 73, 25, 74, 1, 7	0.0	1
25	115717	[81, 1, 0, 37, 82, 1, 33, 33, 22, 23, 9, 1, 10	1.0	1
633	115915	[1192, 1, 238, 11, 1033, 25, 56, 23, 125, 25,	0.0	0
635	115916	[143, 6, 29, 29, 70, 6, 1195, 1, 456, 4, 1196,	0.0	0
643	115918	[0, 1, 1185, 1, 0, 1, 1200, 13, 5, 6, 0, 1, 12	1.0	1
645	115919	[0, 1, 1206, 15, 238, 21, 44, 25, 984, 17, 112	0.0	0
648	115920	[0, 1, 1206, 15, 367, 17, 96, 29, 44, 1, 127,	1.0	1

148 rows × 4 columns

Dividing train and test data

```
train_sent=train_sent.values
test_sent=test_sent.values
train_sent=keras.utils.to_categorical(train_sent)
test_sent=keras.utils.to_categorical(test_sent)
test_text.shape

(37, 225)
```

- Cnn model

```
#Cnn libraries
import tensorflow as tf
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, Activation, Dropout
from keras.layers import Conv1D, MaxPooling1D, Flatten, AveragePooling1D, Embedding
from tensorflow.keras.utils import to_categorical, plot_model
```

Building the model

```
word dict=10000
num labels=2
input shape=test text.shape
batch_size=128
kernel size=3
pool_size=2
filters=64
dropout=0.2
epochs=5
StopWatch.start("cnn_model_building")
cnn model=Sequential()
cnn model.add(Conv1D(filters=filters, kernel size=kernel size,
                activation='relu', input_shape=input_shape,
                padding='same'))
cnn model.add(MaxPooling1D(pool size))
cnn_model.add(Conv1D(filters=filters, kernel_size=kernel_size,
```

```
activation='relu', input shape=input shape,
                padding='same'))
cnn model.add(MaxPooling1D(pool size))
cnn model.add(Conv1D(filters=filters, kernel size=kernel size,
                activation='relu', input_shape=input_shape,
                padding='same'))
cnn model.add(MaxPooling1D(pool size))
cnn model.add(Conv1D(filters=filters, kernel size=kernel size,
                activation='relu', input_shape=input_shape,
                padding='same'))
cnn model.add(Flatten())
cnn_model.add(Dropout(dropout))
cnn model.add(Dense(num labels))
cnn model.add(Activation('softmax'))
cnn model.summary()
plot model(cnn model, to file='cnn model.png', show shapes=True)
StopWatch.stop("cnn model building")
```

Model: "sequential_1"

Layer (type)	Output	Shape	Param #
conv1d_4 (Conv1D)	(None,	37, 64)	43264
max_pooling1d_3 (MaxPooling1	(None,	18, 64)	0
conv1d_5 (Conv1D)	(None,	18, 64)	12352
max_pooling1d_4 (MaxPooling1	(None,	9, 64)	0
conv1d_6 (Conv1D)	(None,	9, 64)	12352
max_pooling1d_5 (MaxPooling1	(None,	4, 64)	0
conv1d_7 (Conv1D)	(None,	4, 64)	12352
flatten_1 (Flatten)	(None,	256)	0
dropout_1 (Dropout)	(None,	256)	0
dense_1 (Dense)	(None,	2)	514
activation_1 (Activation)	(None,	2)	0
_			

Total params: 80,834 Trainable params: 80,834 Non-trainable params: 0

→ Compile

→ Model Fit

```
StopWatch.start("cnn_train")
cnn_model.fit(train_text, train_sent, batch_size=batch_size,epochs=epochs)
StopWatch.stop("cnn_train")

$\subsets \frac{1}{2}$
```

```
Epoch 1/5
```

ValueError

Traceback (most recent call last)

Predicting

▼ Evaluate

```
/usr/local/lih/nvthon3 7/dist-
StopWatch.start("cnn evaluate")
cnn_loss, cnn_accuracy=cnn_model.evaluate(predicted, test_sent, batch_size = batch_size)
print("CNN Accuracy: %.1f%%" %(100.0*cnn_accuracy))
StopWatch.stop("cnn evaluate")
         /usr/local/lib/python3.7/dist-
StopWatch.benchmark()
         /usr/local/lib/python3.7/dist-
        /usr/local/lib/python3.7/dist-
     packages/tensorflow/python/distribute/distribute lib.py:3417 call for each replica
             return fn(*args, **kwargs)
         /usr/local/lib/python3.7/dist-
     packages/tensorflow/python/keras/engine/training.py:788 run step **
             outputs = model.train_step(data)
         /usr/local/lib/python3.7/dist-
     packages/tensorflow/python/keras/engine/training.py:754 train step
             y pred = self(x, training=True)
         /usr/local/lib/python3.7/dist-
     packages/tensorflow/pvthon/keras/engine/base laver.pv:998
```