# term\_term\_pr\_distr\_and\_inference-multinomial-min\_df-3-size-150-max

December 8, 2020

#### autoreload modules and utilities

[1]: %load\_ext autoreload %autoreload 2

# import all necessary libraries/packages

```
import joblib

import numpy as np
import pandas as pd

from tqdm.notebook import tqdm
import matplotlib.pyplot as plt

from scipy.stats import entropy as calculate_entropy

from sklearn.datasets import fetch_20newsgroups
from sklearn.model_selection import StratifiedShuffleSplit

from sklearn.pipeline import Pipeline
from sklearn.pipeline import FeatureUnion
from sklearn.naive_bayes import MultinomialNB, GaussianNB
from sklearn.feature_extraction.text import CountVectorizer, TfidfVectorizer

from sklearn.metrics import classification_report, accuracy_score
from sklearn.metrics import f1_score as calculate_f1_score
from sklearn.model_selection import train_test_split, StratifiedKFold
```

#### **Utility functions**

```
[3]: ## utilities
# from utils import clean_text

import string
from sklearn.base import TransformerMixin
```

```
import nltk
from nltk import word_tokenize
from nltk.stem import WordNetLemmatizer
nltk.download('stopwords')
nltk.download('wordnet')
def clean_text(text: str, lemmatizer = lambda x: x) -> str:
    # removes upper cases
    text = text.lower().strip()
    # removes punctuation
    for char in string.punctuation:
        text = text.replace(char, " ")
    #lematize the words and join back into string text
    text = " ".join([lemmatizer(word) for word in word_tokenize(text)])
    return text
def data_isvalid(text, analyser, min_character_size, max_character_size):
    return min_character_size <= len(analyser(text)) <= max_character_size</pre>
def get_pipeline(vectorizer_type, classifier, use_t2pi, min_df=3,__
→stop_words=None, lemmatizer = lambda x: x):
    vectorizer = CountVectorizer if vectorizer_type == "count" else∟
\hookrightarrowTfidfVectorizer
    models = \Gamma
        ('clean_text', CleanTextTransformer(lemmatizer)),
        ("vectorizers", FeatureUnion([
            ('count_binary', CountVectorizer(stop_words=stop_words,_
 ⇒binary=True, min_df=min_df)),
            ("count", vectorizer(stop_words=stop_words, min_df=min_df))
        ])),
    ]
    if use_t2pi:
        models.append(('t2pi_transformer', T2PITransformer()))
    models.append(('classifier', classifier))
    return Pipeline(models)
```

```
class CleanTextTransformer(TransformerMixin):
   def __init__(self, lemmatizer):
        self._lemmatizer = lemmatizer
   def fit(self, X, y=None, **fit_params):
       return self
   def transform(self, X, y=None, **fit_params):
       return np.vectorize(lambda x: clean_text(x, self._lemmatizer))(X)
   def __str__(self):
       return "CleanTextTransformer()"
   def __repr__(self):
       return self.__str__()
class T2PITransformer(TransformerMixin):
   Ostaticmethod
   def _max_weight(x, pbar, word_word_pr):
       pbar.update(1)
       return (word_word_pr.T * x).max(0)
   Ostaticmethod
   def _sum_weight(x, pbar, word_word_pr):
       pbar.update(1)
       return (word_word_pr.T * x).sum(0)
   def fit(self, X, y=None, **fit_params):
       X = X[:, :int(X.shape[1]/2)].toarray()
       print("creating term-term co-occurence pr matrix")
        terms = np.arange(X.shape[1])
       X = pd.DataFrame(X, columns=terms)
        self.word_word_pr_distr = pd.DataFrame(data=0.0, columns=terms,_
 →index=terms)
       for term in tqdm(terms):
            self.word_word_pr_distr[term] = X[X[term] > 0].sum(0) / X.sum(0)
       return self
   def transform(self, X, y=None, **fit_params):
       X = X[:, int(X.shape[1]/2):].toarray()
        X = pd.DataFrame(X, columns=self.word_word_pr_distr.columns)
```

### 1 Load Data

```
[5]: print(all_docs.data[0])
```

I think that domestication will change behavior to a large degree. Domesticated animals exhibit behaviors not found in the wild. I don't think that they can be viewed as good representatives of the wild animal kingdom, since they have been bred for thousands of years to produce certain behaviors, etc.

#### 1.0.1 Create Dataframe

```
[6]: data = pd.DataFrame(
    data={
        "text":all_docs.data,
        "label":all_docs.target
    }
)
data.head()
```

```
[6]: text label

0 \n\nI think that domestication will change beh... 0

1 \nI don't like this comment about "Typical" th... 3

2 \n<apparently you're not a woman - my husband ... 1

3 While not exactly a service incident, I had a ... 1

4 \n\nI think I can. Largely as a result of effo... 2
```

#### 1.0.2 Label Frequency

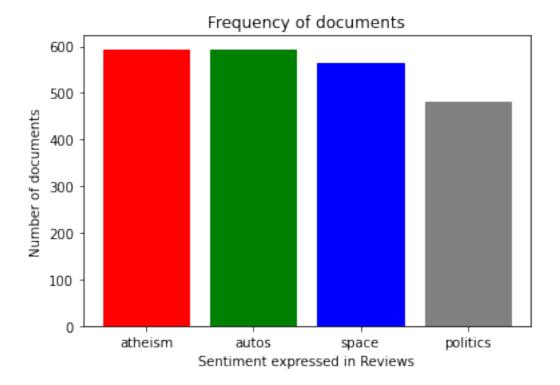
```
[7]: print(data["label"].value_counts())
    print()

barlist = plt.bar(categories, data["label"].value_counts())

plt.title("Frequency of documents")
    plt.xticks(categories, list(map(lambda x: x.split(".")[1], categories)))
    plt.ylabel('Number of documents')
    plt.xlabel('Sentiment expressed in Reviews')

barlist[0].set_color('red')
    barlist[1].set_color('green')
    barlist[2].set_color('blue')
    barlist[3].set_color('grey')
    plt.show()
```

```
1 594
2 593
3 564
0 480
Name: label, dtype: int64
```



The Dataset labels needs to be balanced

### 1.0.3 Parameters

```
[8]: min_df = 3
    stop_words = "english"

def get_classifier():
    # return GaussianNB()
    return MultinomialNB()

def get_lemmatizer():
    # return lambda x: x
    return WordNetLemmatizer().lemmatize
```

# 2 Select Valid Data

```
# make classes balanced
class_indices = []

for index in range(4):
    class_indices.append(np.where((data["label"] == index))[0])

size_per_class = min(max_size_per_class, min(map(len, class_indices)))
indices = np.concatenate([class_ids[:size_per_class] for class_ids in_u class_indices])

data = data.iloc[indices]

data.head()
```

```
[9]: text label

O \n\nI think that domestication will change beh... 0

19 \n\n\tI agree, we spend too much energy on the... 0

30 \n[rest deleted...]\n\nYou were a liberal arts... 0

36 \nWorse? Maybe not, but it is definately a vi... 0

50 \n\n Could you explain what any of the above p... 0

[10]: print(data.iloc[0]["text"])
```

I think that domestication will change behavior to a large degree. Domesticated animals exhibit behaviors not found in the wild. I don't think that they can be viewed as good representatives of the wild animal kingdom, since they have been bred for thousands of years to produce certain behaviors, etc.

```
[11]: print(data["label"].value_counts())
    print()

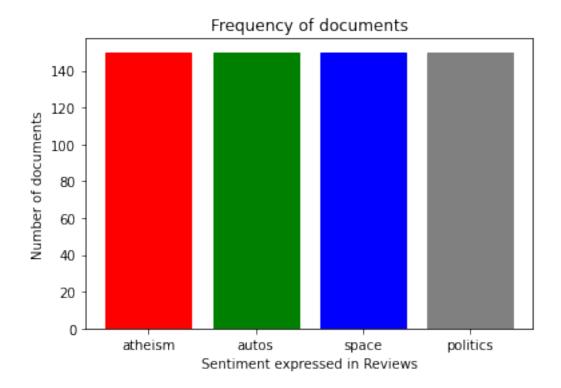
    barlist = plt.bar(categories, data["label"].value_counts())

    plt.title("Frequency of documents")
    plt.xticks(categories, list(map(lambda x: x.split(".")[1], categories)))
    plt.ylabel('Number of documents')
    plt.xlabel('Sentiment expressed in Reviews')

    barlist[0].set_color('red')
    barlist[1].set_color('green')
    barlist[2].set_color('blue')
    barlist[3].set_color('grey')
    plt.show()
```

```
3 150
2 150
1 150
0 150
```

Name: label, dtype: int64



# 2.0.1 initialize input and output

### 2.0.2 initialize recursive word infer model

```
[13]: Pipeline(steps=[('clean_text', CleanTextTransformer()),
                      ('vectorizers',
                       FeatureUnion(transformer_list=[('count_binary',
                                                       CountVectorizer(binary=True,
                                                                        min df=3)),
                                                       ('count',
                                                       CountVectorizer(min df=3))])),
                      ('t2pi_transformer', T2PITransformer()),
                      ('classifier', MultinomialNB())])
[14]: # fit model
      t2pi_model.fit(X_train, y_train)
     creating term-term co-occurence pr matrix
     HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=1136.0),
      →HTML(value='')))
     transforming ...
     HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=450.0),
      →HTML(value='')))
[14]: Pipeline(steps=[('clean_text', CleanTextTransformer()),
                      ('vectorizers',
                       FeatureUnion(transformer_list=[('count_binary',
                                                       CountVectorizer(binary=True,
                                                                        min_df=3)),
                                                       ('count',
                                                       CountVectorizer(min df=3))])),
                      ('t2pi_transformer', T2PITransformer()),
                      ('classifier', MultinomialNB())])
[15]: y_pred = t2pi_model.predict(X_test) #predict testing data
      print(classification_report(y_test, y_pred))
     transforming ...
     HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=150.0),
      →HTML(value='')))
                   precision
                              recall f1-score
                                                    support
                        0.71
                                  0.67
                0
                                             0.69
                                                         45
                        0.76
                1
                                  0.78
                                             0.77
                                                         37
```

2	0.75	0.56	0.64	32
3	0.63	0.81	0.71	36
accuracy			0.71	150
macro avg	0.71	0.70	0.70	150
weighted avg	0.71	0.71	0.70	150

#### 2.0.3 Initialize models

```
[16]: # normal model
     count_model = get_pipeline("count", get_classifier(), use_t2pi=False,__
      →min_df=min_df, stop_words=None, lemmatizer = get_lemmatizer())
     count_sw_model = get_pipeline("count", get_classifier(), use_t2pi=False,_u
      tfidf_model = get_pipeline("tfidf", get_classifier(), use_t2pi=False,__
      →min_df=min_df, stop_words=None, lemmatizer = get_lemmatizer())
     tfidf_sw_model = get_pipeline("tfidf", get_classifier(), use_t2pi=False,_

→min_df=min_df, stop_words=stop_words, lemmatizer = get_lemmatizer())

     # model
     t2pi_count_model = get_pipeline("count", get_classifier(), use_t2pi=True,__
      →min_df=min_df, stop_words=None, lemmatizer = get_lemmatizer())
     t2pi_count_sw_model = get_pipeline("count", get_classifier(), use_t2pi=True,_
      t2pi_tfidf_model = get_pipeline("tfidf", get_classifier(), use_t2pi=True,_u
      →min_df=min_df, stop_words=None, lemmatizer = get_lemmatizer())
     t2pi_tfidf sw_model = get_pipeline("tfidf", get_classifier(), use_t2pi=True,__

_min_df=min_df, stop_words=stop_words, lemmatizer = get_lemmatizer())

     models = {
         "count model": count model,
         "count_sw_model": count_model,
         "tfidf_model": tfidf_model,
         "tfidf_sw_model": tfidf_sw_model,
         "t2pi_count_model": t2pi_count_model,
         "t2pi_count_sw_model": t2pi_count_sw_model,
         "t2pi_tfidf_model": t2pi_tfidf_model,
         "t2pi_tfidf_sw_model": t2pi_tfidf_sw_model
     }
```

## 2.0.4 Running Cross validation on all Models

```
[17]: split size = 7
     skf = StratifiedKFold(n_splits=split_size, shuffle=True, random_state=100)
     macro_f1_scores, weighted_f1_scores, accuracies = [], [], []
     for train_index, test_index in skf.split(X, y):
        index += 1
        x_train_fold, x_test_fold = X.iloc[train_index], X.iloc[test_index]
        y_train_fold, y_test_fold = y.iloc[train_index], y.iloc[test_index]
        accuracies.append([])
        macro_f1_scores.append([])
        weighted_f1_scores.append([])
        for model_name, model in models.items():
            print(f' \rightarrow \{index\}. \{model\_name\} \n{"="*100}\n')
            model.fit(x_train_fold, y_train_fold)
            y_pred = model.predict(x_test_fold)
            accuracy = accuracy_score(y_test_fold, y_pred)
            weighted_f1_score = calculate_f1_score(y_test_fold, y_pred,__
      →average='weighted')
            macro_f1_score = calculate_f1_score(y_test_fold, y_pred,__
      →average='macro')
            weighted_f1_scores[-1].append(weighted_f1_score)
            macro_f1_scores[-1].append(macro_f1_score)
            accuracies[-1].append(accuracy)
    -> 1. count_model
    ______
    -> 1. count_sw_model
    ______
    =============
    -> 1. tfidf model
    -> 1. tfidf_sw_model
    ______
```

```
-> 1. t2pi_count_model
______
creating term-term co-occurence pr matrix
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=1273.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=514.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=86.0),
→HTML(value='')))
-> 1. t2pi_count_sw_model
_______
creating term-term co-occurence pr matrix
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=1059.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=514.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=86.0),
→HTML(value='')))
-> 1. t2pi_tfidf_model
_____
creating term-term co-occurence pr matrix
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=1273.0),
→HTML(value='')))
```

```
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=514.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=86.0),
→HTML(value='')))
-> 1. t2pi_tfidf_sw_model
______
______
creating term-term co-occurence pr matrix
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=1059.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=514.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=86.0),
→HTML(value='')))
-> 2. count model
-> 2. count_sw_model
______
______
-> 2. tfidf_model
_____
-> 2. tfidf_sw_model
______
-> 2. t2pi_count_model
```

```
_____
creating term-term co-occurence pr matrix
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=1275.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=514.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=86.0),
→HTML(value='')))
-> 2. t2pi count sw model
______
______
creating term-term co-occurence pr matrix
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=1056.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=514.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=86.0),
→HTML(value='')))
-> 2. t2pi_tfidf_model
______
creating term-term co-occurence pr matrix
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=1275.0),
→HTML(value='')))
transforming ...
```

```
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=514.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=86.0),
→HTML(value='')))
-> 2. t2pi_tfidf_sw_model
_____
creating term-term co-occurence pr matrix
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=1056.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=514.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=86.0),
→HTML(value='')))
-> 3. count_model
  _____
-> 3. count_sw_model
______
-> 3. tfidf_model
______
-> 3. tfidf_sw_model
______
-> 3. t2pi_count_model
______
===============
```

```
creating term-term co-occurence pr matrix
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=1274.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=514.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=86.0),
→HTML(value='')))
-> 3. t2pi_count_sw_model
______
creating term-term co-occurence pr matrix
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=1057.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=514.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=86.0),
→HTML(value='')))
-> 3. t2pi_tfidf_model
creating term-term co-occurence pr matrix
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=1274.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=514.0),
→HTML(value='')))
```

```
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=86.0),
→HTML(value='')))
-> 3. t2pi_tfidf_sw_model
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_____
creating term-term co-occurence pr matrix
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=1057.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=514.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=86.0),
→HTML(value='')))
-> 4. count_model
______
-> 4. count_sw_model
______
______
-> 4. tfidf model
-> 4. tfidf_sw_model
______
_____
-> 4. t2pi_count_model
______
creating term-term co-occurence pr matrix
```

```
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=1274.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=514.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=86.0),
→HTML(value='')))
-> 4. t2pi_count_sw_model
______
______
creating term-term co-occurence pr matrix
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=1057.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=514.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=86.0),
→HTML(value='')))
-> 4. t2pi_tfidf_model
______
================
creating term-term co-occurence pr matrix
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=1274.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=514.0),
→HTML(value='')))
```

```
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=86.0),
→HTML(value='')))
-> 4. t2pi_tfidf_sw_model
_____
_____
creating term-term co-occurence pr matrix
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=1057.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=514.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=86.0),
→HTML(value='')))
-> 5. count_model
______
-> 5. count_sw_model
______
______
-> 5. tfidf model
-> 5. tfidf_sw_model
______
_____
-> 5. t2pi_count_model
______
creating term-term co-occurence pr matrix
```

```
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=1264.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=514.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=86.0),
→HTML(value='')))
-> 5. t2pi_count_sw_model
______
______
creating term-term co-occurence pr matrix
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=1044.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=514.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=86.0),
→HTML(value='')))
-> 5. t2pi_tfidf_model
______
================
creating term-term co-occurence pr matrix
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=1264.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=514.0),
→HTML(value='')))
```

```
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=86.0),
→HTML(value='')))
-> 5. t2pi_tfidf_sw_model
_____
==============
creating term-term co-occurence pr matrix
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=1044.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=514.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=86.0),
→HTML(value='')))
-> 6. count_model
______
-> 6. count_sw_model
______
______
-> 6. tfidf model
-> 6. tfidf_sw_model
______
_____
-> 6. t2pi_count_model
______
creating term-term co-occurence pr matrix
```

```
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=1292.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=515.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=85.0),
→HTML(value='')))
-> 6. t2pi_count_sw_model
______
______
creating term-term co-occurence pr matrix
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=1073.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=515.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=85.0),
→HTML(value='')))
-> 6. t2pi_tfidf_model
______
================
creating term-term co-occurence pr matrix
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=1292.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=515.0),
→HTML(value='')))
```

```
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=85.0),
→HTML(value='')))
-> 6. t2pi_tfidf_sw_model
_____
==============
creating term-term co-occurence pr matrix
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=1073.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=515.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=85.0),
→HTML(value='')))
-> 7. count_model
______
-> 7. count_sw_model
______
______
-> 7. tfidf model
-> 7. tfidf_sw_model
______
_____
-> 7. t2pi_count_model
______
creating term-term co-occurence pr matrix
```

```
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=1287.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=515.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=85.0),
→HTML(value='')))
-> 7. t2pi_count_sw_model
______
______
creating term-term co-occurence pr matrix
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=1069.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=515.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=85.0),
→HTML(value='')))
-> 7. t2pi_tfidf_model
______
================
creating term-term co-occurence pr matrix
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=1287.0),
→HTML(value='')))
transforming ...
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=515.0),
→HTML(value='')))
```

```
HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=85.0),
      →HTML(value='')))
     -> 7. t2pi_tfidf_sw_model
     ______
     =============
     creating term-term co-occurence pr matrix
     HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=1069.0),
     →HTML(value='')))
     transforming ...
     HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=515.0),
     →HTML(value='')))
     transforming ...
     HBox(children=(HTML(value=''), FloatProgress(value=0.0, max=85.0),
     →HTML(value='')))
[18]: model_names = list(models.keys())
     accuracy = pd.DataFrame(data=np.array(accuracies), columns=model_names)
     weighted_f1_score = pd.DataFrame(data=np.array(weighted_f1_scores),__
      macro_f1_score = pd.DataFrame(data=np.array(macro_f1_scores),__

→columns=model_names)
     accuracy.loc["mean"] = accuracy.mean(0)
     weighted_f1_score.loc["mean"] = weighted_f1_score.mean(0)
     macro_f1_score.loc["mean"] = macro_f1_score.mean(0)
[19]: accuracy.head(split_size+1)
[19]:
           count_model count_sw_model tfidf_model tfidf_sw_model \
                                         0.790698
     0
             0.790698
                            0.790698
                                                        0.813953
     1
             0.848837
                            0.848837
                                         0.790698
                                                        0.813953
     2
             0.767442
                            0.767442
                                         0.720930
                                                        0.779070
     3
             0.755814
                            0.755814
                                         0.744186
                                                        0.744186
     4
             0.639535
                            0.639535
                                        0.651163
                                                        0.755814
             0.705882
                             0.705882
                                         0.717647
                                                        0.729412
```

transforming ...

```
0.743189
                                 0.743189
                                               0.731601
                                                                0.761501
      mean
                                                      t2pi_tfidf_model
            t2pi_count_model
                                t2pi_count_sw_model
      0
                     0.779070
                                            0.837209
                                                               0.790698
      1
                     0.848837
                                            0.860465
                                                               0.837209
      2
                     0.755814
                                                               0.755814
                                            0.767442
      3
                     0.744186
                                            0.732558
                                                               0.732558
      4
                                                               0.662791
                     0.639535
                                            0.732558
      5
                     0.729412
                                            0.729412
                                                               0.729412
      6
                                                               0.694118
                     0.670588
                                            0.694118
                     0.738206
                                            0.764823
                                                               0.743228
      mean
            t2pi_tfidf_sw_model
      0
                        0.825581
      1
                        0.848837
      2
                        0.790698
      3
                        0.720930
      4
                        0.744186
      5
                        0.764706
      6
                        0.705882
                        0.771546
      mean
     weighted_f1_score.head(split_size+1)
[20]:
[20]:
            count model
                          count_sw_model
                                            tfidf model
                                                          tfidf_sw_model
      0
                0.793588
                                 0.793588
                                               0.792005
                                                                0.813405
      1
                0.848491
                                 0.848491
                                               0.791508
                                                                0.813069
      2
                0.765940
                                 0.765940
                                               0.717751
                                                                0.775940
      3
                0.755232
                                 0.755232
                                               0.742546
                                                                0.744384
      4
                0.633116
                                 0.633116
                                               0.644383
                                                                0.751040
      5
                0.703328
                                 0.703328
                                               0.715738
                                                                0.730247
      6
                0.693308
                                 0.693308
                                               0.704881
                                                                0.693997
                0.741857
                                 0.741857
                                               0.729830
                                                                0.760297
      mean
            t2pi_count_model
                                t2pi_count_sw_model
                                                      t2pi_tfidf_model
      0
                     0.780646
                                            0.838026
                                                               0.790735
      1
                     0.848610
                                            0.860938
                                                               0.837316
      2
                     0.752918
                                            0.762584
                                                               0.752351
      3
                     0.744013
                                            0.731442
                                                               0.733177
      4
                     0.633116
                                            0.731213
                                                               0.660776
      5
                     0.727592
                                            0.729392
                                                               0.727805
      6
                     0.669074
                                            0.694041
                                                               0.694385
      mean
                     0.736567
                                            0.763948
                                                               0.742363
            t2pi_tfidf_sw_model
      0
                        0.825307
```

6

0.694118

0.694118

0.705882

0.694118

```
1
                        0.848837
      2
                        0.787803
      3
                        0.717978
      4
                        0.739650
      5
                        0.765708
      6
                        0.705617
                        0.770129
      mean
[21]:
     macro_f1_score.head(split_size+1)
[21]:
            count_model
                          count_sw_model
                                            tfidf_model
                                                         tfidf_sw_model
      0
                0.793105
                                 0.793105
                                               0.791674
                                                                0.813710
      1
                0.848370
                                 0.848370
                                               0.791588
                                                                0.812626
      2
                0.766720
                                 0.766720
                                               0.718401
                                                                0.775478
      3
                0.754968
                                 0.754968
                                               0.742405
                                                                0.743709
      4
                0.632754
                                 0.632754
                                               0.643963
                                                                0.751143
      5
                0.702384
                                 0.702384
                                               0.714735
                                                                0.730082
      6
                0.693096
                                 0.693096
                                               0.704661
                                                                0.693601
                0.741628
                                 0.741628
                                               0.729632
                                                                0.760050
      mean
            t2pi_count_model
                                t2pi_count_sw_model
                                                      t2pi_tfidf_model
      0
                     0.780392
                                            0.838588
                                                               0.791070
      1
                     0.848722
                                            0.860613
                                                               0.837459
      2
                                                               0.752624
                     0.753387
                                            0.762010
      3
                     0.743818
                                            0.730341
                                                               0.733305
      4
                     0.632754
                                            0.731071
                                                               0.660436
      5
                     0.726937
                                            0.729081
                                                               0.726503
      6
                     0.669103
                                            0.693838
                                                               0.694186
                     0.736445
                                            0.763649
                                                               0.742226
      mean
            t2pi_tfidf_sw_model
      0
                        0.825896
      1
                        0.848787
      2
                        0.787701
      3
                        0.716959
      4
                        0.739606
      5
                        0.765829
      6
                        0.705514
                        0.770042
      mean
```