Fake News detection (Vrai and Faux) vs Mix

May 6, 2023

1 Projet final de Machine Learning 1 HAI817I

Ce projet s'inscrit dans le contexte de l'apprentissage supervisé, i.e. les données possèdent des labels. Il vise à trouver les modèles les plus performants pour prédire si des articles de presse sont vrais ou faux. Les articles contiennent des assertions (une assertion est une proposition que l'on avance et que l'on soutient comme vraie) faites, par exemple, par des hommes politiques

```
[75]: # ! pip install langdetect
# !pip install contractions
# ! pip install wordcloud
```

1.1 Importation des bibliotheques et modules necessaires

```
[2]: # Importation des différentes librairies utiles pour le notebook
     #Sickit learn met réqulièrement à jour des versions et
     #indique des futurs warnings.
     #ces deux lignes permettent de ne pas les afficher.
     import warnings
     warnings.filterwarnings("ignore", category=FutureWarning)
     import pandas as pd
     import seaborn as sns
     import matplotlib.pyplot as plt
     import sys
     # librairies générales
     import pickle
     import pandas as pd
     from scipy.stats import randint
     import numpy as np
     import string
     import time
     import base64
     import re
     import sys
     import contractions
```

```
# librairie BeautifulSoup
from bs4 import BeautifulSoup
# librairie affichage
## detection de language
import langdetect
import wordcloud
import nltk
from nltk import sent_tokenize
from nltk import RegexpParser
from nltk import pos_tag
from nltk import word_tokenize
from nltk.tokenize import word_tokenize
from nltk.corpus import stopwords
# Lemmatisateur
from nltk.stem import WordNetLemmatizer
# Racinisateur
from nltk.stem.porter import PorterStemmer
# TF-IDF
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.feature_extraction.text import TfidfTransformer
from sklearn.feature_extraction.text import CountVectorizer
import spacy
from spacy.tokens import Span
from spacy.lang.en import English
from sklearn.base import BaseEstimator
from sklearn.base import TransformerMixin
# Modeles
from sklearn.linear_model import LogisticRegression
from sklearn.svm import SVC
from sklearn.tree import DecisionTreeClassifier
from sklearn.ensemble import RandomForestClassifier
from sklearn.ensemble import GradientBoostingClassifier
from sklearn.model_selection import GridSearchCV
from sklearn.model_selection import train_test_split
```

```
# Metrics
from sklearn.metrics import accuracy_score
from sklearn.metrics import ConfusionMatrixDisplay
from sklearn.metrics import classification_report
from sklearn.model_selection import cross_val_score
from sklearn.metrics import roc_curve
from sklearn.metrics import confusion_matrix
from sklearn.metrics import precision_recall_fscore_support as score
from tabulate import tabulate
# il est possible de charger l'ensemble des librairies en une seule fois
# décocher le commentaire de la ligne ci-dessous
#nltk.download('all')
# nltk.download('punkt')
# nltk.download('averaged_perceptron_tagger')
# nltk.download('tagsets')
# nltk.download("stopwords")
# nltk.download('wordnet')
```

```
[3]: import pandas as pd
     import numpy as np
     import seaborn as sns
     import matplotlib.pyplot as plt
     import sys
     import sklearn
     # import some classifier
     from sklearn.naive bayes import MultinomialNB
     from sklearn.linear_model import LogisticRegression
     from sklearn.naive_bayes import GaussianNB
     from sklearn.tree import DecisionTreeClassifier
     from sklearn.neighbors import KNeighborsClassifier
     from sklearn.ensemble import RandomForestClassifier
     from sklearn.svm import SVC
     from sklearn.linear_model import LinearRegression
     from sklearn.linear_model import SGDClassifier
     from sklearn.linear_model import PassiveAggressiveClassifier
     # import modules for vectorizing and pipe
     from sklearn.feature_extraction.text import CountVectorizer, TfidfVectorizer
     from sklearn.pipeline import Pipeline
     from sklearn import preprocessing
     from sklearn.feature extraction.text import TfidfTransformer
```

```
from sklearn.pipeline import make_pipeline
# scale data
from sklearn.preprocessing import StandardScaler
# upsampling downsampling
from sklearn.utils import resample
# modules for model selection
from sklearn.model_selection import train_test_split
from sklearn.model selection import KFold
from sklearn.model_selection import GridSearchCV
from sklearn.model_selection import cross_val_score
# modules for evaluation
from sklearn.metrics import accuracy_score
from sklearn.metrics import confusion_matrix
from sklearn.metrics import classification_report
from sklearn.metrics import precision_recall_fscore_support as score
# modules for vizualisation
import seaborn as sns
import matplotlib.pyplot as plt
from tabulate import tabulate
# others
import itertools
import random
from sklearn.exceptions import ConvergenceWarning
# import some classifier
from sklearn.naive_bayes import MultinomialNB
from sklearn.linear_model import LogisticRegression
from sklearn.naive_bayes import GaussianNB
from sklearn.tree import DecisionTreeClassifier
from sklearn.neighbors import KNeighborsClassifier
from sklearn.ensemble import RandomForestClassifier
from sklearn.svm import SVC
from sklearn.linear_model import LinearRegression
from sklearn.linear model import SGDClassifier
from sklearn.linear_model import PassiveAggressiveClassifier
from sklearn.pipeline import Pipeline
from sklearn import preprocessing
```

```
from sklearn.pipeline import make_pipeline
 [4]: # Dataset initial
      train_df_init = pd.read_csv('HAI817_Project_data/HAI817_Projet_train.csv')
      test_df_init = pd.read_csv('HAI817_Project_data/HAI817_Projet_test.csv')
      # Dataset recent
      train_df_latest = pd.read_csv('HAI817_Project_data/HAI817_Project_data/
       →HAI817_Projet_train.csv')
      test_df_latest = pd.read_csv('HAI817_Project_data/HAI817_Project_data/
       ⇔HAI817_Projet_train.csv')
      train_df_init = train_df_init.fillna(' ')
      test_df_init = test_df_init.fillna(' ')
 [5]: # Initial
      train_df_init.describe()
      #
      # Recent
      test_df_latest.describe()
 [5]:
             public_id
                                                                      text
      count
                  1264
                                                                      1264 \
      unique
                  1115
                                                                      1086
              cd9cd5e8 The late Robin Williams once called cocaine "G...
      top
      freq
                                                                         4
                              title our rating
      count
                               1241
                                          1264
     unique
                               1070
              - The Washington Post
      top
                                         false
      freq
                                           578
[76]: # Un petit echantillon de 15 lignes pour commencer
      \# sample = train_df_init.sample(15)
[77]: # sample
[74]: # sample.info()
      # sample = sample.values.tolist()
      # sample
 [9]: # sample
```

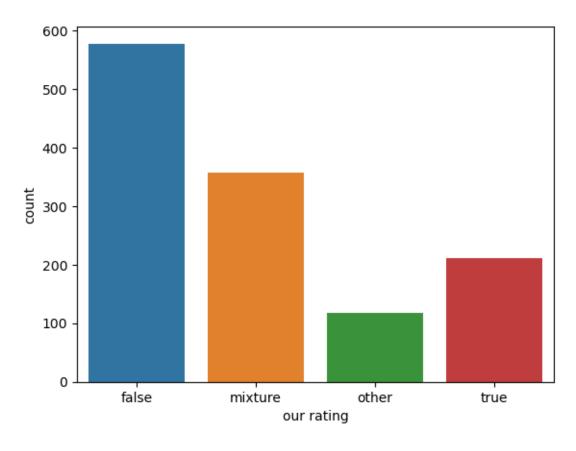
from sklearn.feature_extraction.text import TfidfTransformer

```
[10]: # For the content
      # for record in sample:
            text = record[1]
            title = record[2]
              display(text)
              wc = wordcloud.WordCloud(background_color='black', max_font_size=30,_
       \hookrightarrow max words=100)
              text_wc = wc.generate(str(text))
              title_wc = wc.generate(str(title))
              print("Text size: ",len(text), " Title size: ", len(title), end="\n")
            text_wc = wordcloud.WordCloud(background color='black', max_font_size=30,__
       →max_words=100).generate_from_text(text=text)
            title wc = wordcloud.WordCloud(background color='black',,,
       →max_font_size=30, max_words=100).generate_from_text(text=title)
      # #
              fig = plt.figure(num=1)
            # Create a figure with two subplots
            fig, axs = plt.subplots(ncols=2, figsize=(15, 10)) # figsize=(10, 5)
            fig.suptitle(title)
            # Plot the first wordcloud in the first subplot
            axs[0].imshow(text_wc, cmap=None) # , interpolation='bilinear'
      #
            axs[0].set title('Text')
      #
            # Plot the second wordcloud in the second subplot
            axs[1].imshow(title wc, cmap=None)
            axs[1].set title('Title')
            # Remove axis ticks and labels
      #
            for ax in axs:
      #
                ax.set_xticks([])
      #
                ax.set_yticks([])
                ax.set_xticklabels([])
      #
                ax.set_yticklabels([])
              plt.axis('off')
              plt.imshow(text_wc, title_wc, cmap=None)
            plt.show()
              display(record)
      # #
```

1.2 !Apparemment le texte et titre sont mal positionnes dans certains row et certains rows ont un texte de taille presqu'egale a celle du titre, Est-ce du bruit?

```
[11]: sns.countplot(x='our rating', data=train_df_init)
```

[11]: <AxesSubplot:xlabel='our rating', ylabel='count'>



2 Ingenierie des donnees

2.0.1 Renommons le rating

```
[15]: train_df_init.rating
[15]: 0
                false
      1
              mixture
      2
              mixture
      3
                false
                false
      1259
                 true
      1260
                 true
      1261
                false
      1262
                 true
      1263
                  true
      Name: rating, Length: 1264, dtype: object
```

2.1 Elements utilitaires du cours

TextCleaner

```
[16]: nltk.download('wordnet')
      nltk.download('stopwords')
      nltk.download('punkt')
      stop_words = set(stopwords.words('english'))
      def MyCleanText(X,
                     lowercase=False, # mettre en minuscule
                     removestopwords=False, # supprimer les stopwords
                     removedigit=False, # supprimer les nombres
                     getstemmer=False, # conserver la racine des termes
                     getlemmatisation=False # lematisation des termes
                    ):
          sentence=str(X)
          # suppression des caractères spéciaux
          sentence = re.sub(r'[^\w\s]',' ', sentence)
          # suppression de tous les caractères uniques
          sentence = re.sub(r'\s+[a-zA-Z]\s+', ' ', sentence)
          # substitution des espaces multiples par un seul espace
          sentence = re.sub(r'\s+', ' ', sentence, flags=re.I)
          # decoupage en mots
          tokens = word_tokenize(sentence)
          if lowercase:
                tokens = [token.lower() for token in tokens]
          # suppression ponctuation
          table = str.maketrans('', '', string.punctuation)
```

```
words = [token.translate(table) for token in tokens]
    # suppression des tokens non alphabetique ou numerique
    words = [word for word in words if word.isalnum()]
    # suppression des tokens numerique
    if removedigit:
        words = [word for word in words if not word.isdigit()]
    # suppression des stopwords
    if removestopwords:
        words = [word for word in words if not word in stop_words]
    # lemmatisation
    if getlemmatisation:
        lemmatizer=WordNetLemmatizer()
        words = [lemmatizer.lemmatize(word)for word in words]
    # racinisation
    if getstemmer:
        ps = PorterStemmer()
        words=[ps.stem(word) for word in words]
    sentence= ' '.join(words)
    return sentence
[nltk_data] Downloading package wordnet to /home/richard/nltk_data...
```

```
[nltk_data] Downloading package wordnet to /home/richard/nltk_data...
[nltk_data] Package wordnet is already up-to-date!
[nltk_data] Downloading package stopwords to
[nltk_data] /home/richard/nltk_data...
[nltk_data] Package stopwords is already up-to-date!
[nltk_data] Downloading package punkt to /home/richard/nltk_data...
[nltk_data] Package punkt is already up-to-date!
```

TextNormailizer

```
self.getstemmer=getstemmer
    self.removestopwords=removestopwords
    self.getlemmatisation=getlemmatisation
    self.removedigit=removedigit
def transform(self, X, **transform_params):
    # Nettoyage du texte
    X=X.copy() # pour conserver le fichier d'origine
    return [MyCleanText(text,lowercase=self.lowercase,
                        getstemmer=self.getstemmer,
                        removestopwords=self.removestopwords,
                        getlemmatisation=self.getlemmatisation,
                        removedigit=self.removedigit) for text in X]
def fit(self, X, y=None, **fit_params):
    return self
def fit_transform(self, X, y=None, **fit_params):
    return self.fit(X).transform(X)
def get_params(self, deep=True):
    return {
        'lowercase':self.lowercase,
        'getstemmer':self.getstemmer,
        'removestopwords':self.removestopwords,
        'getlemmatisation':self.getlemmatisation,
        'removedigit':self.removedigit
    }
def set_params (self, **parameters):
    for parameter, value in parameters.items():
        setattr(self,parameter,value)
    return self
```

${\bf True False Mixer}$

```
[18]: def TrueFalseMixer(x):

    if x=="true" or x=="false":
        return "true_or_false"
    else:
        return "mixture"
```

```
[19]: # train_df_init = pd.read_csv('HAI817_Project_data/HAI817_Projet_train.csv')
# train_df_init = train_df_init.rename(columns={"our rating": "rating"})
# test_df_init = pd.read_csv("HAI817_Project_data/HAI817_Projet_test.csv")
# test_df_init = test_df_init.rename(columns={"our rating": "rating"})
```

```
# test_df_init.head()
[20]: | train_df_init = pd.read_csv('HAI817_Project_data/HAI817_Projet_train.csv')
      train_df_init = train_df_init.rename(columns={"our rating": "rating"})
      test_df_init = pd.read_csv("HAI817_Project_data/HAI817_Projet_test.csv")
      test_df_init = test_df_init.rename(columns={"our rating": "rating"})
      train_df_init = train_df_init.fillna(' ')
      test_df_init = test_df_init.fillna(' ')
      train_df_init['text'] = train_df_init['title'] + " "+train_df_init['text']
      test_df_init['text'] = test_df_init['title'] + " "+test_df_init['text']
      test_df_init.head()
[20]:
      0 122653045997905671927713471889615536378 \
      1 275389285957305997321446227088442471741
      2 333248764296609831067233855420575814716
      3 264019763253447756851916399533799891538
      4 158073737187690682830899773280916034317
                                                      text
      O US Treasury deputy sec warns that shortages li... \
      1 CNN Praises Taliban For Wearing Masks During A...
      2 Tennessee Has Just LEGALIZED Government COVID ...
      3 MEDICAL SHOCKER: Scientists at Sloan Kettering...
      4 Study Results: Facemasks are Ineffective to Bl...
                                                     title
                                                             rating
     O US Treasury deputy sec warns that shortages li... mixture
      1 CNN Praises Taliban For Wearing Masks During A...
                                                             other
      2 Tennessee Has Just LEGALIZED Government COVID ...
                                                             false
      3 MEDICAL SHOCKER: Scientists at Sloan Kettering...
                                                            false
      4 Study Results: Facemasks are Ineffective to Bl...
                                                            false
     Resampler
[21]: def Resampler( df, classif_type=0):
          smallest size = 0
          smallest = ""
           # true false other mixture
          df["rating"].value_counts().plot(kind='pie',
                                        autopct='%1.1f%%',
```

label = "Classe",

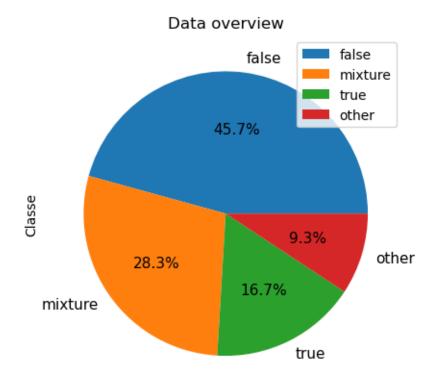
```
title='Data overview',
                                  fontsize=11,
                                  legend=True)
   plt.show()
   if classif_type==0:
       11 11 11
       resampling pour Vrai vs Faux
       df
       falses = df[df["rating"] == "false"]
       trues = df[df["rating"] == "true"]
       # find smallest between trues and falses and dropping other types
       smallest = "true" if trues.shape[0] <= falses.shape[0] else "false"</pre>
       smallest_size = trues.shape[0] if trues.shape[0] <= falses.shape[0]__
→else falses.shape[0]
       df = pd.concat([falses, trues]) # df[df["rating"] == smallest]
       df["rating"].value_counts().plot(kind='pie',
                                      autopct='%1.1f%%',
                                      label = "Classe",
                                      title='Before',
                                      fontsize=11,
                                      legend=True)
       plt.show()
        print("True: ", trues.shape[0]," False: ", falses.shape[0], "Smallest:
→ ", smallest, " Size:", smallest_size)
       df = df[df["rating"] == smallest]
       if smallest == "true":
           df = pd.concat([df, falses.sample(smallest_size)])
       else:
           df = pd.concat([df, trues.sample(smallest_size)]) # df.append(trues.
\hookrightarrow sample(smallest_size))
   if classif_type==1:
       resampling pour (Vrai, Faux) vs mixture
       df["rating"] = df["rating"].apply(TrueFalseMixer)
       df["rating"].value_counts().plot(kind='pie',
                                      autopct='%1.1f%%',
                                      label = "Classe",
                                      title='Before',
```

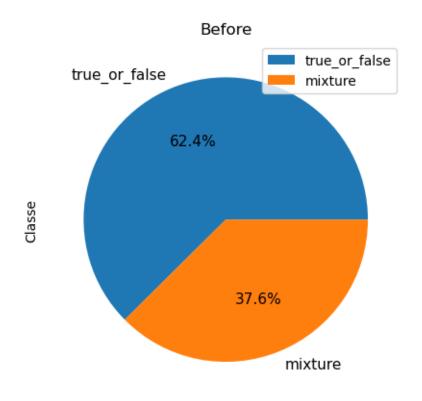
```
fontsize=11,
                                      legend=True)
       plt.show()
       true_and_false = df[df["rating"] == "true_or_false"]
       mixtures = df[df["rating"] == "mixture"]
       # find smallest between trues and falses and dropping other types
       smallest = "true_or_false" if true_and_false.shape[0] <= mixtures.</pre>
⇔shape[0]
            else "mixture"
       smallest_size = true_and_false.shape[0] if true_and_false.shape[0] <=_u
mixtures.shape[0] else mixtures.shape[0]
         print("True fs: ", true_and_false.shape[0]," Mix: ", mixtures.
⇔shape[0], "Smallest: ", smallest, " Size:", smallest_size)
         display(df)
       df = df[df["rating"] == smallest]
       if "true_or_false" in smallest:
           df = pd.concat([df, mixtures.sample(smallest_size)])
       else:
           df = pd.concat([df, true_and_false.sample(smallest_size)])
   if classif_type==2:
       11 11 11
       resampling pour Vrai vs Faux
       # true false other mixture
       falses = df[df["rating"] == "false"]
       trues = df[df["rating"] == "true"]
       others = df[df["rating"] == "other"]
       mixtures = df[df["rating"] == "mixture"]
       classes_dict = {
           "true": trues.shape[0],
           "false": falses.shape[0],
            "other": others.shape[0],
           "mixture": mixtures.shape[0],
       }
       # find smallest between trues and falses and dropping other types
       smallest_size = list(classes_dict.values())[0]
```

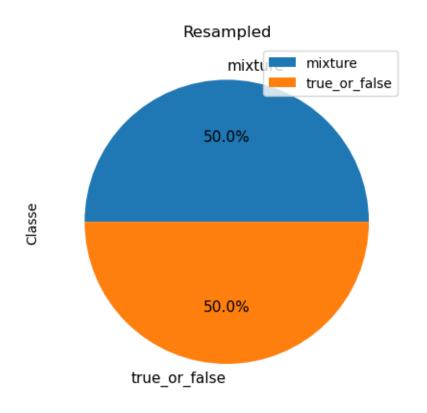
```
smallest = list(classes_dict.keys())[0]
    for key, val in classes_dict.items():
        if val <= smallest_size:</pre>
            smallest_size = val
            smallest = key
    df = df[df["rating"] == smallest]
    if smallest == "true":
        df = pd.concat([df,
                        falses.sample(smallest_size),
                        others.sample(smallest_size),
                        mixtures.sample(smallest_size)])
    elif smallest == "false":
        df = pd.concat([df,
                        true.sample(smallest_size),
                         others.sample(smallest_size),
                        mixtures.sample(smallest_size)])
    elif smallest == "other":
        df = pd.concat([df,
                        trues.sample(smallest_size),
                        falses.sample(smallest_size),
                        mixtures.sample(smallest_size)])
    else:
        df = pd.concat([df,
                        trues.sample(smallest_size),
                        falses.sample(smallest_size),
                         others.sample(smallest_size)])
# Shuffle the df to make data random
df = df.sample(frac=1)
df["rating"].value_counts().plot(kind='pie',
                               autopct='%1.1f%%',
                               label = "Classe",
                              title='Resampled',
                              fontsize=11,
                              legend=True)
plt.show()
return df
```

2.1.1 DataCleanerForClassification

```
[23]: dft = Resampler(df=train_df_init, classif_type=1)
    display(dft)
```







```
907
           863897d7
                     Dr. Didier Raoult published new results of 1,0... \
                     SUPREME COURT OF THE UNITED STATES No. 18A335 ...
     918
           054fefab
     384
           6f6addcc Marine Corps. Rebukes Pelosi: "WE DON'T WORK F...
     1155 claf4bf0 When Will The Planet Be Too Hot For Humans? Mu...
     1245
           d0b0e459
                     Global Ocean Circulation Appears To Be Collaps...
                     SAD: Biden Introduces Granddaughter by Saying,...
     32
           a329a5b3
     794
           f353ec98
                     Project G-2101: Pentagon biolab discovered MER...
     3
           f14e8eb6
                     Obama's Daughters Caught on Camera Burning US ...
     352
           6612a10d
                     Deliveroo drivers demand union recognition and...
     465
           OaO6O4c8 FG bans alcohol in satchets, polythene Covid-1...
                                                         title
                                                                       rating
     907
           Dr. Didier Raoult published new results of 1,0...
                                                                    mixture
           SUPREME COURT OF THE UNITED STATES No. 18A335 ...
     918
                                                                    mixture
     384
           Marine Corps. Rebukes Pelosi: "WE DON'T WORK F... true_or_false
     1155 When Will The Planet Be Too Hot For Humans? Mu...
                                                                    mixture
     1245
           Global Ocean Circulation Appears To Be Collaps... true_or_false
     32
           SAD: Biden Introduces Granddaughter by Saying,...
                                                                    mixture
     794
           Project G-2101: Pentagon biolab discovered MER...
                                                                    mixture
           Obama's Daughters Caught on Camera Burning US ... true_or_false
     352
           Deliveroo drivers demand union recognition and...
                                                                    mixture
     465
                       FG bans alcohol in satchets, polythene
                                                                      mixture
     [950 rows x 4 columns]
[24]: def preprocess_selection(model_name,model,X,y,lowercase=False):
          CV_brut = Pipeline([('cleaner', TextNormalizer()),
                               ('count_vectorizer', CountVectorizer(lowercase=False)),
                               (model name, model)])
          CV_lowcase = Pipeline([('cleaner', __
       →TextNormalizer(removestopwords=False,lowercase=True,
       ⇒getstemmer=False,removedigit=False)),
                               ('count_vectorizer', u

→CountVectorizer(lowercase=lowercase)),
                               (model_name, model)])
          CV_lowStop = Pipeline([('cleaner',__
       →TextNormalizer(removestopwords=True,lowercase=True,
       ⇒getstemmer=False,removedigit=False)),
```

text

public_id

```
('count_vectorizer',_
→CountVectorizer(lowercase=lowercase)),
                       (model_name, model)])
  CV_lowStopstem = Pipeline([('cleaner', __
→TextNormalizer(removestopwords=True,lowercase=True,
⇒getstemmer=True,removedigit=False)),
                       ('count_vectorizer', _
→CountVectorizer(lowercase=lowercase)),
                       (model_name, model)])
  CV_lowStopna = Pipeline([('cleaner', __
→TextNormalizer(removestopwords=True,lowercase=True,
→getstemmer=True,removedigit=True)),
                       ('count_vectorizer',_
→CountVectorizer(lowercase=lowercase)),
                       (model name, model)])
  TFIDF_brut = Pipeline ([('cleaner', TextNormalizer()),
                       ('tfidf_vectorizer',_
→TfidfVectorizer(lowercase=lowercase)),
                       (model_name, model)])
  TFIDF_lowcase = Pipeline([('cleaner', __
→TextNormalizer(removestopwords=False,lowercase=True,
→getstemmer=False,removedigit=False)),
                       ('tfidf_vectorizer', __
→TfidfVectorizer(lowercase=lowercase)),
                       (model_name, model)])
  TFIDF_lowStop = Pipeline([('cleaner', __
→TextNormalizer(removestopwords=True,lowercase=True,
⇒getstemmer=False,removedigit=False)),
                       ('tfidf_vectorizer',u
→TfidfVectorizer(lowercase=lowercase)),
                       (model_name, model)])
  TFIDF_lowStopstem = Pipeline([('cleaner', __
→TextNormalizer(removestopwords=True,lowercase=True,
('tfidf_vectorizer', u
→TfidfVectorizer(lowercase=lowercase)),
```

```
(model_name, model)])
         →TextNormalizer(removestopwords=True,lowercase=True,

→getstemmer=True,removedigit=True)),
                              ('tfidf_vectorizer',__
       →TfidfVectorizer(lowercase=lowercase)),
                              (model name, model)])
         all models = [
              ("CV_brut", CV_brut),
              ("CV_lowcase", CV_lowcase),
              ("CV_lowStop", CV_lowStop),
              ("CV_lowStopstem", CV_lowStopstem),
              ("CV_lowStopna", CV_lowStopna),
              ("TFIDF_lowStopna", TFIDF_lowStopna),
              ("TFIDF_lowcase", TFIDF_lowcase),
              ("TFIDF_lowStop", TFIDF_lowStop),
              ("TFIDF_lowStopstem", TFIDF_lowStopstem),
              ("TFIDF_brut", TFIDF_brut),
         ]
         print ("Evaluation des différentes configurations : ")
         unsorted_scores = [(name, cross_val_score(model, X, y, cv=10).mean()) for__
       →name, model in all_models]
          scores = sorted(unsorted_scores, key=lambda x: -x[1])
         print(tabulate(scores, floatfmt='.4f', headers=('Pipeline', 'Score')))
[25]: X = dft['title']+" "+dft['text']
     \# X = X.drop("rating", axis=1)
     # display(X)
     y = dft['rating']
     у
[25]: 907
                   mixture
     918
                   mixture
     384
             true_or_false
     1155
                   mixture
     1245
             true_or_false
     32
                   mixture
     794
                   mixture
     3
             true_or_false
     352
                   mixture
```

```
465
                    mixture
      Name: rating, Length: 950, dtype: object
[26]: X_s = X # X.sample(200)
      y_s = y.loc[X_s.index]
      preprocess_selection("multinomial_nb", MultinomialNB(), X_s, y_s)
     Evaluation des différentes configurations :
     Pipeline
                           Score
     TFIDF_brut
                          0.7105
     TFIDF_lowStop
                          0.7084
     CV_brut
                          0.7042
     TFIDF_lowcase
                          0.7011
     CV lowcase
                          0.6947
     CV_lowStop
                          0.6937
     TFIDF_lowStopna
                          0.6895
     TFIDF_lowStopstem
                          0.6895
     CV_lowStopstem
                          0.6853
     CV_lowStopna
                          0.6853
 []:
```

Matrice de confusion

```
[27]: def plot_confusion_matrix(cm, classes=[],
                                normalize=False,
                                 title='Confusion matrix',
                                 cmap=plt.cm.Blues):
          HHHH
          See full source and example:
          http://scikit-learn.org/stable/auto_examples/model_selection/
       \neg plot\_confusion\_matrix.html
          This function prints and plots the confusion matrix.
          Normalization can be applied by setting `normalize=True`.
          plt.imshow(cm, interpolation='nearest', cmap=cmap)
          plt.title(title)
          plt.colorbar()
          tick_marks = np.arange(len(classes))
          plt.xticks(tick_marks, classes, rotation=45)
          plt.yticks(tick_marks, classes)
          if normalize:
              cm = cm.astype('float') / cm.sum(axis=1)[:, np.newaxis]
              print("Normalized confusion matrix")
```

```
else:
       print('Confusion matrix, without normalization')
   thresh = 2*cm.max() / 3.
   for i, j in itertools.product(range(cm.shape[0]), range(cm.shape[1])):
       plt.text(j, i, cm[i, j],
                 horizontalalignment="center",
                 color="white" if cm[i, j] > thresh else "black")
   plt.tight_layout()
   plt.ylabel('Predicted label')
   plt.xlabel('True label')
def encode_true_and_false(val):
   if val == "true_or_false":
       return 1
    else:
       return 0
def plot_quad_error(X,y,model):
   arr1=[];arr2=[]
      print("X: ", X.shape," y: ", y.shape)
   size = np.linspace(200 ,int(len(X)), 25).astype('int32') # if X.shape[0] >=__
 →200 else X.shape[0]
   for m in size:
       X_s = X.sample(m)
        y s = y.loc[X s.index.intersection(y.index)] # .intersection(y.index)
       X_train,X_val,y_train,y_val = train_test_split(X_s,y_s,train_size=0.

¬7,random_state=42)

       clf = model.fit(X_train, y_train)
        y_pred_train = clf.predict(X_train)
          y pred train = y pred.shape[0]
       y_pred_val = clf.predict(X_val)
          y_pred_val = y_pred.shape[0]
       y_pred_train=np.vectorize(encode_true_and_false)(y_pred_train)
        y_pred_val=np.vectorize(encode_true_and_false)(y_pred_val)
        y_train=np.vectorize(encode_true_and_false)(y_train)
        y_val=np.vectorize(encode_true_and_false)(y_val)
          print("y pred train: ", y pred train, "y train: ", y train, sep='\n')
        squarred_error_train = (1/len(y_pred_train))*np.sum((y_pred_train -_

y_train)**2)

        squared_error_CV = (1/len(y_pred_val))*np.sum((y_pred_val - y_val)**2)
        arr1.append(squarred_error_train)
        arr2.append(squared_error_CV)
```

```
plt.plot(size,arr1,label='train error')
          plt.plot(size,arr2,label='cv error')
          plt.xlabel("Training Size")
          plt.gca().set_xlim([100,len(X)])
          plt.ylabel("Quadratic Error")
          plt.title("Model Evaluation")
          plt.legend()
          plt.show()
                print("X_s.index: ", X_s.index, " y_index: ", y_s.index, "X.index:", ____
[28]: #
       \hookrightarrow X.index, "y.index:", y.index, sep='\n')
                y s = y.sample(m)
      #
                X_s = X.loc[y_s.index] # .intersection(X.index)
                print("X_s: ", X_s.shape," y_s: ", y_s.shape)
[29]: dft.head()
[29]:
           public_id
                                                                     text
      907
            863897d7 Dr. Didier Raoult published new results of 1,0... \
            054fefab SUPREME COURT OF THE UNITED STATES No. 18A335 ...
      918
      384
            6f6addcc Marine Corps. Rebukes Pelosi: "WE DON'T WORK F...
      1155 c1af4bf0 When Will The Planet Be Too Hot For Humans? Mu...
      1245 dObOe459 Global Ocean Circulation Appears To Be Collaps...
                                                         title
                                                                       rating
      907
            Dr. Didier Raoult published new results of 1,0...
                                                                    mixture
            SUPREME COURT OF THE UNITED STATES No. 18A335 ...
      918
                                                                    mixture
      384
            Marine Corps. Rebukes Pelosi: "WE DON'T WORK F... true_or_false
      1155 When Will The Planet Be Too Hot For Humans? Mu...
      1245 Global Ocean Circulation Appears To Be Collaps... true_or_false
[30]: X = dft["text"]
      y = dft["rating"]
      text normalizer=
      -TextNormalizer(removestopwords=False,lowercase=True,getstemmer=False,removedigit=False)
      X=text_normalizer.fit_transform(X)
      vectorizer = TfidfVectorizer()
      X = vectorizer.fit_transform(X)
[31]: \# print(X, y)
[32]: from sklearn.decomposition import TruncatedSVD
      from sklearn.preprocessing import LabelEncoder
      svd = TruncatedSVD(n_components=2)
      X_svd = svd.fit_transform(X)
```

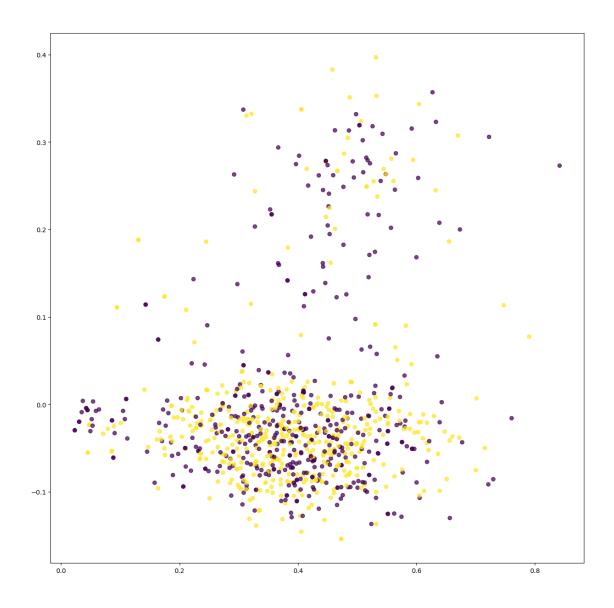
```
# Encode the classes' label as integers
le = LabelEncoder()
y_encoded = le.fit_transform(y)

print("variance ratio: ",svd.explained_variance_ratio_)

print("variance ratio sum: ",svd.explained_variance_ratio_.sum())
print("singular values:" ,svd.singular_values_)

fig,ax = plt.subplots(figsize=(15,15))
ax.scatter(X_svd[:,0], X_svd[:,1], alpha=0.72, c=y_encoded)
plt.show()
```

variance ratio: [0.02051781 0.01229229]
variance ratio sum: 0.0328100962242348
singular values: [12.72291038 3.14966499]

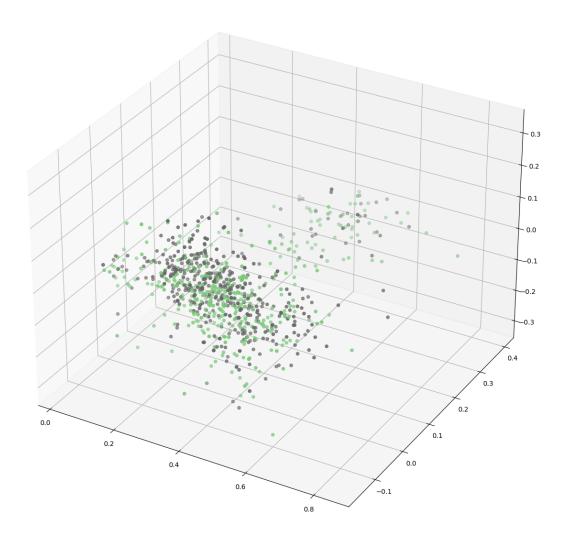


```
[33]: from mpl_toolkits.mplot3d import Axes3D

# Encode the classes' label as integers
le = LabelEncoder()
y_encoded = le.fit_transform(y)

svd = TruncatedSVD(n_components=3)
X_svd = svd.fit_transform(X)

fig = plt.figure(figsize=(15, 15))
ax = fig.add_subplot(projection='3d')
```



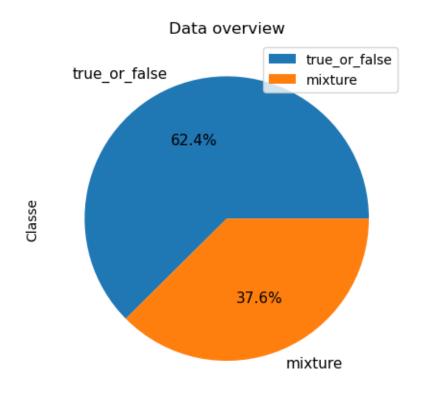
3 Classifications

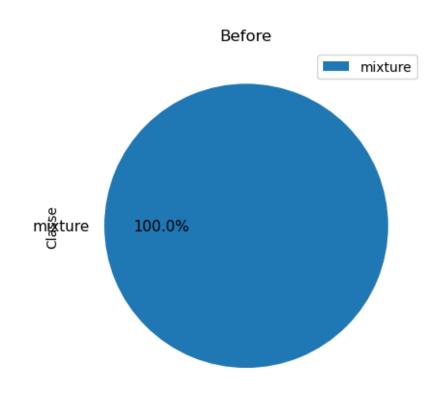
3.1 Classification Vrai vs Faux

3.1.1 Logistic Regression

[]:

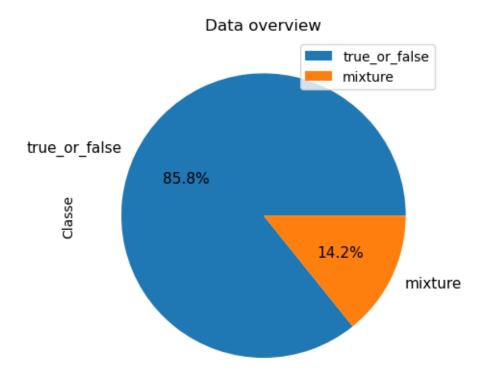
```
[34]: df_train = Resampler(train_df_init, classif_type=1)
      df_test = DataCleanerForClassification(test_df_init, classif_type=1)
      text = df_train['text']
      rating = df_train['rating']
      arr = [i for i in range(len(df_train) + len(df_test))]
      # print(arr, text)
      idx = np.random.choice(arr, 250) # int( len(df_train) / 2)
      # print(idx)
      text = np.concatenate((df_train["text"],df_test["text"]))
      text = text[idx]
      rating = np.concatenate(([df_train['rating'], df_test['rating']]))
      rating = rating[idx]
      # pd.DataFrame(rating).head()
      text_normalizer=TextNormalizer(lowercase=True)
      text=text_normalizer.fit_transform(text)
      text = np.array(text)
      # print(text)
      vectorizer = TfidfVectorizer()
      vectors = vectorizer.fit transform(text)
      corpus = np.array(vectorizer.get_feature_names())
      print(corpus)
```





Resampled

Classe



```
['00' '000' '0008' ... 'óbito' 'g' ' '[
[35]: def get_fake_no_fake_text(text):
          fake_text = np.array(text[rating == 'mixture'])
          non_fake_text = np.array(text[rating == 'true_and_false'])
          return fake_text,non_fake_text
      def compute_frequency(fake_text,non_fake_text):
        # Computes 2 different frequency use of dictionary for O(1) time acces to
       →value of a paticular word
          fake_text = " ".join(fake_text)
          fake_text = word_tokenize(fake_text)
          non_fake_text = " ".join(non_fake_text)
          non_fake_text = word_tokenize(non_fake_text)
          corpus = np.unique(np.concatenate((fake_text,non_fake_text),axis=0))
          FakeFreq = {w:0 for w in corpus}
          NonFakeFreq = {w:0 for w in corpus}
          for word in fake_text:
```

```
FakeFreq[word] += 1
          for word in non_fake_text:
              NonFakeFreq[word] += 1
          return FakeFreq, NonFakeFreq
[36]: fake_text,non_fake_text= get_fake_no_fake_text(text)
      FakeFreq,NonFakeFreq = compute_frequency(fake_text,non_fake_text)
[37]: def feat_extraction_fake(row):
          row_text = word_tokenize(row)
          sum_fake_freq = 0
          for word in row_text:
              if word in FakeFreq.keys():
                    print("word: ", word, " FakeFreq[word]:", FakeFreq[word])
                  sum fake freq += FakeFreq[word]
          return sum_fake_freq
      def feat_extraction_no_fake(row):
          row_text = word_tokenize(row)
          sum_non_fake_freq = 0
          for word in row_text:
              if word in NonFakeFreq.keys():
                  sum_non_fake_freq += NonFakeFreq[word]
          return sum_non_fake_freq
[38]: def conf_matrix(model):
          ConfusionMatrixDisplay.from_estimator(
              model,
              X_test,
              y_test
          )
      def class_report(model):
          print(classification_report(
              Y_test,
              model.predict(X_test)
          ))
[39]: # text.shape
[40]: dfbis train = pd.DataFrame(text,columns=["text"])
```

 $\# dfbis_train["bias"] = 1$

Training

```
[41]: # ps = PorterStemmer()
# def stemming(content):
# stemmed_content = re.sub('[^a-zA-Z]',' ',content)
# stemmed_content = stemmed_content.lower()
# stemmed_content = stemmed_content.split()
# stemmed_content = [ps.stem(word) for word in stemmed_content if not word_u
in stopwords.words('english')]
# stemmed_content = ' '.join(stemmed_content)
# return stemmed_content
```

```
[42]: # # dfbis_train.sample(2)
    # dfbis_train["text"] = dfbis_train["text"].apply(stemming)

# X = dfbis_train["text"]
    # y = rating

# vector = TfidfVectorizer()
    # vector.fit(X)
    # X = vector.transform(X)

# # X.sample(2)
    # print(X)
```

```
clf = LogisticRegression(random_state=42).fit(X_train, y_train)

y_pred = clf.predict(X_test)

target_names = ["mixture","true_or_false"]

print('accuracy %s' % accuracy_score(y_pred, y_test))
print(classification_report(y_test, y_pred)) # target_names=target_names
```

accuracy 0.89333333333333333

	precision	recall	f1-score	support
mixture	0.00	0.00	0.00	8
true_or_false	0.89	1.00	0.94	67
accuracy			0.89	75
macro avg	0.45	0.50	0.47	75
weighted avg	0.80	0.89	0.84	75

/home/richard/anaconda3/lib/python3.9/site-

packages/sklearn/utils/validation.py:993: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n_samples,), for example using ravel().

y = column_or_1d(y, warn=True)

/home/richard/anaconda3/lib/python3.9/site-

packages/sklearn/metrics/_classification.py:1318: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.

_warn_prf(average, modifier, msg_start, len(result))

/home/richard/anaconda3/lib/python3.9/site-

packages/sklearn/metrics/_classification.py:1318: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.

_warn_prf(average, modifier, msg_start, len(result))

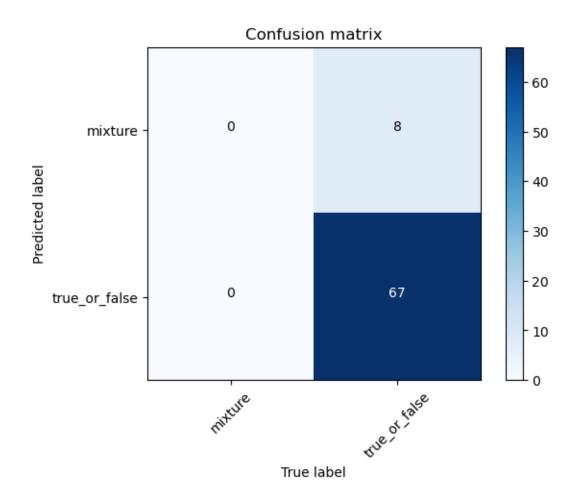
/home/richard/anaconda3/lib/python3.9/site-

packages/sklearn/metrics/_classification.py:1318: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.

_warn_prf(average, modifier, msg_start, len(result))

```
[44]: plot_confusion_matrix(confusion_matrix(y_test, y_pred), classes=target_names) # confusion_matrix(y_test, y_pred) # conf_matrix(clf)
```

Confusion matrix, without normalization



3.1.2 Naives Bayes

```
[45]: train_df_init = pd.read_csv('HAI817_Project_data/HAI817_Projet_train.csv')
    train_df_init = train_df_init.rename(columns={"our rating": "rating"})

test_df_init = pd.read_csv("HAI817_Project_data/HAI817_Projet_test.csv")
    test_df_init = test_df_init.rename(columns={"our rating": "rating"})

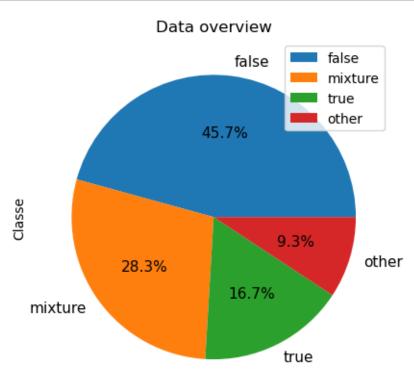
train_df_init = train_df_init.fillna(' ')

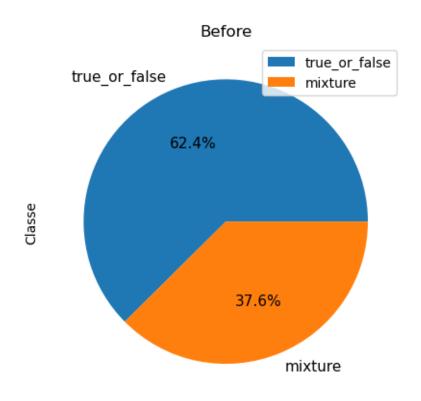
test_df_init = test_df_init.fillna(' ')

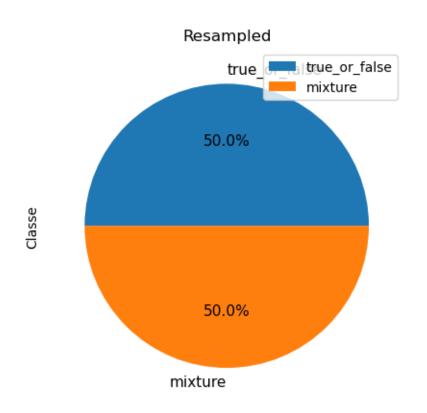
train_df_init['text'] = train_df_init['title'] + " "+train_df_init['text']

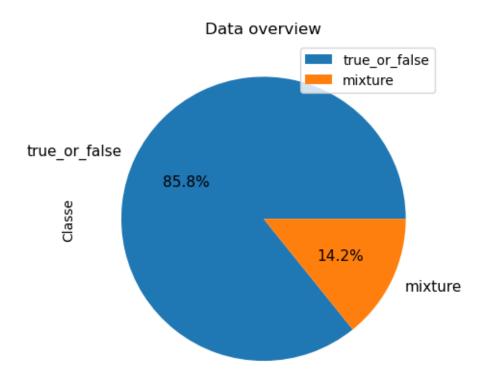
test_df_init['text'] = test_df_init['title'] + " "+test_df_init['text']

df_train = Resampler(train_df_init, classif_type=1)
    df_test = DataCleanerForClassification(test_df_init, classif_type=1)
```









858 467	Up to a third of millennials 'face renting the Khrushchev tours America: 'No sour cabbage sou
273	New Mexico House Committee Passes Bill to Lega
561	Rand Paul: We Must Demilitarize the Police The
222	Pro-Life Groups Upset UN Coronavirus Relief Fu
	•••
167	Bibles Pulled From Shelves For Outdated Idea T
722	A very fresh look at climate change Milwaukee
1202	Australia's Great Barrier Reef has worst coral
232	Langevin, Cicilline Call for Stronger Backgrou
1158	The three-degree world: cities that will be dr
Name:	text, Length: 950, dtype: object

```
[46]: df_train
```

```
[46]: public_id text

858 6600699c Up to a third of millennials 'face renting the... \

467 45525605 Khrushchev tours America: 'No sour cabbage sou...

273 5dc31757 New Mexico House Committee Passes Bill to Lega...

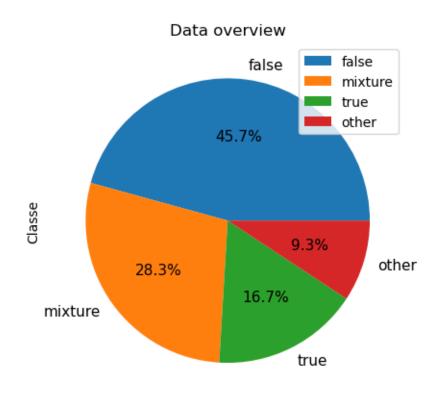
561 0192a5cc Rand Paul: We Must Demilitarize the Police The...
```

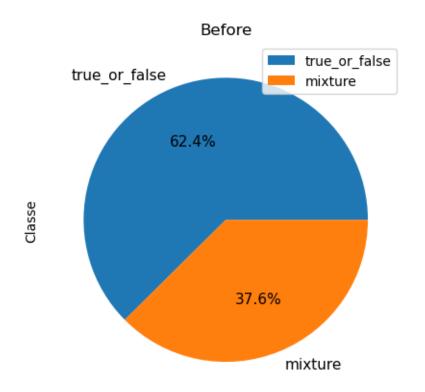
```
222
            b906fccf Pro-Life Groups Upset UN Coronavirus Relief Fu...
      167
            6c95fb76 Bibles Pulled From Shelves For Outdated Idea T...
            95020291 A very fresh look at climate change Milwaukee ...
      722
      1202 Oa68da63 Australia's Great Barrier Reef has worst coral...
      232
            bfd9ddf6 Langevin, Cicilline Call for Stronger Backgrou...
      1158 93db83ce The three-degree world: cities that will be dr...
                                                         title
      858
            Up to a third of millennials 'face renting the... true_or_false
      467
            Khrushchev tours America: 'No sour cabbage sou...
      273
            New Mexico House Committee Passes Bill to Lega... true_or_false
                   Rand Paul: We Must Demilitarize the Police true_or_false
      561
      222
            Pro-Life Groups Upset UN Coronavirus Relief Fu... true_or_false
      167
            Bibles Pulled From Shelves For Outdated Idea T... true_or_false
      722
                          A very fresh look at climate change true_or_false
      1202 Australia's Great Barrier Reef has worst coral... true_or_false
      232
            Langevin, Cicilline Call for Stronger Backgrou...
      1158
           The three-degree world: cities that will be dr...
                                                                    mixture
      [950 rows x 4 columns]
[73]: \# print(X)
[48]: | X_train, X_val, y_train, y_val = train_test_split(X, y, train_size=0.
       →7,random_state=42)
      nb = Pipeline([('vect', TfidfVectorizer()),
                     ('tfidf', TfidfTransformer()),
                     ('clf', MultinomialNB()),
      nb.fit(X_train, y_train)
      y_pred = nb.predict(X_val)
      print('accuracy %s' % accuracy_score(y_pred, y_val))
      print(classification_report(y_val, y_pred))
     accuracy 0.6736842105263158
                    precision
                                  recall f1-score
                                                     support
                          0.63
                                    0.80
                                              0.70
           mixture
                                                          139
     true_or_false
                          0.74
                                    0.55
                                              0.64
                                                          146
                                              0.67
          accuracy
                                                         285
                                    0.68
                                              0.67
                                                         285
                          0.69
         macro avg
```

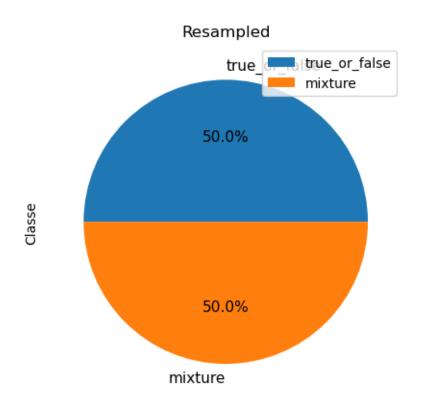
weighted avg 0.69 0.67 0.67 285

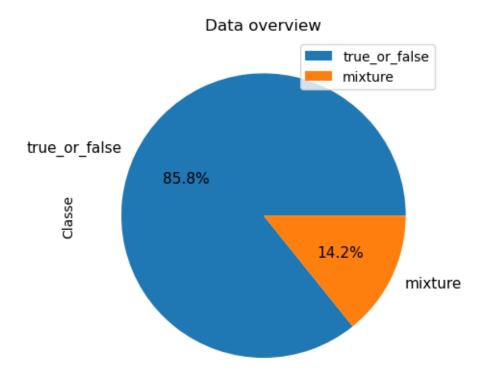
3.1.3 La Cross Validation

```
[49]: train_df_init = pd.read_csv('HAI817_Project_data/HAI817_Projet_train.csv')
      train_df_init = train_df_init.rename(columns={"our rating": "rating"})
      test_df_init = pd.read_csv("HAI817_Project_data/HAI817_Projet_test.csv")
      test_df_init = test_df_init.rename(columns={"our rating": "rating"})
      train_df_init = train_df_init.fillna(' ')
      test_df_init = test_df_init.fillna(' ')
      train_df_init['text'] = train_df_init['title'] + " "+train_df_init['text']
      test_df_init['text'] = test_df_init['title'] + " "+test_df_init['text']
      df_train = Resampler(train_df_init, classif_type=1)
      df_test = DataCleanerForClassification(test_df_init, classif_type=1)
      X = df train["text"]
      y = df_train["rating"]
      # found when testing different parameters in the preprocessing
      text_normalizer=_
       →TextNormalizer(removestopwords=False,lowercase=True,getstemmer=False,removedigit=False)
      X=text_normalizer.fit_transform(X)
      models = []
      models.append(('LRegression', LogisticRegression()))
      models.append(('KNN', KNeighborsClassifier()))
      models.append(('MultinomialNB', MultinomialNB()))
     models.append(('DecisionTreeClassifier', DecisionTreeClassifier()))
      models.append(('SVM', SVC()))
      models.append(('SGDClassifier',SGDClassifier()))
      models.append(('RandomForest', RandomForestClassifier()))
```





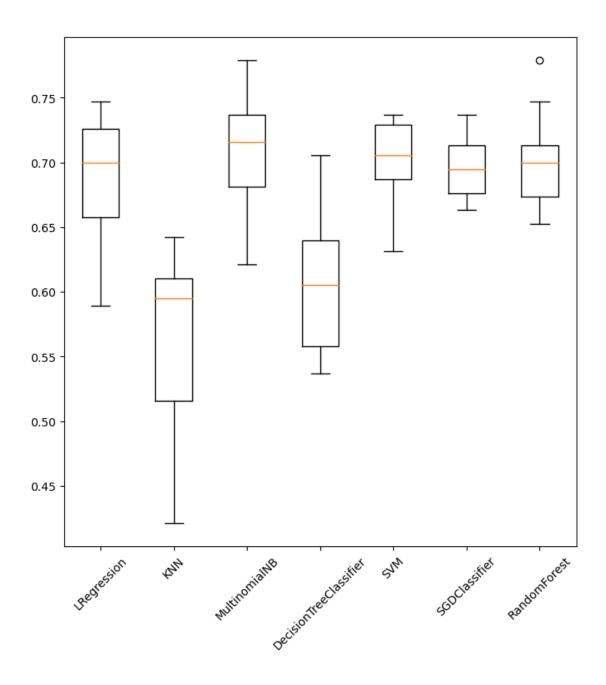




```
names = []
      scoring = 'accuracy'
      for name, model in models:
       kfold = KFold(n_splits=10, shuffle=True)
       model = make_pipeline(TfidfVectorizer(), model)
        cv_results = cross_val_score(model, X, y, cv=kfold, scoring=scoring)
       results.append(cv_results)
       names.append(name)
       msg = "%s: %f (%f)" % (name, cv_results.mean(), cv_results.std())
        print(msg)
     LRegression: 0.689474 (0.047834)
     KNN: 0.564211 (0.072807)
     MultinomialNB: 0.706316 (0.047880)
     DecisionTreeClassifier: 0.602105 (0.052166)
     SVM: 0.702105 (0.030526)
     SGDClassifier: 0.695789 (0.024211)
     RandomForest: 0.700000 (0.038316)
[51]: classes = ["true_or_false", "mixture"]
      fig = plt.figure(figsize=(8,8))
```

[50]: results = []

Comparison of models true_or_false vs mixture



3.1.4 Hyperparameters finding process

3.1.5 Grid Search CV

3.1.6 Logistic Regression

```
[52]: X = df_train["text"]
      y = df_train["rating"]
      X s = X \# X.sample(300) \# X if X.shape[0] <= 700 else X.sample(700)
      y_s = y.loc[X_s.index]
      preprocess_selection("logistic_regression", LogisticRegression(
            n_jobs=-1
      ),X_s,y_s)
     Evaluation des différentes configurations :
     /home/richard/anaconda3/lib/python3.9/site-
     packages/sklearn/linear_model/_logistic.py:814: ConvergenceWarning: lbfgs failed
     to converge (status=1):
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STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
Increase the number of iterations (max_iter) or scale the data as shown in:
   https://scikit-learn.org/stable/modules/preprocessing.html
Please also refer to the documentation for alternative solver options:
   https://scikit-learn.org/stable/modules/linear_model.html#logistic-
regression
 n_iter_i = _check_optimize_result(
/home/richard/anaconda3/lib/python3.9/site-
packages/sklearn/linear_model/_logistic.py:814: ConvergenceWarning: lbfgs failed
to converge (status=1):
STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
```

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regression
 n_iter_i = _check_optimize_result(
Pipeline
                   Score
_____
TFIDF_brut
                  0.7063
TFIDF_lowStop
                 0.7000
TFIDF_lowStopstem 0.7000
TFIDF_lowcase
                 0.6947
TFIDF_lowStopna 0.6863
                 0.6821
CV_brut
```

CV_lowcase 0.6821

```
CV lowStop
                         0.6674
     CV_lowStopstem
                         0.6674
     CV_lowStopna
                         0.6674
[53]: X = df train["text"]
      y = df_train["rating"]
      # put text normalizer in the pipe is a bad idea because of the very long time,
       ⇔processing
      text_normalizer =__
       →TextNormalizer(removestopwords=False,lowercase=True,getstemmer=False,removedigit=False)
      X = text normalizer.fit transform(X)
      X_train,X_val,y_train,y_val = train_test_split(X,y,train_size=0.
       →7, random_state=42)
      pipe = Pipeline([('vect', TfidfVectorizer()),
                     ('clf', LogisticRegression(solver='lbfgs')),
                     1)
      grid = {"clf_penalty": ['12' ,'none'],
              "clf__C":np.logspace(-2,2,4),
              "clf__max_iter":[100,1000]}
      gd_srLR = GridSearchCV(pipe,
                           param_grid=grid,
                           scoring='accuracy',
                           cv=10,
      #
                             n_jobs=-1,
                          return_train_score=True)
      gd_srLR.fit(X_train, y_train)
      print('meilleur score ',
      gd_srLR.best_score_,'\n')
      print('meilleurs paramètres',
       gd_srLR.best_params_,'\n')
      print('meilleur estimateur',
      gd_srLR.best_estimator_,'\n')
     /home/richard/anaconda3/lib/python3.9/site-
     packages/sklearn/linear_model/_logistic.py:1483: UserWarning: Setting
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```

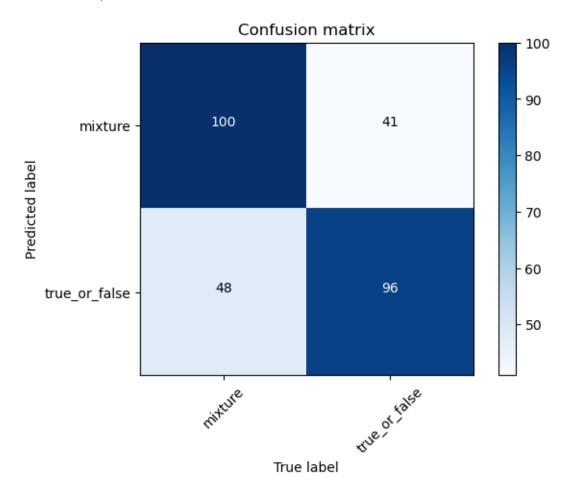
```
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     meilleur score 0.6811849841700589
     meilleurs paramètres {'clf__C': 4.6415888336127775, 'clf__max_iter': 100,
     'clf__penalty': '12'}
     meilleur estimateur Pipeline(steps=[('vect', TfidfVectorizer()), ('tfidf',
     TfidfTransformer()),
                     ('clf', LogisticRegression(C=4.6415888336127775))])
[54]: | # Creation d'une instance de l'algorithme en utilisant les meilleurs paramètres
      lr = gd_srLR.best_estimator_
      lr.fit(X_train, y_train)
      y_pred = lr.predict(X_val)
      print('\n accuracy: ', accuracy_score(y_pred, y_val),'\n')
      print('\n',classification_report(y_val, y_pred))
      plot_confusion_matrix(confusion_matrix(y_val, y_pred),classes = target_names)
```

/home/richard/anaconda3/lib/python3.9/site-

accuracy: 0.6877192982456141

	precision	recall	f1-score	support
mixture	0.68	0.71	0.69	141
true_or_false	0.70	0.67	0.68	144
accuracy			0.69	285
macro avg	0.69	0.69	0.69	285
weighted avg	0.69	0.69	0.69	285



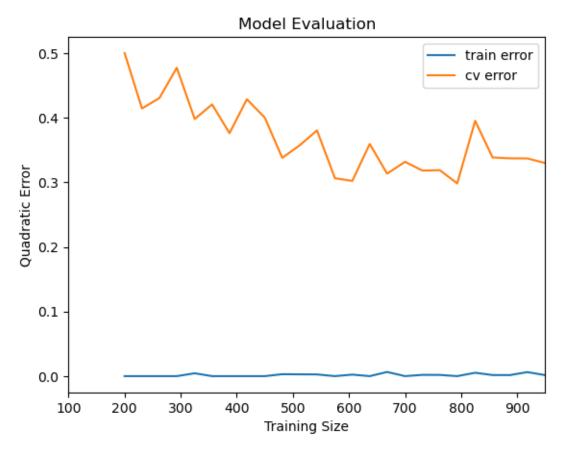
```
# print("From call: X.index: ",X.index, "y.index", y.index)
X = X.reset_index(drop=True)
y = y.reset_index(drop=True)

X = text_normalizer.fit_transform(X)

# X = X.reset_index(drop=True)
# y = y.reset_index(drop=True)

X = pd.Series(X)

plot_quad_error(X,y,lr)
```

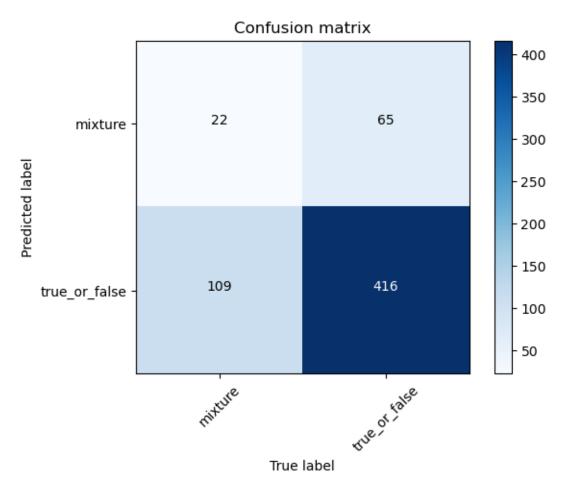


```
[56]: X_test,y_test = df_test["text"],df_test["rating"]
y_pred = lr.predict(X_test)
print('\n accuracy: ', accuracy_score(y_pred, y_test),'\n')

print('\n',classification_report(y_test, y_pred)) # ,target_names=classes
plot_confusion_matrix(confusion_matrix(y_test, y_pred),classes = target_names)
```

accuracy: 0.7156862745098039

	precision	recall	f1-score	support
mixture	0.17	0.25	0.20	87
true_or_false	0.86	0.79	0.83	525
accuracy			0.72	612
macro avg	0.52	0.52	0.51	612
weighted avg	0.77	0.72	0.74	612

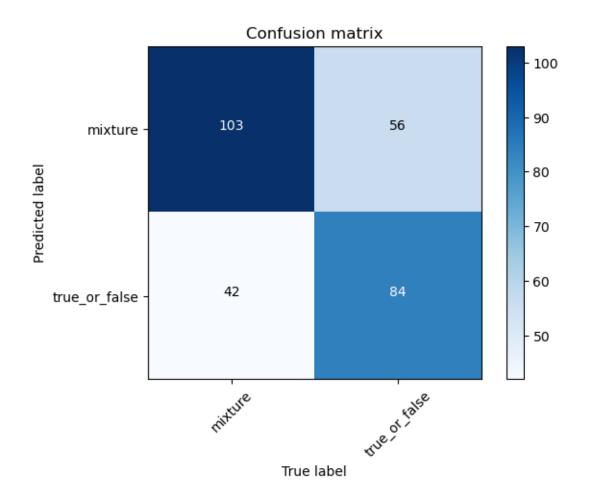


3.1.7 Svc

```
[57]: X = df train["text"]
      y = df_train["rating"]
      X = X + X.sample(300) + X if X.shape[0] <= 700 else X.sample(700)
      y_s = y.loc[X_s.index]
      preprocess_selection("SVC",SVC(),X_s,y_s)
     Evaluation des différentes configurations :
     Pipeline
                          Score
     TFIDF_brut
                         0.7242
     TFIDF_lowcase
                         0.7116
     TFIDF_lowStop
                         0.7084
     TFIDF_lowStopstem
                         0.7053
     TFIDF_lowStopna
                         0.7000
     CV_lowStop
                         0.6432
     CV_lowStopstem
                         0.6337
     CV_lowStopna
                         0.6316
     CV_brut
                         0.5884
     CV_lowcase
                         0.5842
[58]: X = df_train["text"]
      y = df_train["rating"]
      # put text normalizer in the pipe is a bad idea because of the very long time_
       ⇔processing
      text normalizer = ___
       -TextNormalizer(removestopwords=True,lowercase=True,getstemmer=True,removedigit=False)
      X = text_normalizer.fit_transform(X)
      X_train,X_val,y_train,y_val = train_test_split(X,y,train_size=0.
       →7,random_state=0)
      pipe = Pipeline([('vect', TfidfVectorizer()),
                     ('clf', SVC()),
                    ])
      grid = {'clf__C': [0.001, 0.01, 0.1, 1, 10, 100, 1000],
              'clf__kernel': ['linear'],
              'clf__gamma': [ 0.0001, 0.001, 0.01, 0.1, 1],
              'clf_kernel': ['linear','rbf','poly','sigmoid']}
      gd_srSVC = GridSearchCV(pipe,
                           param_grid=grid,
                           scoring='accuracy',
                           cv=10.
      #
                             n_{jobs}=-1,
```

```
return_train_score=True)
      gd_srSVC.fit(X_train, y_train)
      print('meilleur score ',
      gd_srSVC.best_score_,'\n')
      print('meilleurs paramètres',
      gd_srSVC.best_params_,'\n')
      print('meilleur estimateur',
      gd_srSVC.best_estimator_,'\n')
     meilleur score 0.670872908186341
     meilleurs paramètres {'clf__C': 100, 'clf__gamma': 0.1, 'clf__kernel': 'rbf'}
     meilleur estimateur Pipeline(steps=[('vect', TfidfVectorizer()), ('tfidf',
     TfidfTransformer()),
                     ('clf', SVC(C=100, gamma=0.1))])
[59]: svm = gd_srSVC.best_estimator_
      svm.fit(X_train, y_train)
      y_pred = svm.predict(X_val)
      print('\n accuracy: ', accuracy_score(y_pred, y_val),'\n')
      print('\n',classification_report(y_val, y_pred))
      plot_confusion_matrix(confusion_matrix(y_val, y_pred),classes = target_names)
      accuracy: 0.656140350877193
```

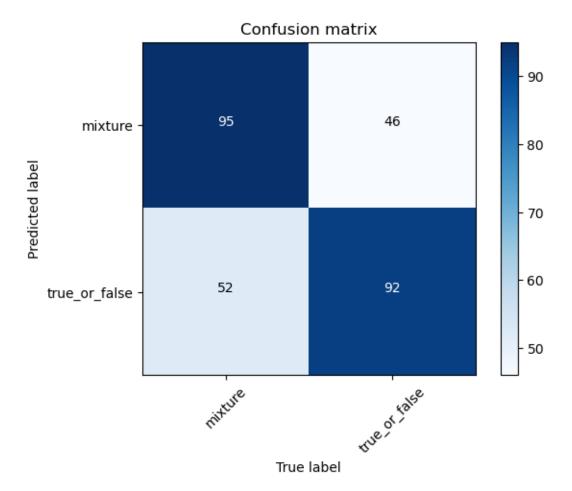
	precision	recall	f1-score	support
mixture	0.71	0.65	0.68	159
true_or_false	0.60	0.67	0.63	126
accuracy			0.66	285
macro avg	0.66	0.66	0.65	285
weighted avg	0.66	0.66	0.66	285

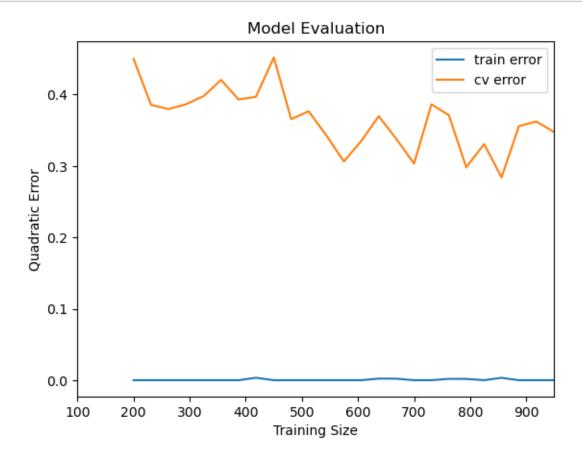


```
[60]: X = df_train["text"]
      y = df_train["rating"]
      # put text normalizer in the pipe is a bad idea because of the very long time_
      →processing
      text_normalizer = TextNormalizer(removestopwords=True,
                                       lowercase=True,
                                       getstemmer=True,
                                       removedigit=False)
      X = text_normalizer.fit_transform(X)
      X_train,X_val,y_train,y_val = train_test_split(X,y,train_size=0.
       →7,random_state=42)
     pipe = Pipeline([('vect', TfidfVectorizer()),
                     ('clf', SVC()),
                    ])
      grid = {
         'vect_stop_words':['english',None],
```

```
'vect__lowercase': [True,False],
          'clf__C': [1, 10],
          'clf__gamma' : [1],
          'clf__kernel': ['rbf']
             }
      gd_srSVC = GridSearchCV(pipe,
                           param_grid=grid,
                           scoring='accuracy',
                           cv=10,
                             n_{jobs}=-1,
                          return_train_score=True)
      gd_srSVC.fit(X_train, y_train)
      print('meilleur score ',
      gd_srSVC.best_score_,'\n')
      print('meilleurs paramètres',
      gd_srSVC.best_params_,'\n')
      print('meilleur estimateur',
      gd_srSVC.best_estimator_,'\n')
     meilleur score 0.694843962008141
     meilleurs paramètres {'clf__C': 10, 'clf__gamma': 1, 'clf__kernel': 'rbf',
     'vect_lowercase': True, 'vect_stop_words': 'english'}
     meilleur estimateur Pipeline(steps=[('vect',
     TfidfVectorizer(stop_words='english')),
                     ('tfidf', TfidfTransformer()), ('clf', SVC(C=10, gamma=1))])
[61]: # Creation d'une instance de l'algorithme en utilisant les meilleurs paramètres
      svm = gd_srSVC.best_estimator_
      svm.fit(X_train, y_train)
      y_pred = svm.predict(X_val)
      print('\n accuracy: ', accuracy_score(y_pred, y_val),'\n')
      print('\n',classification_report(y_val, y_pred))
      plot_confusion_matrix(confusion_matrix(y_val, y_pred),classes = target_names)
      accuracy: 0.656140350877193
                     precision recall f1-score
                                                    support
                         0.65
                                 0.67
                                             0.66
                                                        141
           mixture
```

true_or_false	0.67	0.64	0.65	144
accuracy			0.66	285
macro avg	0.66	0.66	0.66	285
weighted avg	0.66	0.66	0.66	285





```
[63]: X_test,y_test = df_test["text"],df_test["rating"]
y_pred = svm.predict(X_test)
print('\n accuracy: ', accuracy_score(y_pred, y_test),'\n')

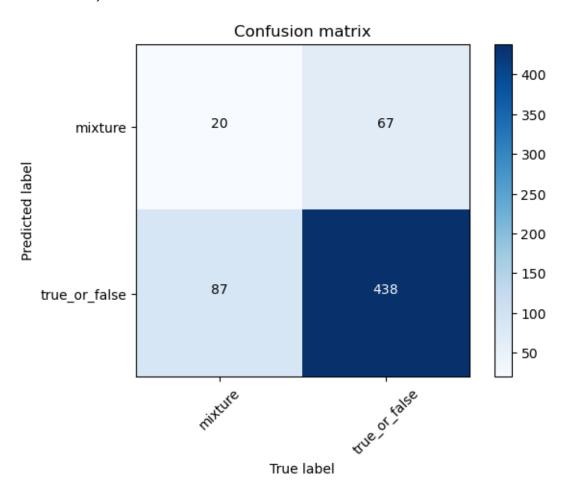
print('\n',classification_report(y_test, y_pred))
plot_confusion_matrix(confusion_matrix(y_test, y_pred),classes = target_names)
```

accuracy: 0.7483660130718954

	precision	recall	f1-score	support
mixture	0.19	0.23	0.21	87
true_or_false	0.87	0.83	0.85	525
accuracy			0.75	612
macro avg	0.53	0.53	0.53	612

weighted avg 0.77 0.75 0.76 612

Confusion matrix, without normalization



3.1.8 MultinominalNB

```
[64]: X = df_train["text"]
y = df_train["rating"]

X_s = X # X.sample(300) # X if X.shape[0] <=700 else X.sample(700)
y_s = y.loc[X_s.index]
preprocess_selection("MuntinomialNB", MultinomialNB(), X_s, y_s)</pre>
```

Evaluation des différentes configurations :

Pipeline	Score
TFIDF_brut	0.7105
TFIDF lowStop	0.7063

```
TFIDF_lowcase
                         0.7021
     CV_brut
                         0.6968
     TFIDF_lowStopstem
                         0.6947
     TFIDF_lowStopna
                         0.6937
     CV lowcase
                         0.6884
     CV lowStop
                         0.6863
     CV lowStopstem
                         0.6832
     CV_lowStopna
                         0.6821
[65]: X = df_train["text"]
      y = df_train["rating"]
      # put text normalizer in the pipe is a bad idea because of the very long time_
       ⇔processing
      text_normalizer =_
       -TextNormalizer(removestopwords=True,lowercase=True,getstemmer=True,removedigit=False)
      X = text_normalizer.fit_transform(X)
      X_train,X_val,y_train,y_val = train_test_split(X,y,train_size=0.
       →7,random_state=42)
      pipe = Pipeline([('vect', TfidfVectorizer()),
                     ('clf', MultinomialNB()),
                    ])
      grid = {'clf_alpha': np.linspace(0.5, 1.5, 6),
              'clf__fit_prior': [True, False],}
      gd_srMNB = GridSearchCV(pipe,
                           param_grid=grid,
                           scoring='accuracy',
                           cv=10,
      #
                             n_{jobs=-1},
                          return_train_score=True)
      gd_srMNB.fit(X_train, y_train)
      print('meilleur score ',
       gd_srMNB.best_score_,'\n')
      print('meilleurs paramètres',
      gd_srMNB.best_params_,'\n')
      print('meilleur estimateur',
      gd_srMNB.best_estimator_,'\n')
     meilleur score 0.6797829036635006
     meilleurs paramètres {'clf_alpha': 0.5, 'clf_fit_prior': False}
     meilleur estimateur Pipeline(steps=[('vect', TfidfVectorizer()), ('tfidf',
     TfidfTransformer()),
```

('clf', MultinomialNB(alpha=0.5, fit_prior=False))])

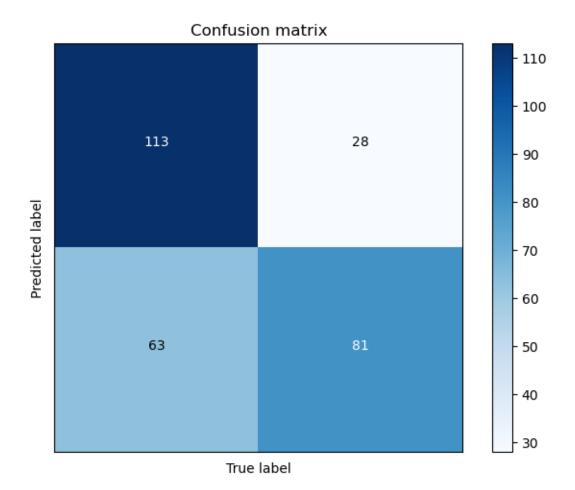
```
[66]: #Creation d'une instance de l'algorithme en utilisant les meilleurs paramètres
mNB = gd_srMNB.best_estimator_

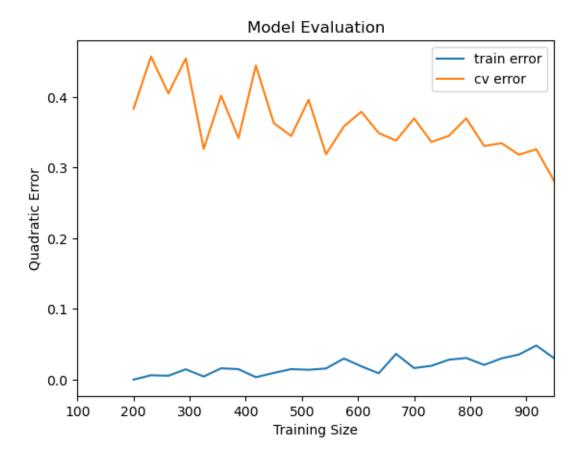
mNB.fit(X_train, y_train)
y_pred = mNB.predict(X_val)
print('\n accuracy: ', accuracy_score(y_pred, y_val),'\n')

print('\n',classification_report(y_val, y_pred))
plot_confusion_matrix(confusion_matrix(y_val, y_pred))
```

accuracy: 0.6807017543859649

	precision	recall	f1-score	support
mixture	0.64	0.80	0.71	141
true_or_false	0.74	0.56	0.64	144
accuracy			0.68	285
macro avg	0.69	0.68	0.68	285
weighted avg	0.69	0.68	0.68	285



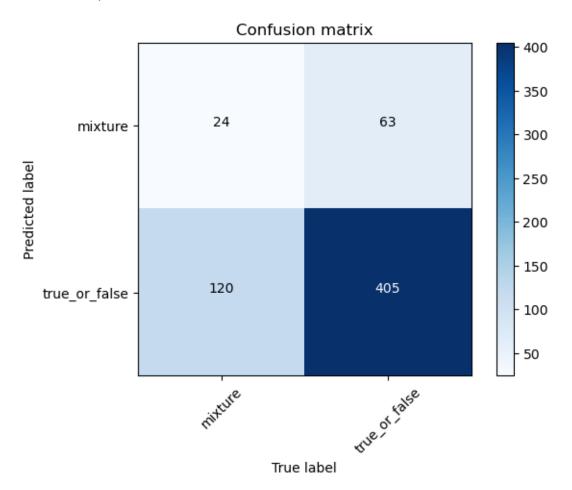


```
[68]: X_test,y_test = df_test["text"],df_test["rating"]
y_pred = mNB.predict(X_test)
print('\n accuracy: ', accuracy_score(y_pred, y_test),'\n')

print('\n',classification_report(y_test, y_pred,target_names=classes))
plot_confusion_matrix(confusion_matrix(y_test, y_pred),classes = target_names)
```

accuracy: 0.7009803921568627

	precision	recall	f1-score	support
true_or_false	0.17	0.28	0.21	87
mixture	0.87	0.77	0.82	525
accuracy			0.70	612
macro avg	0.52	0.52	0.51	612
weighted avg	0.77	0.70	0.73	612



3.1.9 RandomForest

```
[69]: X = df_train["text"]
y = df_train["rating"]

X_s = X # X.sample(300) # X if X.shape[0] <=700 else X.sample(700)
y_s = y.loc[X_s.index]
preprocess_selection("RandomForest", RandomForestClassifier(random_state=42), X_s, y_s)</pre>
```

Evaluation des différentes configurations :

Pipeline	Score
CV_lowStop	0.7084
TFIDF_lowcase	0.7084
CV_lowStopstem	0.7063
TFIDF_lowStopstem	0.6926

```
TFIDF_lowStop
                         0.6884
     CV_lowStopna
                         0.6811
     TFIDF_brut
                         0.6789
     CV_lowcase
                         0.6747
     CV brut
                         0.6747
     TFIDF_lowStopna
                         0.6705
[70]: X = df_train["text"]
      y = df_train["rating"]
      # put text normalizer in the pipe is a bad idea because of the very long time_
       ⇔processing
      text_normalizer =__
      →TextNormalizer(removestopwords=True,lowercase=True,getstemmer=True,removedigit+False)
      X = text normalizer.fit transform(X)
      X_train,X_val,y_train,y_val = train_test_split(X,y,train_size=0.
       →7,random_state=42)
      pipe = Pipeline([('vect', TfidfVectorizer()),
                     ('tfidf', TfidfTransformer()),
                     ('clf', RandomForestClassifier()),
                    1)
      parameters = {
          'clf__n_estimators': list(range(50,125,25)), #[500, 1200],
          'clf_max_depth': list(range(60,81,2)), # [25, 30],
          'clf__min_samples_split': [5, 10, 15],
          'clf__min_samples_leaf' : [1, 2],
          }
      gd_srMNB = GridSearchCV(pipe,
                           param_grid=parameters,
                           scoring='accuracy',
                           cv=10,
      #
                             n jobs=-1,
                          return_train_score=True)
      gd_srMNB.fit(X_train, y_train)
      print('meilleur score ',
      gd_srMNB.best_score_,'\n')
      print('meilleurs paramètres',
       gd_srMNB.best_params_,'\n')
      print('meilleur estimateur',
      gd_srMNB.best_estimator_,'\n')
```

meilleur score 0.6916553595658075

```
[71]: rfc = gd_srMNB.best_estimator_

rfc.fit(X_train, y_train)
y_pred = rfc.predict(X_val)
print('\n accuracy: ', accuracy_score(y_pred, y_val),'\n')

print('\n',classification_report(y_val, y_pred))
plot_confusion_matrix(confusion_matrix(y_val, y_pred,classes = target_names))
```

accuracy: 0.6421052631578947

	precision	recall	f1-score	support
mixture	0.63	0.65	0.64	141
true_or_false	0.65	0.63	0.64	144
accuracy			0.64	285
macro avg	0.64	0.64	0.64	285
weighted avg	0.64	0.64	0.64	285

```
[]: X = df_train["text"]
     y = df_train["rating"]
     text_normalizer =__
      →TextNormalizer(removestopwords=True,lowercase=True,getstemmer=True,removedigit=False)
     X = X.reset_index(drop=True)
     y = y.reset_index(drop=True)
     X = text_normalizer.fit_transform(X)
     X = pd.Series(X)
     plot_quad_error(X,y,rfc)
[]: X_test,y_test = df_test["text"],df_test["rating"]
     y_pred = rfc.predict(X_test)
     print('\n accuracy: ', accuracy_score(y_pred, y_test),'\n')
     print('\n',classification_report(y_test, y_pred))
     plot_confusion_matrix(confusion_matrix(y_test, y_pred),classes = target_names)
[]: # ('RandomForest', RandomForestClassifier())
[]: models = pd.DataFrame({
         "Models": ["Logistic Regression" , "SVM", "Multinominal Naive_{\sqcup}
      ⇔Bayes", "Random Forest Classifier"],
         "Score": [lr.score(X_test,y_test), svm.score(X_test,y_test), mNB.
      ⇒score(X_test,y_test), rfc.score(X_test,y_test)]
     models.sort_values(by="Score" , ascending=False)
[]:
```