# LTU\_PhD\_CyberSecAI\_ProgTask\_SMondal\_v0.003

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# 1 Programming Task:

# 1.1 ## Language Recognition Using Distributed High Dimensional Representations

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# 1.1.1 Objective:

Classify the language of an input text. The language recognition will be done for 21 European languages. The list of languages being - Bulgarian, Czech, Danish, German, Greek, English, Estonian, Finnish, French, Hungarian, Italian, Latvian, Lithuanian, Dutch, Polish, Portuguese, Romanian, Slovak, Slovene, Spanish, Swedish. The training and testing data is based on the Wortschatz Corpora: https://wortschatz.uni-leipzig.de/en/download

#### 1.1.2 Approach:

- Smaller size subsets of original data sets were used for faster processing. In fact selected randomly 500 rows from each language corpora and combined together to form train data set for the classification data model.
- Data preprocessing was necessary. The data procured from Leipzig corpora had some noise in it. Used Notepad++ for data cleaning. Removed symbols, punctuation marks, digits from the data. Later checked for punctuation marks using Jupyter notebook
- A 70-30 split was done for model training and testing purpose.
- Using tri-grams were memory intensive especially when used with Count Vectorizier, a sklearn library
- Ridge Classification was used to train the model, an implementation from sklearn library
- Test data were taken from euparl repository
- Confusion matrix, accuracy and F1-scores were calculated and plotted

#### 1.1.3 Model Building

#### Step-1: Import the necessary Libraries

```
[1]: import re import string
```

```
import itertools
import seaborn as sns
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.preprocessing import LabelEncoder
from sklearn.feature_extraction.text import CountVectorizer
from sklearn.model_selection import train_test_split
from sklearn.linear_model import RidgeClassifier
from sklearn.metrics import accuracy_score, confusion_matrix,u
classification_report
from sklearn import metrics
```

#### Step-2: Load data

Data was preprocessed. It was obtained by combining news data in selected languages from the Leipzig Corpora repository.

```
[2]: data = pd.read_csv("../input/leipzig21_500_sub.csv")
   data.head(5)
```

```
[2]: Text Language

0 ... Bulgarian

1 ... Bulgarian

2 ... Bulgarian

3 ... Bulgarian

4 ... Bulgarian
```

Outline of data. It shows no of rows and memory usage.

```
[3]: print(data.info())
```

Rows by language type.

```
[4]: data['Language'].value_counts()
```

[4]: Language
Bulgarian 500
Greek 500

```
Swedish
              500
              500
Spanish
Slovenian
              500
Slovak
              500
Romanian
              500
Portuguese
              500
Polish
              500
English
              500
Dutch
              500
Czech
              500
Lithuanian
              500
Lativian
              500
Italian
              500
French
              500
Finnish
              500
Estonian
              500
German
              500
Danish
              500
              500
Hungarian
Name: count, dtype: int64
```

# Step-3: Separating Label from the data

Labels are the predictor variables, dependent variable. While the other features are independent variables.

```
[5]: X = data["Text"]
y = data["Language"]
```

Label Encoding

```
[6]: le = LabelEncoder()
y = le.fit_transform(y)
```

#### Step-4: Text Preprocessing

```
C:\Users\shara\AppData\Local\Temp\ipykernel_26108\1572569825.py:7:
FutureWarning: Possible nested set at position 1
  text = re.sub(r'[[]]', ' ', text)
```

#### Bag of Words

Output feature and input feature should be of the numerical form. So we are converting text into numerical form by creating a Bag of Words model using CountVectorizer.

```
[8]: cv = CountVectorizer(ngram_range=(1,1), analyzer="word", min_df=1, max_df=1.0)

# unigram (1,1) for word analyzer

X = cv.fit_transform(data_list).toarray()

X.shape #(10500, 157042)
```

[8]: (10500, 64952)

#### 1.1.4 Visualization: WordCloud

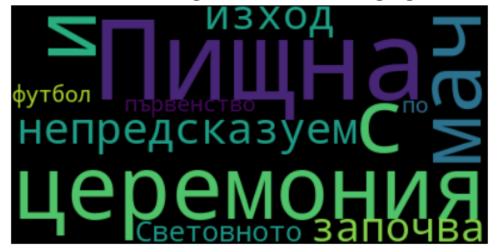
```
[9]: from wordcloud import WordCloud
```

```
[10]: # Start with one review:
    text = data.Text[0]

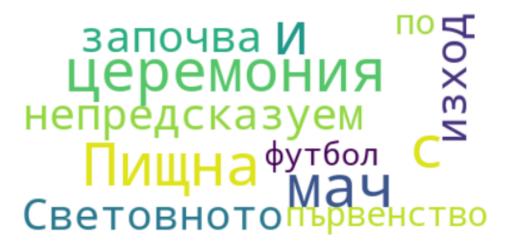
# Create and generate a word cloud image:
    wordcloud = WordCloud().generate(text)

# Display the generated image:
    plt.imshow(wordcloud, interpolation='bilinear')
    plt.axis("off")
    plt.title("WordCloud Of A Single Sentence From Language Text")
    plt.show()
```

# WordCloud Of A Single Sentence From Language Text



WordCloud Of One Hundred Sentences From Language Text



```
[12]: text = " ".join(review for review in data.Text)
print ("There are {} words in the combination of all review.".format(len(text)))
```

There are 1125189 words in the combination of all review.

[13]: from wordcloud import STOPWORDS

```
max_font_size= 100, background_color="white",
max_words=5000).generate(text)

# Display the generated image:
# the matplotlib way:
plt.figure(figsize = (15, 8))
plt.imshow(wordcloud, interpolation='bilinear')
plt.title("WordCloud of Language Text from 21 European Languages")
plt.axis("off")
#plt.axes(linewidth=1.0)
plt.show()
```



```
[15]: # Save the image in the img folder:
wordcloud.to_file("../img/five_thousand_sentences.png")
```

[15]: <wordcloud.wordcloud.WordCloud at 0x254bd5500b0>

# Step-5: Train Test Data Splitting

```
[16]: x_train, x_test, y_train, y_test = train_test_split(X, y, test_size = 0.30, u →random_state=0)
```

Step-6: Model Training and Prediction

```
[17]: model = RidgeClassifier()
model.fit(x_train, y_train)
```

[17]: RidgeClassifier()

# Prediction

```
[18]: y_pred = model.predict(x_test)
```

# Step-7: Model Evaluation

Model Accuracy

```
[19]: lang_labels = le.inverse_transform(np.unique(y_test))
```

```
[20]: ac = accuracy_score(y_test, y_pred)
    cm = confusion_matrix(y_test, y_pred)
    clf_report = classification_report(y_test, y_pred, target_names=lang_labels)
    print("Accuracy is :",ac)
```

Accuracy is : 0.9482539682539682

# [21]: print(clf\_report)

	precision	recall	f1-score	support
Bulgarian	1.00	0.99	1.00	153
Czech	0.92	0.90	0.91	129
Danish	0.95	0.95	0.95	145
Dutch	0.93	0.96	0.94	147
English	0.96	0.97	0.97	159
Estonian	0.94	0.91	0.93	150
Finnish	0.95	0.93	0.94	155
French	0.96	0.97	0.96	137
German	0.92	0.97	0.95	151
Greek	1.00	0.98	0.99	150
Hungarian	0.99	0.94	0.96	145
Italian	0.92	0.97	0.95	150
Lativian	0.97	0.95	0.96	164
Lithuanian	0.95	0.94	0.94	147
Polish	0.90	0.94	0.92	150
Portuguese	0.93	0.93	0.93	153
Romanian	0.98	0.93	0.95	170
Slovak	0.91	0.89	0.90	140
Slovenian	0.93	0.93	0.93	139
Spanish	0.98	0.97	0.98	163
Swedish	0.92	0.95	0.94	153
accuracy			0.95	3150
macro avg	0.95	0.95	0.95	3150

weighted avg 0.95 0.95 0.95 3150

Confusion Matrix

```
[22]: #df_cm = pd.DataFrame(cm, columns=np.unique(y_test), index = np.unique(y_test))

df_cm = pd.DataFrame(cm, columns=lang_labels, index = lang_labels)

df_cm.index.name = 'Actual'

df_cm.columns.name = 'Predicted'

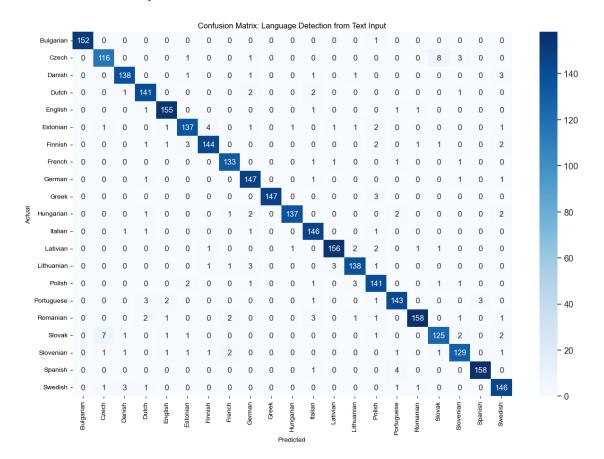
plt.figure(figsize = (15,10))

plt.title("Confusion Matrix: Language Detection from Text Input")

sns.set(font_scale=1.2) #for label size

sns.heatmap(df_cm, cmap="Blues", annot=True,annot_kws={"size": 12}, fmt="d")#__

$\infty font size$
```



**Step-8: Model Validation** 

```
[23]: def predict(text):
    x = cv.transform([text]).toarray() # converting text to bag of words model
    ∴(Vector)
    lang = model.predict(x) # predicting the language
    lang = le.inverse_transform(lang) # finding the language corresponding the
    ∴the predicted value
    print("The language is in",lang[0]) # printing the language
```

The langauge is in Bulgarian

```
[25]: predict("komandų žaidėjai rungtynes pradėjo be didesnės žvalgybos")
```

The langauge is in Lithuanian

# 1.2 Data Pre-Processing Step: Read & Combine data to form a single csv file for building the classification model

```
[27]: for filename, language in zip(data_files, languages):
    df_name = language + '_df'
    file = "../input/data/{}.csv".format(filename)
    df_dic[df_name] = pd.read_csv(file)
```

```
[28]: df = pd.concat(df_dic.values(), ignore_index=True)
print(df.info())
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 630000 entries, 0 to 629999

```
Column
                    Non-Null Count
                                     Dtype
                    _____
      0
                    630000 non-null object
          Text
      1
          Language 630000 non-null object
     dtypes: object(2)
     memory usage: 9.6+ MB
     None
[29]: df['Language'].value_counts()
[29]: Language
      Bulgarian
                    30000
      Greek
                    30000
      Swedish
                    30000
      Spanish
                    30000
      Slovenian
                    30000
      Slovak
                    30000
      Romanian
                    30000
     Portuguese
                    30000
     Polish
                    30000
     English
                    30000
     Dutch
                    30000
      Czech
                    30000
     Lithuanian
                    30000
      Lativian
                    30000
      Italian
                    30000
      French
                    30000
      Finnish
                    30000
      Estonian
                    30000
      German
                    30000
      Danish
                    30000
      Hungarian
                    30000
      Name: count, dtype: int64
          Creating a subset of data by selecting a few rows randomly
     1.3
[30]: dfsub_dic = {}
      for name, dat in df_dic.items():
          dfsub_name = name + '_subdf'
          dfsub_dic[dfsub_name] = dat.sample(n=500, random_state=3).
       →reset_index(drop=True)
[31]: subdf = pd.concat(dfsub_dic.values(), ignore_index=True)
      print(subdf.info())
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 10500 entries, 0 to 10499
```

Data columns (total 2 columns):

```
Column
                    Non-Null Count Dtype
                    _____
      0
          Text
                    10500 non-null object
      1
          Language 10500 non-null
     dtypes: object(2)
     memory usage: 164.2+ KB
     None
[32]: subdf['Language'].value_counts()
[32]: Language
      Bulgarian
                    500
      Greek
                    500
      Swedish
                    500
      Spanish
                    500
      Slovenian
                    500
      Slovak
                    500
      Romanian
                    500
     Portuguese
                    500
     Polish
                    500
     English
                    500
     Dutch
                    500
      Czech
                    500
     Lithuanian
                    500
     Lativian
                    500
      Italian
                    500
      French
                    500
     Finnish
                    500
      Estonian
                    500
      German
                    500
      Danish
                    500
      Hungarian
                    500
      Name: count, dtype: int64
[33]: subdf.to_csv("../input/leipzig21_500_sub.csv", index=False)
[34]: df_ = pd.read_csv("../input/leipzig21_500_sub.csv")
      print(df.info())
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 630000 entries, 0 to 629999
     Data columns (total 2 columns):
          Column
                    Non-Null Count
                                     Dtype
          _____
                    _____
      0
          Text
                    630000 non-null object
          Language 630000 non-null object
      1
     dtypes: object(2)
```

Data columns (total 2 columns):

```
memory usage: 9.6+ MB
```

None

```
[35]: df_.head()

[35]: Text Language

0 ... Bulgarian
1 ... Bulgarian
2 ... Bulgarian
3 ... Bulgarian
4 ... Bulgarian
[]:
```