uage-detection-capstone-project-1

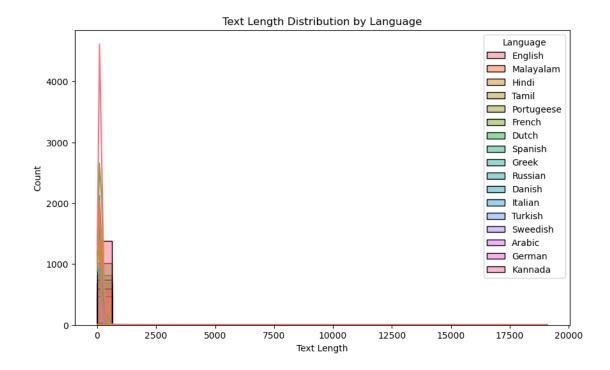
November 19, 2024

```
[62]: import pandas as pd
      import numpy as np
      from sklearn.feature_extraction.text import CountVectorizer
      from sklearn.model_selection import train_test_split
      from sklearn.naive_bayes import MultinomialNB
      data = pd.read_csv("Language Detection.csv")
      print(data.head())
                                                      Text Language
        Nature, in the broadest sense, is the natural... English
     1 "Nature" can refer to the phenomena of the phy... English
     2 The study of nature is a large, if not the onl... English
     3 Although humans are part of nature, human acti...
                                                          English
     4 [1] The word nature is borrowed from the Old F...
                                                          English
[63]: data.isnull().sum()
[63]: Text
                  0
      Language
                  0
      dtype: int64
[64]: data["Language"].value_counts()
[64]: Language
      English
                    1385
      French
                    1014
      Spanish
                     819
      Portugeese
                     739
      Italian
                     698
      Russian
                     692
      Sweedish
                     676
     Malayalam
                     594
      Dutch
                     546
      Arabic
                     536
      Turkish
                     474
      German
                     470
      Tamil
                     469
      Danish
                     428
```

```
Greek
                     365
      Hindi
                      63
      Name: count, dtype: int64
[65]: import matplotlib.pyplot as plt
      import seaborn as sns
      # Text Length Distribution
      data['Text Length'] = data['Text'].apply(len)
      plt.figure(figsize=(10, 6))
      sns.histplot(data=data, x='Text Length', hue='Language', bins=30, kde=True)
      plt.title('Text Length Distribution by Language')
      plt.show()
      # Word Frequency Analysis
      from collections import Counter
      # Combine all texts for analysis
      all_text = " ".join(data['Text'])
      # Tokenize text (split into words)
      words = all_text.split()
      # Count word frequencies
      word freq = Counter(words)
      # Print the most common words
      print("Most common words:")
      for word, freq in word_freq.most_common(10):
          print(f"{word}: {freq}")
      # Character Distribution
      all_characters = " ".join(" ".join(data['Text']).split())
      # Count character frequencies
      char_freq = Counter(all_characters)
      # Plot character frequencies
      char_freq_df = pd.DataFrame.from_dict(char_freq, orient='index').reset_index()
      char_freq_df.columns = ['Character', 'Frequency']
      char_freq_df = char_freq_df.sort_values(by='Frequency', ascending=False)
      plt.figure(figsize=(10, 6))
      sns.barplot(data=char_freq_df.head(20), x='Character', y='Frequency')
      plt.title('Top 20 Most Frequent Characters')
      plt.show()
```

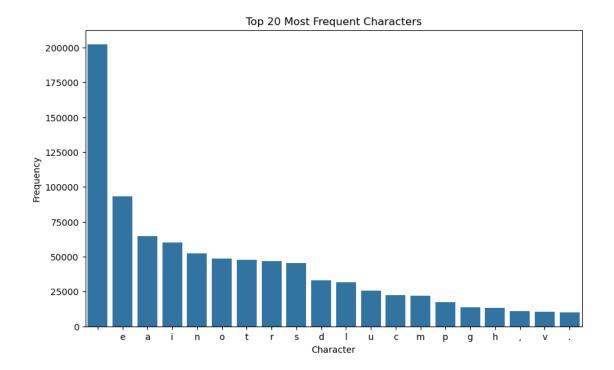
Kannada

369



Most common words:

de: 3428 a: 1722 en: 1395 the: 1368 la: 1316 que: 1259 of: 1127 in: 1071 and: 836 du: 782



```
[66]: from sklearn.model_selection import train_test_split, cross_val_score
      from sklearn.feature_extraction.text import CountVectorizer
      from sklearn.naive_bayes import MultinomialNB
      # Data Splitting
      x = data["Text"]
      y = data["Language"]
      # Create CountVectorizer
      cv = CountVectorizer()
      X = cv.fit_transform(x)
      # Split the data into training and testing sets
      X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,_
       →random_state=42)
      # Train a Multinomial Naive Bayes model
      model = MultinomialNB()
      model.fit(X_train, y_train)
      # Perform cross-validation for evaluation
      # You can change the number of folds (e.g., cv=5) as needed
      cross_val_scores = cross_val_score(model, X, y, cv=10, scoring='accuracy')
```

```
# Print cross-validation results
      print("Cross-validation scores:", cross_val_scores)
      print("Mean accuracy:", cross_val_scores.mean())
     Cross-validation scores: [0.98355899 0.99226306 0.99419729 0.99032882 0.98742747
     0.98065764
      0.96518375 0.94191675 0.95159729 0.97967086]
     Mean accuracy: 0.9766801919630904
 []: from sklearn.preprocessing import LabelEncoder
      # Define a maximum text length for padding or truncation
      max_text_length = 100 # Adjust as needed
      # Ensure all texts have a consistent length by padding or truncating
      x = [text[:max text length] if len(text) > max text length else text.
       →ljust(max_text_length) for text in x]
      # Create a label encoder to encode language labels
      label encoder = LabelEncoder()
      #label_encoder.fit(y)
      y_encoded = label_encoder.fit_transform(y)
      # Example: Language detection for a short text
      user_input = "Bonjour"
      user_input = user_input[:max_text_length] if len(user_input) > max_text_length_
       ⇔else user_input.ljust(max_text_length)
      user_input_encoded = cv.transform([user_input]).toarray()
      predicted_language_encoded = model.predict(user_input_encoded)
      predicted_language = label_encoder.inverse_transform(predicted_language_encoded)
      print("Predicted Language:", predicted_language[0])
[69]: import re
      # Data Cleaning
      data.dropna(inplace=True) # Remove rows with missing values
      data.drop_duplicates(inplace=True) # Remove duplicate rows
      # Text Preprocessing
      def preprocess_text(text):
          # Convert text to lowercase
          text = text.lower()
          # Remove special characters and digits using regular expressions
          text = re.sub(r'[^a-z\s]', '', text)
```

return text

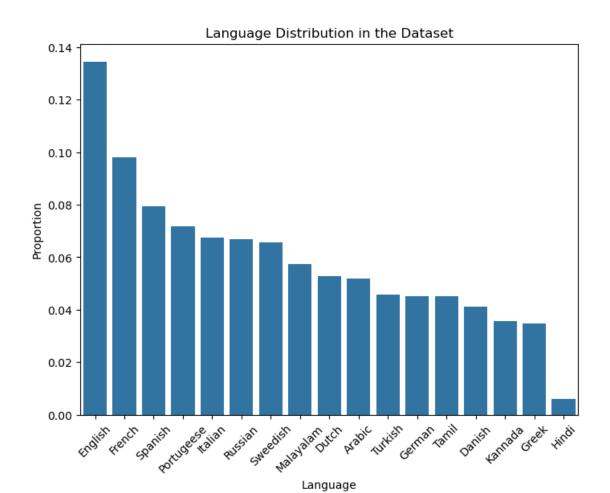
```
# Apply text preprocessing to the "Text" column
data['Text'] = data['Text'].apply(preprocess_text)

# Model Evaluation
accuracy = model.score(X_test, y_test)
print("Model Accuracy:", accuracy)
```

Model Accuracy: 0.9825918762088974

```
[70]: # Bias Analysis
# Calculate the distribution of languages in the dataset
language_distribution = data['Language'].value_counts(normalize=True)

# Plot the distribution of languages
plt.figure(figsize=(8, 6))
sns.barplot(x=language_distribution.index, y=language_distribution.values)
plt.title('Language Distribution in the Dataset')
plt.xlabel('Language')
plt.ylabel('Proportion')
plt.xticks(rotation=45)
plt.show()
```



```
[71]: # Noise Analysis
    # Check for empty or very short text samples
    data['Text_Length'] = data['Text'].apply(len)
    short_text_threshold = 10  # Adjust as needed
    noise_short_texts = data[data['Text_Length'] < short_text_threshold]

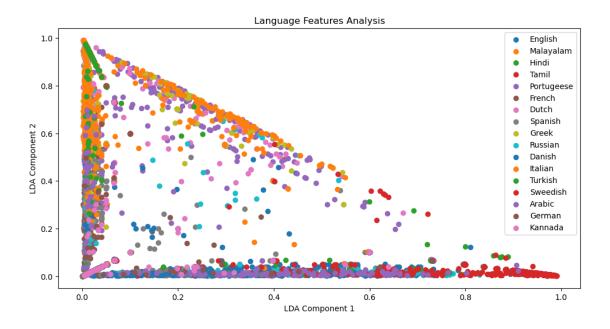
# Check for text samples with very long lengths
    long_text_threshold = 1000  # Adjust as needed
    noise_long_texts = data[data['Text_Length'] > long_text_threshold]

# Display the noisy short and long text samples
    print("Noisy Short Texts (length <", short_text_threshold, "):")
    print(noise_short_texts[['Text', 'Language']])

print("\nNoisy Long Texts (length >", long_text_threshold, "):")
    print(noise_long_texts[['Text', 'Language']])
```

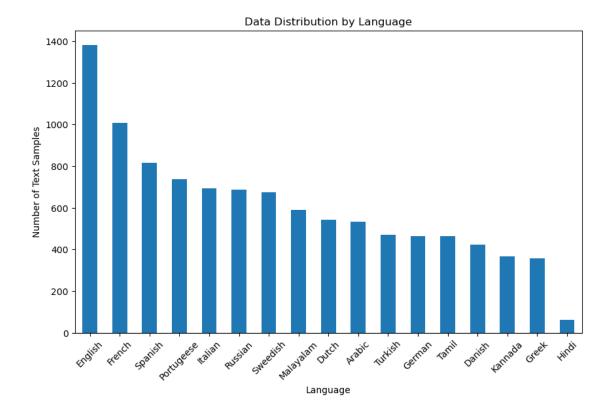
Noisy Short Texts (length < 10):

```
Text Language
     347
              kennedy English
     750
                  gne English
     1125
             its fine English
             im sorry English
     1133
            oh my god English
     1138
                       Kannada
     10294
     10313
                       Kannada
     10320
                       Kannada
     10331
                       Kannada
     10332
                       Kannada
     [1605 rows x 2 columns]
     Noisy Long Texts (length > 1000 ):
                                                         Text
                                                                 Language
     1979
                  \n
                                                                  Hindi
                            \n
                                    \n
                                                     \n
                                           \n
     1997
                                             \n
                                                                  Hindi
     2041
                                                                  Hindi
            o mediawiki tem documentao sobre wikipdia ing... Portugeese
     2889
     3858 ces programmes selon leur degr de perfectionne...
                                                                 French
[72]: from sklearn.decomposition import LatentDirichletAllocation
      # Apply Latent Dirichlet Allocation (LDA) for feature analysis
      lda = LatentDirichletAllocation(n_components=5, random_state=42) # You can_
       ⇔adjust the number of components
      X_lda = lda.fit_transform(X)
      # Visualize language features using LDA components
      plt.figure(figsize=(12, 6))
      for language in data['Language'].unique():
          plt.scatter(X_lda[y == language][:, 0], X_lda[y == language][:, 1],__
       →label=language)
      plt.title('Language Features Analysis')
      plt.xlabel('LDA Component 1')
      plt.ylabel('LDA Component 2')
      plt.legend()
      plt.show()
```



```
[73]: # Data Distribution Analysis
language_distribution = data['Language'].value_counts()

# Plot the distribution of languages
plt.figure(figsize=(10, 6))
language_distribution.plot(kind='bar')
plt.title('Data Distribution by Language')
plt.xlabel('Language')
plt.ylabel('Number of Text Samples')
plt.xticks(rotation=45)
plt.show()
```



Model Accuracy: 0.6935103244837758

```
[76]: import tkinter as tk

from tkinter import messagebox
import webbrowser
import pandas as pd
import numpy as np
from sklearn.feature_extraction.text import CountVectorizer
from sklearn.naive_bayes import MultinomialNB
```

```
# Load the dataset
data = pd.read_csv("Language Detection.csv")
# Create the CountVectorizer and train the MultinomialNB model
x = np.array(data["Text"])
y = np.array(data["Language"])
cv = CountVectorizer()
X = cv.fit transform(x)
model = MultinomialNB()
model.fit(X, y)
# Global variable to store detected language
detected_language = None
# Function to perform language detection
def detect_language():
    global detected_language
    user_input = text_entry.get(1.0, tk.END).strip()
    if not user_input:
        messagebox.showwarning("Input Error", "Please enter some text for_
 ⇔detection.")
        return
    data = cv.transform([user_input]).toarray()
    detected_language = model.predict(data)[0]
    language_result.config(text=f"The language is: {detected_language}")
# Function to open Google with search for the detected language
def view_details():
    if not detected_language:
        messagebox.showwarning("No Language Detected", "Please detect all
 ⇔language first.")
        return
    query = f"{detected language} language"
    webbrowser.open(f"https://www.google.com/search?q={query}")
# Function to switch windows
def show frame(frame):
    frame.tkraise()
# Create the root window
root = tk.Tk()
root.title("Stylish Language Detection App")
root.geometry("800x600")
# Create a container for stacking frames
container = tk.Frame(root)
```

```
container.pack(fill="both", expand=True)
# Configure grid layout for the container
container.grid_rowconfigure(0, weight=1)
container.grid_columnconfigure(0, weight=1)
# Function to set background color for frames
def set_background(frame, color):
   frame.configure(bg=color)
# Create the frames
main menu = tk.Frame(container)
language_detection = tk.Frame(container)
about_window = tk.Frame(container)
# Assign different colors to frames
set_background(main_menu, "#FAF6E3") # Light Cream
set_background(language_detection, "#FAF6E3") # Light Cream
set_background(about_window, "#FAF6E3") # Light Cream
for frame in (main_menu, language_detection, about_window):
   frame.grid(row=0, column=0, sticky="nsew")
# Main Menu Frame
main_label = tk.Label(main_menu, text="Welcome to Language Detection App", __
 main_label.pack(pady=20)
detect_button_main = tk.Button(main_menu, text="Go to Language Detection", __
 ⇔command=lambda: show_frame(language detection), bg="#D8DBBD", fg="#2A3663", □
detect button main.pack(pady=10)
about button main = tk.Button(main menu, text="About/Help", command=lambda:___
 ⇒show_frame(about_window), bg="#D8DBBD", fg="#2A3663", font=("Helvetica", 14))
about_button_main.pack(pady=10)
# Language Detection Frame
input_label = tk.Label(language_detection, text="Enter a text:",__

¬font=("Helvetica", 14), bg="#FAF6E3", fg="#2A3663")

input_label.pack(pady=10)
text_entry = tk.Text(language_detection, height=10, width=70, bg="#FFFFFF", __
 ⇒fg="black", font=("Helvetica", 12)) # Updated to White
text_entry.pack(pady=10)
```

```
detect_button = tk.Button(language_detection, text="Detect Language",_
 ⇔command=detect_language, bg="#D8DBBD", fg="#2A3663", font=("Helvetica", 14))
detect_button.pack(pady=10)
language_result = tk.Label(language_detection, text="", font=("Helvetica", 14),__
 ⇔bg="#FAF6E3", fg="#2A3663", wraplength=600)
language_result.pack(pady=10)
details_button = tk.Button(language_detection, text="View Details", __
 ocommand=view_details, bg="#D8DBBD", fg="#2A3663", font=("Helvetica", 14))
details_button.pack(pady=10)
back_button = tk.Button(language_detection, text="Back to Main Menu", __
Germand=lambda: show_frame(main_menu), bg="#D8DBBD", fg="#2A3663", L
 ⇔font=("Helvetica", 14))
back_button.pack(pady=10)
# About/Help Frame
about_label = tk.Label(about_window, text="About/Help", font=("Helvetica", 18, __
about_label.pack(pady=20)
help_text = (
    "This application detects the language of a given text using a trained "
   "Naive Bayes model. To use:\n\n"
   "1. Enter some text in the input box.\n"
    "2. Click the 'Detect Language' button.\n"
    "3. The detected language will be displayed below the button.\n"
   "4. Click 'View Details' to search for more information about the detected \sqcup
 ⇔language on Google.\n\n"
    "Navigate between windows using the buttons."
help_label = tk.Label(about_window, text=help_text, font=("Helvetica", 12), ___
 ⇔bg="#FAF6E3", fg="#2A3663", wraplength=700, justify="left")
help label.pack(pady=10)
back_button_about = tk.Button(about_window, text="Back to Main Menu",_
 ⇒command=lambda: show_frame(main_menu), bg="#D8DBBD", fg="#2A3663",⊔
 ⇔font=("Helvetica", 14))
back button about.pack(pady=10)
# Start with the main menu
show_frame(main_menu)
# Run the application
root.mainloop()
```

[]:[