

# CISB5123 Text Analytics

## Lab 4

### Basic Text Pre-Processing

Text pre-processing is a process to clean and prepare the textual data before they can be used as the input for text modelling techniques. In this lab, we will apply text preprocessing techniques step-by-step on a dataset and store the result.

#### **Step 1: Read the source data**

```
# Load dataset import  
pandas as pd
```

```
file_path = "Review.csv" df  
= pd.read_csv(file_path)
```

```
# Display column content without truncation  
pd.set_option('display.max_colwidth', None) # Set to None for unlimited width  
print(df)
```

#### **Step 2: Perform Text Pre-Processing**

### **a. Convert text to lowercase**

```
# Lowercase conversion def
convert_to_lowercase(text):
    return text.lower()

df["lowercased"] = df["Review"].apply(convert_to_lowercase)

# Display column content without truncation
pd.set_option('display.max_colwidth', None) # Set to None for unlimited width
print(df["lowercased"])
```

### **b. Remove URLs**

```
# Removal of URLs import
re

# remove any URLs that start with "http" or "www" from the text def
remove_urls(text):
    return re.sub(r'http\S+|www\S+', '', text)

df["urls_removed"] = df["lowercased"].apply(remove_urls)

# Display column content without truncation
pd.set_option('display.max_colwidth', None) # Set to None for unlimited width
print(df["urls_removed"])
```

### **c. Remove HTML tags**

```
# Removal of HTML tags from
bs4 import BeautifulSoup

# extracts only the text, removing all HTML tags def
remove_html_tags(text):
    return BeautifulSoup(text, "html.parser").get_text()

df["html_removed"] = df["urls_removed"].apply(remove_html_tags)
```

```
# Display column content without truncation
pd.set_option('display.max_colwidth', None) # Set to None for unlimited width
print(df["html_removed"])
```

#### **d. Remove emojis**

```
# Removal of emojis (if any) import
emoji
```

```
# replace emoji with "" def
remove_emojis(text):
    return emoji.replace_emoji(text, replace="")
```

```
df["emojis_removed"] = df["html_removed"].apply(remove_emojis)
```

```
# Display column content without truncation
pd.set_option('display.max_colwidth', None) # Set to None for unlimited width
print(df["emojis_removed"])
```

#### **e. Replace internet slang/chat words**

```
# Replace internet slang/chat words
# Dictionary of slang words and their replacements slang_dict
= {
    "tbh": "to be honest",
    "omg": "oh my god",
    "lol": "laugh out loud",
    "idk": "I don't know",
    "brb": "be right back",
    "btw": "by the way",
    "imo": "in my opinion",
    "smh": "shaking my head",
    "fyi": "for your information",
    "np": "no problem",
    "ikr": "I know right",
    "asap": "as soon as possible",
    "bff": "best friend forever",
    "gg": "good game",
```

```

    "hmu": "hit me up",
    "rofl": "rolling on the floor laughing"
}

```

```

# Function to replace slang words def

```

```

replace_slang(text):

```

```

    # Create a list of escaped slang words    escaped_slang_words = []

```

```

# Empty list to store escaped slang words

```

```

    for word in slang_dict.keys():

```

```

        escaped_word = re.escape(word) # Ensure special characters are escaped

```

```

    escaped_slang_words.append(escaped_word) # Add to list

```

```

    # Join the words using '|'    slang_pattern = r'\b(' +
    '|'join(escaped_slang_words) + r')\b'

```

```

    # Define a replacement function

```

```

def replace_match(match):

```

```

    slang_word = match.group(0) # Extract matched slang word

```

```

    return slang_dict[slang_word.lower()] # Replace with full form

```

```

    # Use regex to replace slang words with full forms    replaced_text =
    re.sub(slang_pattern, replace_match, text, flags=re.IGNORECASE)

```

```

    return replaced_text

```

```

# Apply the function to the column df["slangs_replaced"] =
df["emojis_removed"].apply(replace_slang)

```

```

# Display column content without truncation

```

```

pd.set_option('display.max_colwidth', None) # Set to None for unlimited width

```

```

print(df["slangs_replaced"])

```

## f. Replace contractions

```

# Replace Contractions

```

```

contractions_dict = {

```

```

    "wasn't": "was not",

```

"isn't": "is not",  
"aren't": "are not",  
"weren't": "were not",  
"doesn't": "does not",  
"don't": "do not",  
"didn't": "did not",  
"can't": "cannot",  
"couldn't": "could not",  
"shouldn't": "should not",  
"wouldn't": "would not",  
"won't": "will not",  
"haven't": "have not",  
"hasn't": "has not",  
"hadn't": "had not",  
"i'm": "i am",  
"you're": "you are",  
"he's": "he is",  
"she's": "she is",  
"it's": "it is",  
"we're": "we are",  
"they're": "they are",  
"i've": "i have",  
"you've": "you have",  
"we've": "we have",  
"they've": "they have",  
"i'd": "i would",  
"you'd": "you would",  
"he'd": "he would",  
"she'd": "she would",  
"we'd": "we would",  
"they'd": "they would",  
"i'll": "i will",  
"you'll": "you will",  
"he'll": "he will",  
"she'll": "she will",  
"we'll": "we will",  
"they'll": "they will",  
"let's": "let us",

```

    "that's": "that is",
    "who's": "who is",
    "what's": "what is",
    "where's": "where is",
    "when's": "when is",
    "why's": "why is"
}

# Build the regex pattern for contractions escaped_contractions
= [] # List to store escaped contractions

for contraction in contractions_dict.keys():
    escaped_contraction = re.escape(contraction) # Escape special characters (e.g.,
    apostrophes)    escaped_contractions.append(escaped_contraction) # Add to list

# Join the escaped contractions with '|' joined_contractions
= "|".join(escaped_contractions)

# Create a regex pattern with word boundaries (\b) contractions_pattern
= r'\b(' + joined_contractions + r')\b'

# Compile the regex compiled_pattern = re.compile(contractions_pattern,
flags=re.IGNORECASE)

# Define a function to replace contractions def
replace_contractions(text):
    # Function to handle each match found
    def replace_match(match):
        matched_word = match.group(0) # Extract matched contraction
        lower_matched_word = matched_word.lower() # Convert to lowercase
        expanded_form = contractions_dict[lower_matched_word] # Get full form from
        dictionary    return expanded_form # Return the expanded form

    # Apply regex substitution    expanded_text =
    compiled_pattern.sub(replace_match, text)

    return expanded_text # Return modified text

```

```
# Apply the function to a DataFrame column df["contractions_replaced"] =
df["slangs_replaced"].apply(replace_contractions)

# Display column content without truncation
pd.set_option('display.max_colwidth', None) # Set to None for unlimited width
print(df["contractions_replaced"])
```

### **g. Remove punctuations and special characters**

```
# Remove punctuations and special characters import
string

# Function to remove punctuation def
remove_punctuation(text):
    return text.translate(str.maketrans("", "", string.punctuation))

# Apply the function to the column df["punctuations_removed"] =
df["contractions_replaced"].apply(remove_punctuation)

# Display column content without truncation
pd.set_option('display.max_colwidth', None) # Set to None for unlimited width
print(df["punctuations_removed"])
```

### **h. Remove numbers #**

```
Remove numbers def
remove_numbers(text):
    return re.sub(r'\d+', '', text) # Removes all numeric characters

# Apply the function to the column df["numbers_removed"] =
df["punctuations_removed"].apply(remove_numbers)
```

```
# Display column content without truncation
pd.set_option('display.max_colwidth', None) # Set to None for unlimited width
print(df["numbers_removed"])
```

### **i. Correct spelling mistakes**

```
# Correct spelling mistakes from
autocorrect import Speller

# Initialize spell checker spell
= Speller(lang='en')

# Function to correct spelling def
correct_spelling(text):
    return spell(text) # Apply correction

# Apply the function to the column df["spelling_corrected"] =
df["numbers_removed"].apply(correct_spelling) # Display column
content without truncation pd.set_option('display.max_colwidth', None)
# Set to None for unlimited width print(df["spelling_corrected"])
```

### **j. Remove stopwords**

```
# Remove stopwords import nltk
from nltk.corpus import stopwords
```



```

# Download stopwords if not already downloaded nltk.download('stopwords')

# Define stopwords list stop_words =
set(stopwords.words('english'))

# Function to remove stopwords def
remove_stopwords(text):

    words = text.split() # Split text into words    filtered_words = [] # Create an
empty list to store words after stopword removal

    for word in words: # Loop through each word in the list of words

        lower_word = word.lower() # Convert the word to lowercase for uniform
comparison

        if lower_word not in stop_words: # Check if the lowercase word is NOT in the
stopwords list            filtered_words.append(word) # If it's not a stopword, add it to
the filtered list

    return " ".join(filtered_words) # Join words back into a sentence

# Apply the function to the column df["stopwords_removed"] =
df["spelling_corrected"].apply(remove_stopwords)

# Display column content without truncation

pd.set_option('display.max_colwidth', None) # Set to None for unlimited width

print(df["stopwords_removed"])

```

### **k. Stemming - - reduces words to their base root by chopping off suffixes**

# Stemming - reduces words to their base root by chopping off suffixes from

`nlTK.stem import PorterStemmer`

# Initialize the stemmer stemmer

`= PorterStemmer()`

# Function to apply stemming

`def stem_text(text):` if not

`isinstance(text, str):`

`return ""`

`words = text.split() stemmed_words = [stemmer.stem(word) for word in`  
`words]` # Apply stemming `return " ".join(stemmed_words)`

# Apply the function `df["stemmed_words"] =`

`df["stopwords_removed"].apply(stem_text)`

# Display column content without truncation

`pd.set_option('display.max_colwidth', None)` # Set to None for unlimited width

`print(df["stemmed_words"])`

### **l. Lemmatization - reduces words to their base dictionary form (lemma) import**

`nlTK`

```

# Download the required resources nltk.download('wordnet')
# For lemmatization nltk.download('omw-1.4')          #
WordNet lexical database
nltk.download('averaged_perceptron_tagger_eng') # For POS
tagging nltk.download('punkt_tab')                # For tokenization

# Lemmatization - reduces words to their base dictionary form (lemma)
from nltk.stem import WordNetLemmatizer from nltk.corpus import
wordnet from nltk.tokenize import word_tokenize from nltk import
pos_tag

# Initialize the lemmatizer lemmatizer
= WordNetLemmatizer()

# Function to map NLTK POS tags to WordNet POS tags def
get_wordnet_pos(nltk_tag):
    if nltk_tag.startswith('J'): # Adjective
return wordnet.ADJ    elif
nltk_tag.startswith('V'): # Verb
return wordnet.VERB    elif
nltk_tag.startswith('N'): # Noun
return wordnet.NOUN    elif
nltk_tag.startswith('R'): # Adverb
return wordnet.ADV

    else:

```

```

        return wordnet.NOUN # Default to noun

# Function to lemmatize text with POS tagging def
lemmatize_text(text):
    if not isinstance(text, str): # Ensure input is a string
        return ""

    words = word_tokenize(text) # Tokenize text into words
    pos_tags = pos_tag(words) # Get POS tags

    # Lemmatize each word with its correct POS tag    lemmatized_words =
    [lemmatizer.lemmatize(word, get_wordnet_pos(tag)) for word, tag in pos_tags]

    return " ".join(lemmatized_words) # Join words back into a sentence

# Apply the function to the column df["lemmatized"] =
df["stopwords_removed"].apply(lemmatize_text)

# Display column content without truncation
pd.set_option('display.max_colwidth', None) # Set to None for unlimited width
print(df["lemmatized"])

Step 3: Save the result to a file df.to_csv("Processed_Reviews.csv", index=False)
# Saves without the index column Putting it all together

```

```

import pandas as
pd import re import
emoji import string
import nltk

from bs4 import BeautifulSoup from
autocorrect import Speller from nltk.corpus
import stopwords, wordnet from nltk.stem
import WordNetLemmatizer from
nltk.tokenize import word_tokenize from
nltk import pos_tag

# Download required NLTK resources nltk.download('stopwords')
nltk.download('wordnet') # For lemmatization
nltk.download('omw-1.4') # WordNet lexical database
nltk.download('averaged_perceptron_tagger_eng') # For POS
tagging nltk.download('punkt_tab') # For tokenization

# Initialize tools spell = Speller(lang='en')
stop_words =
set(stopwords.words('english')) lemmatizer =
WordNetLemmatizer()

# Dictionary of slang words and their replacements slang_dict
= {
    "tbh": "to be honest",
    "omg": "oh my god",
    "lol": "laugh out loud",
    "idk": "I don't know",
    "brb": "be right back",
    "btw": "by the way",
    "imo": "in my opinion",
    "smh": "shaking my head",
    "fyi": "for your information",
    "np": "no problem",
    "ikr": "I know right",
    "asap": "as soon as possible",

```

```
"bff": "best friend forever",
"gg": "good game",
"hmu": "hit me up",
"rofl": "rolling on the floor laughing"
}
```

```
#          Contractions
           dictionary
```

```
contractions_dict = {
"wasn't": "was not",
  "isn't": "is not",
  "aren't": "are not",
  "weren't": "were not",
"doesn't": "does not",
  "don't": "do not",
  "didn't": "did not",
  "can't": "cannot",
  "couldn't": "could not",
  "shouldn't": "should not",
  "wouldn't": "would not",
  "won't": "will not",
  "haven't": "have not",
  "hasn't": "has not",
  "hadn't": "had not",
  "i'm": "i am",
  "you're": "you are",
  "he's": "he is",
  "she's": "she is",
  "it's": "it is",
  "we're": "we are",
  "they're": "they are",
  "i've": "i have",
  "you've": "you have",
  "we've": "we have",
  "they've": "they have",
  "i'd": "i would",
  "you'd": "you would",
```

```

    "he'd": "he would",
    "she'd": "she would",
    "we'd": "we would",
    "they'd": "they would",
    "i'll": "i will",
    "you'll": "you will",
    "he'll": "he will",
    "she'll": "she will",
    "we'll": "we will",
    "they'll": "they will",
    "let's": "let us",
    "that's": "that is",
    "who's": "who is",
    "what's": "what is",
    "where's": "where is",
    "when's": "when is",
    "why's": "why is"
}

# Remove any URLs that start with "http" or "www" from the text
def remove_urls(text):
    return re.sub(r'http\S+|www\S+', '', text)

# extracts only the text, removing all HTML tags
def remove_html(text):
    return BeautifulSoup(text, "html.parser").get_text()

# replace emoji with ''
def remove_emojis(text):
    return emoji.replace_emoji(text, replace='')

# Replace internet slang/chat words
def replace_slang(text):
    # Create a list of escaped slang words    escaped_slang_words = []
    # Empty list to store escaped slang words

```

```

    for word in slang_dict.keys():
        escaped_word = re.escape(word) # Ensure special characters are escaped
    escaped_slang_words.append(escaped_word) # Add to list

    # Join the words using '|'    slang_pattern = r'\b(' +
    '|'.join(escaped_slang_words) + r')\b'

    # Define a replacement function
    def replace_match(match):
        slang_word = match.group(0) # Extract matched slang word
    return slang_dict[slang_word.lower()] # Replace with full form

    # Use regex to replace slang words with full forms    replaced_text =
    re.sub(slang_pattern, replace_match, text, flags=re.IGNORECASE)

    return replaced_text

# Function to expand contractions # Build the regex pattern
for contractions escaped_contractions = [] # List to store
escaped contractions

for contraction in contractions_dict.keys():
    escaped_contraction = re.escape(contraction) # Escape special characters
    (e.g., apostrophes)    escaped_contractions.append(escaped_contraction) # Add
    to list

    # Join the escaped contractions with '|'
    joined_contractions = "|".join(escaped_contractions)

    # Create a regex pattern with word boundaries (\b)
    contractions_pattern = r'\b(' + joined_contractions + r')\b'

    # Compile the regex compiled_pattern = re.compile(contractions_pattern,
    flags=re.IGNORECASE)

```



```

# Define a function to replace contractions
def replace_contractions(text):
    # Function to handle each match found
    def replace_match(match):
        matched_word = match.group(0) # Extract matched contraction
        lower_matched_word = matched_word.lower() # Convert to lowercase
        expanded_form = contractions_dict[lower_matched_word] # Get full form from
        dictionary      return expanded_form # Return the expanded form

    # Apply regex substitution    expanded_text =
    compiled_pattern.sub(replace_match, text) return
    expanded_text # Return modified text

# Function to remove punctuation
def remove_punctuation(text):
    return text.translate(str.maketrans("", "", string.punctuation))

# Function to remove numbers
def remove_numbers(text):
    return re.sub(r'\d+', "", text)

# Function to correct spelling using AutoCorrect
def correct_spelling(text):
    return spell(text) # Apply correction

# Function to remove stopwords
def remove_stopwords(text):
    words = text.split()    filtered_words = [word for word in words if
    word.lower() not in stop_words]    return " ".join(filtered_words)

# Function to map NLTK POS tags to WordNet POS tags
def get_wordnet_pos(nltk_tag):
    if nltk_tag.startswith('J'): #
    Adjective      return wordnet.ADJ
    elif nltk_tag.startswith('V'): # Verb
    return wordnet.VERB    elif
    nltk_tag.startswith('N'): # Noun

```

```

return wordnet.NOUN    elif
nltk_tag.startswith('R'): # Adverb
return wordnet.ADV    else:
    return wordnet.NOUN # Default to noun

# Function to lemmatize text with POS tagging
def lemmatize_text(text):
    if not isinstance(text, str): # Ensure input is a string
        return ""

    words = word_tokenize(text) # Tokenize text into words
    pos_tags = pos_tag(words) # Get POS tags

    # Lemmatize each word with its correct POS tag    lemmatized_words =
    [lemmatizer.lemmatize(word, get_wordnet_pos(tag)) for word, tag in pos_tags]

    return " ".join(lemmatized_words) # Join words back into a sentence

# Function to tokenize text
def tokenize_text(text):
    if not isinstance(text, str): # Ensure the input is a string
        return []
    return word_tokenize(text) # Tokenize text into words

# Function to apply all preprocessing steps
def preprocess_text(text):
    text = text.lower()          # Step 1: Lowercasing    text =
    remove_urls(text)           # Step 2: Remove URLs    text =
    remove_html(text)           # Step 3: Remove HTML tags    text =
    remove_emojis(text)         # Step 4: Remove Emojis    text =
    replace_slang(text)         # Step 5: Replace Slang    text =
    replace_contractions(text)   # Step 6: Expand Contractions    text =
    remove_punctuation(text)     # Step 7: Remove Punctuation    text
    = remove_numbers(text)       # Step 8: Remove Numbers    text =
    correct_spelling(text)       # Step 9: Correct Spelling    text =
    remove_stopwords(text)       # Step 10: Remove Stopwords    text

```

---

```
= lemmatize_text(text)      # Step 11: Lemmatization    text =  
tokenize_text(text)        # Step 12: Tokenization    return text  
  
# Load dataset df = pd.read_csv("Review.csv") #  
Replace with your file  
  
# Apply preprocessing pipeline df["processed"] =  
df["Review"].apply(preprocess_text)  
  
# Save the cleaned dataset df.to_csv("Processed_Reviews2.csv",  
index=False)  
  
# Display the first few rows print(df[["Review",  
"processed"]].head())
```

## **Exercise**

1. Identify the issues with the "Review" column in the UNITENReview.csv file  
These are the issues:
  1. Mixed uppercase and lowercase letters (e.g., "UNITEN", "Im happy")
  2. Contractions (e.g., "I'm", "it's")
  3. Missing punctuation consistency
  4. Possible spelling mistakes
  5. Stopwords (e.g., "the", "is", "in", "and")
  6. Extra spaces and inconsistent formatting
  7. Informal writing style
2. Perform the necessary text pre-processing steps based on the identified issues
3. Save the result in a .csv file