You use autocorrect every day on your cell phone and computer. In this project:

- Get a word count given a corpus
- · Get a word probability in the corpus
- · Manipulate strings
- · Filter strings
- · Implement Minimum edit distance to compare strings and to help find the optimal path for the edits.

0.1 - Edit Distance

we will implement models that correct words that are 1 and 2 edit distances away.

· We say two words are n edit distance away from each other when we need n edits to change one word into another.

An edit could consist of one of the following options:

- Delete (remove a letter): 'hat' => 'at, ha, ht'
- Switch (swap 2 adjacent letters): 'eta' => 'eat, tea,...'
- Replace (change 1 letter to another): 'jat' => 'hat, rat, cat, mat, ...'
- Insert (add a letter): 'te' => 'the, ten, ate, ...'

You will be using the four methods above to implement an Auto-correct.

• To do so, we will compute probabilities that a certain word is correct given an input.

The goal of our spell check model is to compute the following probability:

$$P(c|w) = \frac{P(w|c) \times P(c)}{P(w)}$$
 (Eqn-1)

- Equation 1 says that the probability of a word being correct P(c|w) is equal to the probability of having a certain word w, given that it is correct P(w|c), multiplied by the probability of being correct in general P(C) divided by the probability of that word w appearing P(w) in general.
- To compute equation 1, you will first import a data set and then create all the probabilities that you need using that data set.

1 - Data Preprocessing

```
import re
from collections import Counter
import numpy as np
import pandas as pd
```

a - process_data

Implement the function process_data which

- 1) Reads in a corpus (text file)
- 2) Changes everything to lowercase
- 3) Returns a list of words.

```
def process_data(file_name):
    words = []
    with open(file_name) as fh:
        data=fh.read()

    data=data.lower()
    #Convert every word to lower case and return them in a list.
    words=re.findall(r'\w+', data)
    # we split the data into a list words while removing the new line character
    return words
```

```
word_l = process_data('shakespeare.txt')
vocab = set(word_l)  # this will be your new vocabulary
print(f"The first ten words in the text are: \n{word_l[0:10]}")
print(f"There are {len(vocab)} unique words in the vocabulary.")

The first ten words in the text are:
   ['this', 'is', 'the', '100th', 'etext', 'file', 'presented', 'by', 'project', 'gutenberg']
There are 23902 unique words in the vocabulary.
```

b - get_count

Implement a get_count function that returns a dictionary

- The dictionary's keys are words
- The value for each word is the number of times that word appears in the corpus.

```
def get_count(word_1):
    word_count_dict = {}
    word_count_dict=Counter(word_1)
    # counter for every unique word in the text document
    return word_count_dict

word_count_dict = get_count(word_1)
print(f"There are {len(word_count_dict)} key values pairs")
print(f"The count for the word 'thee' is {word_count_dict.get('thee',0)}")
    There are 23902 key values pairs
    The count for the word 'thee' is 3181
```

c - get_probs

Given the dictionary of word counts, compute the probability that each word will appear if randomly selected from the corpus of words.

$$P(w_i) = \frac{C(w_i)}{M} \tag{Eqn-2}$$

where

 $C(w_i)$ is the total number of times w_i appears in the corpus.

M is the total number of words in the corpus.

```
def get_probs(word_count_dict):
    probs = {}

    m=sum(word_count_dict.values())
    for word in word_count_dict:
        probs[word]=word_count_dict.get(word)/m
    # probability of each word occuring in the text document return probs

probs = get_probs(word_count_dict)
print(f"Length of probs is {len(probs)}")
print(f"P('thee') is {probs['thee']:.4f}")
    Length of probs is 23902
    P('thee') is 0.0034
```

2 - String Manipulations

Now that you have computed $P(w_i)$ for all the words in the corpus,we gave writtn e a few functions to manipulate strings so that we can edit the erroneous strings and return the right spellings of the words.WE will use functions:

- delete_letter: given a word, it returns all the possible strings that have one character removed.
- · switch_letter: given a word, it returns all the possible strings that have two adjacent letters switched.
- replace letter: given a word, it returns all the possible strings that have one character replaced by another different letter.
- insert_letter: given a word, it returns all the possible strings that have an additional character inserted.

a - delete letter

given a word it returns a list of strings with one character deleted.

For example, given the word nice, it would return the set: {'ice', 'nce', 'nic', 'nie'}.

b - switch_letter

It takes in a word and returns a list of all the possible switches of two letters that are adjacent to each other

• For example, given the word 'eta', it returns {'eat', 'tea'}, but does not return 'ate'.

c - replace_letter

we implement a function that takes in a word and returns a list of strings with one replaced letter from the original word.

```
def replace_letter(word, verbose=False):
    letters = 'abcdefghijklmnopqrstuvwxyz'
    split_1 = []
```

d- insert letter

Now we implement a function that takes in a word and returns a list with a letter inserted at every offset.

3 - Edit one letter

Now that we have implemented the string manipulations, we will create two functions that, given a string, will return all the possible single and double edits on that string. These will be edit_one_letter() and edit_two_letters().

```
def edit_one_letter(word, allow_switches = True):
    edit_one_set = set()

    edit_one_set.update(delete_letter(word))
    edit_one_set.update(switch_letter(word))
    edit_one_set.update(insert_letter(word))
    edit_one_set.update(replace_letter(word))

    return set(edit_one_set)

tmp_word = "at"
tmp_edit_one_set = edit_one_letter(tmp_word)

tmp_edit_one_l = sorted(list(tmp_edit_one_set))

print(f"input word {tmp_word} \nedit_one_l \n{tmp_edit_one_l}\n")
```

```
input word at
edit_one_1
['a', 'aa', 'aat', 'ab', 'abt', 'ac', 'act', 'ad', 'adt', 'ae', 'aft', 'aft', 'ag', 'agt', 'ah', 'aht', 'ai', 'ait', 'aj', 'ajt',
```

a - Edit Two Letters

Exercise 9 - edit_two_letters

wegeneralize this to implement to get two edits on a word. To do so, we would have to get all the possible edits on a single word and then for each modified word, we would have to modify it again.

```
def edit_two_letters(word, allow_switches = True):
    edit_two_set = set()
    edit_one_set=set()
    edit_one_set.update(delete_letter(word))
    edit_one_set.update(switch_letter(word))
    edit_one_set.update(insert_letter(word))
    edit_one_set.update(replace_letter(word))
    L=list(edit_one_set)
    for i in L:
        edit_two_set.update(insert_letter(i))
        edit two set.update(replace letter(i))
        edit two set.update(switch letter(i))
        edit_two_set.update(delete_letter(i))
    return set(edit_two_set)
tmp edit two set = edit two letters("a")
tmp_edit_two_l = sorted(list(tmp_edit_two_set))
print(f"Number of strings with edit distance of two: {len(tmp_edit_two_1)}")
print(f"First 10 strings {tmp_edit_two_l[:10]}")
print(f"Last 10 strings {tmp_edit_two_l[-10:]}")
     Number of strings with edit distance of two: 2654
     First 10 strings ['', 'a', 'aa', 'aaa', 'aab', 'aac', 'aad', 'aae', 'aaf', 'aag']
Last 10 strings ['zv', 'zva', 'zw', 'zwa', 'zx', 'zxa', 'zy', 'zya', 'zz', 'zza']
```

4 - Suggest Spelling Suggestions

Now you will use your edit_two_letters function to get a set of all the possible 2 edits on your word. You will then use those strings to get the most probable word you meant to type a.k.a your typing suggestion.

```
def get_corrections(word, probs, vocab, n=4, verbose = False):
   suggestions = []
   n_best = []
   ### START CODE HERE ###
   #Step 1: create suggestions as described above
   one_edit=list(edit_one_letter(word))
   two_edit=list(edit_two_letters(word))
   list1=set()
   for i in vocab:
       if(i ==word):
           list1.add(i)
           break
   list2=set()
   for i in one_edit:
       if i in vocab:
           list2.add(i)
```

```
list3=set()
   for i in two_edit:
       if i in vocab:
           list3.add(i)
   suggestions=list1 or list2 or list3
   best_words={}
   for i in list(suggestions):
       if i in probs:
           best_words[i]=probs.get(i)
           best words[i]=0
   keys = list(best_words.keys())
   values = list(best_words.values())
   sorted_value_index = np.argsort(values)[::-1]
   sorted_dict = {keys[i]: values[i] for i in sorted_value_index}
   if (n>len(sorted_dict)):
        for i in sorted_dict:
           n_best.append((i,sorted_dict.get(i)))
   else:
       d1= dict(list(sorted dict.items())[0:n])
        for i in d1:
           n_best.append((i,d1.get(i)))
   if verbose: print("entered word = ", word, "\nsuggestions = ", suggestions)
   return n_best
my_word = 'dys'
tmp corrections = get corrections(my word, probs, vocab, 2, verbose=True)
for i, word_prob in enumerate(tmp_corrections):
   print(f"word {i}: {word_prob[0]}, probability {word_prob[1]:.6f}")
print(f"data type of corrections {type(tmp_corrections)}")
    entered word = dvs
    suggestions = {'dye', 'dy', 'dis', 'des', 'days'}
    word 0: days, probability 0.000225
    word 1: dis, probability 0.000005
    data type of corrections <class 'list'>
```

4a - Minimum Edit Distance

Now that we have implemented our auto-correct, we need to evaluate the similarity between two strings. For example: 'waht' and 'what'

```
def min_edit_distance(source, target, ins_cost = 1, del_cost = 1, rep_cost = 2):
   m = len(source)
   n = len(target)
   #initialize cost matrix with zeros and dimensions (m+1,n+1)
   D = np.zeros((m+1, n+1), dtype=int)
   \mbox{\tt\#} Fill in column 0, from row 1 to row m, both inclusive
    for row in range(1,m+1):
        D[row,0] = D[row-1,0]+del_cost
   # Fill in row 0, for all columns from 1 to n, both inclusive
    for col in range(1,n+1):
        D[0,col] = D[0,col-1] + ins_cost
    # Loop through row 1 to row m, both inclusive
    for row in range(1,m+1):
        # Loop through column 1 to column n, both inclusive
        for col in range(1,n+1):
            \# Intialize r_cost to the 'replace' cost that is passed into this function
            r_cost = rep_cost
```

```
# Check to see if source character at the previous row
           # matches the target character at the previous column,
           if (source[row-1]==target[col-1]):
               # Update the replacement cost to 0 if source and target are the same
               r_{cost} = 0
           # Update the cost at row, col based on previous entries in the cost matrix
           \label{eq:decomposition} D[row,col] = \min(D[row-1,col]+del\_cost,D[row,col-1]+ins\_cost,D[row-1,col-1]+r\_cost)
   \# Set the minimum edit distance with the cost found at row m, column n
   med = D[m,n]
   return D, med
source = 'play'
target = 'stay'
matrix, min_edits = min_edit_distance(source, target)
print("minimum edits: ",min_edits, "\n")
idx = list('#' + source)
cols = list('#' + target)
df = pd.DataFrame(matrix, index=idx, columns= cols)
print(df)
minimum edits: 4
       # s t a y
     # 0 1 2 3 4
    p 1 2 3 4 5
    1 2 3 4 5 6
    a 3 4 5 4 5
    y 4 5 6 5 4
                                                               Code
                                                                           Text
```

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