# CSCI 544 - Applied Natural language processing Assignment-3

Name: Anne Sai Venkata Naga Saketh

USC Email: annes@usc.edu

USC ID: 3725520208

```
In [1]: # Importing the required libraries
import csv
# Library for Hidden Markov Models (HMM)
import hmmlearn
# Library for creating defaultdicts, a subclass of dict
from collections import defaultdict, Counter
# Library providing functions that map Python operators to corresponding
import operator
# Importing the JSON library to read and write the json files
import json
```

## **Task 1: Vocabulary Creation**

```
In [2]: # Read the training data
        with open('./data/train', 'r') as file:
            train_data = file.readlines()
        # Print first two sentences for demonstration
        print("Train data:")
        for sentence in train data[:2]:
            print(sentence)
        # Read the testing data
        with open('./data/test', 'r') as file:
            test_data = file.readlines()
        # Print first two sentences for demonstration
        print("Test data:")
        for sentence in test_data[:2]:
            print(sentence)
        # Read the testing data
        with open('./data/dev', 'r') as file:
            dev data = file.readlines()
        # Print first two sentences for demonstration
        print("Dev data:")
        for sentence in dev data[:2]:
            print(sentence)
```

```
Vinken NNP
        Test data:
                Influential
                members
        Dev data:
                The
                        DТ
               Arizona NNP
In [3]: # Set the threshold for word frequency
        frequency_threshold = 3
        # Initialize dictionaries to store word and tag frequencies, and count se
        w_f, t_f, s_c = defaultdict(int), defaultdict(int), 0
        # Initialize a list to store file data with start tokens
        f_d = ["<s>"]
        # Iterate through each line in the training data
        for line in train_data:
            # Check if the line is empty, indicating end of a sentence
            if line.strip() == "":
                 # Add start token to file data
                f_d.append("<s>")
                # Increment sentence count
                s c += 1
            else:
                speech = line.strip().split("\t")
                # Check if the line has three parts (word, tag, frequency)
                # Skip this line if it doesn't have three parts
                if len(speech) != 3:
                    continue
                # Extract word and tag
                w, t = speech[1], speech[2]
                # Increment word frequency
                w_f[w] += 1
                # Increment tag frequency
                t f[t] += 1
                # Add line to file data
                f d.append(line.strip())
        # Filter vocabulary based on frequency threshold
```

Train data:

Pierre NNP

```
vocabulary = {w: f for w, f in w_f.items() if f >= frequency_threshold}

# Add <unk> token for infrequent words
vocabulary["<unk>"] = sum(f for w, f in w_f.items() if f < frequency_thre

# Write vocabulary to a file
with open("vocab.txt", "w") as vocabulary_file:
    # Write <unk> token with frequency
    vocabulary_file.write("<unk>\t0\t{}\n".format(vocabulary["<unk>"]))

# Write each word with its index and frequency to the file, sorted by
    sorted_vocabulary = sorted(vocabulary.items(), key=lambda x: x[1], re
    for idx, (w, f) in enumerate(sorted_vocabulary, start=1):
        vocabulary_file.write("{}\t{}\t{}\n".format(w, idx, f))

# Print vocabulary size after replacement and total occurrences of '<unk>
print(f"The Vocabulary size after replacement of the least occurring occu
print(f"Total occurrences of '<unk>' token are: {vocabulary['<unk>']}")
```

The Vocabulary size after replacement of the least occurring occurrences by the threshold: 16920
Total occurrences of '<unk>' token are: 32537

## Task 2: Model Learning

```
In [4]: # Initialize dictionaries to store emission and transition probabilities
        e probabilities, t probabilities = defaultdict(int), defaultdict(int)
        # Initialize previous tag variable with start token
        prev_tag = "<s>"
        # Iterate through each line in the file data
        for line in f d:
            # Check if the line is a start token
            if line == "<s>":
                prev tag = "<s>" # Reset previous tag to start token
                continue
            # Split the line by tab
            speech = line.split("\t")
            w, cur tag = speech[1], speech[2]
            # Replace infrequent words with <unk>
            w = w if w in vocabulary else "<unk>"
            # Increment emission probability
            e_probabilities[(cur_tag, w)] += 1
            # Update transition probabilities based on previous and current tags
            if prev tag != "<s>":
                t probabilities[(prev tag, cur tag)] += 1
            else:
                 # Increment start transition probability
                t_probabilities[("start", cur_tag)] += 1
```

```
prev tag = cur tag
# Normalize emission probabilities by tag frequency
for key in e probabilities:
    e_probabilities[key] /= t_f[key[0]]
# Normalize transition probabilities by tag frequency or sentence count
for key in t probabilities:
    if key[0] == "start":
        t_probabilities[key] /= s_c
    else:
        t_probabilities[key] /= t_f[key[0]]
# Print number of transition and emission parameters
print(f"Number of transition parameters in the hmm.json file are: {len(e
print(f"Number of emission parameters in the hmm.json file are: {len(t pr
# Convert keys to string for emission and transition probabilities
e k = {f"({tag}, {word})": prob for (tag, word), prob in e probabilities.i
t_k = {f"({prev_tag}, {next_tag})": prob for (prev_tag, next_tag), prob in
# Create an HMM model dictionary
model = {"Transition": t_k, "Emission": e_k}
# Save HMM model to a JSON file
with open('hmm.json', 'w') as json file:
    json.dump(model, json_file, indent=4)
```

Number of transition parameters in the hmm.json file are: 23373 Number of emission parameters in the hmm.json file are: 1392

### Task 3: Greedy Decoding with HMM

```
In [5]: def load_hmm_from_json(file_path):
    with open(file_path, 'r') as f:
        data = json.load(f)

# Create a dictionary containing HMM parameters with keys "Transition
    return {"Transition": data["Transition"], "Emission": data["Emission"

# Path to the JSON file containing HMM parameters
hmm_file_path = 'hmm.json'

# Load HMM parameters from the JSON file
model = load_hmm_from_json(hmm_file_path)
```

#### Testing on the Dev Data

```
In [6]: # Specify the output file name
  output_file_name = "greedy_dev.out"

# Initialize a list to store data to be written
  w_d = []
```

```
# Initialize previous tag variable with start token
prev tag = "start"
# Open the development data file for reading and output file for writing
with open(output_file_name, "w") as output_file:
    # Iterate through each line in the development data
    for line in dev data:
       w = line.split("\t")
        # Check if the line contains only one element, indicating start o
        if len(w) == 1:
            # Reset previous tag to start token
            prev tag = "start"
            # Append line to write data
           w_d.append(line)
            continue
        else:
            # Extract index and current word
            idx, cur_word = w[0].strip(), w[1].strip()
            # Replace infrequent words with <unk>
            if cur word not in vocabulary:
                cur word = "<unk>"
            # Initialize probability and temporary tag variables
            prob val, temp tag = 0, ""
            # Iterate through each tag in the tag frequency dictionary
            for tag_iter in t_f:
                # Check emission and transition probabilities for the cur
                e_c = (tag_iter, cur_word)
                emission_prob_value = e_probabilities.get(e_c, 0)
                t c = (prev tag, tag iter)
                transition prob value = t probabilities.get(t c, 0)
                current prob val = emission prob value * transition prob
                # Update probability and temporary tag if current probabi
                if current prob val >= prob val:
                    prob val, temp tag = current prob val, tag iter
            # Update previous tag with temporary tag
            prev_tag = temp_tag
            # Construct the line to be written
            cur line = f"{idx}\t{cur word}\t{prev tag}\n"
            # Append line to write data
            w_d.append(cur_line)
    # Write all the data to the output file
    output file.writelines(w d)
# Printing the output
```

```
print("Output file: greedy_dev.out has been generated successfully")
Output file: greedy_dev.out has been generated successfully
In [7]: !python eval.py -p greedy_dev.out -g ./data/dev
total: 131768, correct: 122390, accuracy: 92.88%
```

#### Testing on the Test Data

```
In [8]: # Specify the output file name
        output file name = "greedy.out"
        # Initialize a list to store data to be written
        wd = []
        # Initialize previous tag variable with start token
        prev_tag = "start"
        # Open the development data file for reading and output file for writing
        with open(output file name, "w") as output file:
            # Iterate through each line in the development data
            for line in test data:
                w = line.split("\t")
                # Check if the line contains only one element, indicating start o
                if len(w) == 1:
                    # Reset previous tag to start token
                    prev_tag = "start"
                    # Append line to write data
                    w d.append(line)
                    continue
                else:
                    # Extract index and current word
                    idx, cur word = w[0].strip(), w[1].strip()
                    # Replace infrequent words with <unk>
                    if cur word not in vocabulary:
                        cur_word = "<unk>"
                    # Initialize probability and temporary tag variables
                    prob_val, temp_tag = 0, ""
                    # Iterate through each tag in the tag frequency dictionary
                    for tag_iter in t_f:
                         # Check emission and transition probabilities for the cur
                        e c = (tag iter, cur word)
                        emission prob value = e probabilities.get(e c, 0)
                        t c = (prev tag, tag iter)
                        transition_prob_value = t_probabilities.get(t_c, 0)
                        current prob_val = emission_prob_value * transition_prob_
```

Output file: greedy.out has been generated successfully

## Task 4: Viterbi Decoding with HMM

```
In [9]: def viterbi(o w, s l, e probabilities, t probabilities):
            # Calculate the total number of observations and states
            tot obs = len(o w)
            tot_s = len(s_l)
            # Initialize the Viterbi matrix and backpointers matrix
            v_m = [[0 for _ in range(tot_s)] for _ in range(tot_obs)]
            backtracking = [[0 for _ in range(tot_s)] for _ in range(tot obs)]
            # Initialization step
            for ind in range(tot_s):
                # Calculate transition and emission probabilities for the first o
                t_prob = t_probabilities.get(('start', s_l[ind]), 1e-10)
                e prob = e probabilities.get((s l[ind], o w[0]), 1e-10)
                # Initialize the first column of the Viterbi matrix and backpoint
                v_m[0][ind] = t_prob * e_prob
                backtracking[0][ind] = 0
            # Recursion step
            for t_s in range(1, tot_obs):
                for ind in range(tot_s):
                    # Calculate the maximum probability and the corresponding pre
                    m p, optimal state = max(
                        (v_m[t_s-1][prev_state] * t_probabilities.get((s_l[prev_s
                        for prev_state in range(tot_s))
                    # Update the current cell in the Viterbi matrix and backpoint
                    v_m[t_s][ind] = m_p
                    backtracking[t s][ind] = optimal state
            # Termination step
            final_t_s = tot_obs - 1
            # Find the final state with the highest probability
            optimal_state = max(range(tot_s), key=lambda s: v_m[final_t_s][s])
            # Path backtracking
            minimal_route = [optimal_state]
            for t_s in range(tot_obs - 1, 0, -1):
                # Insert the best previous state at the beginning of the optimal
                minimal route.insert(0, backtracking[t s][minimal route[0]])
            # Convert state indices to state labels and return the optimal path
            return [s_l[region] for region in minimal_route]
        # Assuming emission probs is a dictionary where keys are tuples (tag, wor
        # Extract unique tags from the keys of emission probs
        states = set(tag for tag, _ in e_probabilities.keys())
```

```
In [10]: # If 'states' was originally defined as a set, convert it to a list
         states = list(states)
         # Determine the output file based on the data type
         output name = "viterbi dev.out"
         viterbi_result = []
         idx_data, cw_data, o_w = [], [], []
         for line in dev_data:
             # Sentence boundary
             if len(line.strip()) == 0:
                 if o w:
                     # Perform Viterbi decoding on observed words
                     wots = viterbi(o_w, states, e_probabilities, t_probabilities)
                     # Append decoded tags along with word indices and original wo
                     viterbi_result.extend(f"{idx}\t{word}\t{tag}\n" for idx, word
                 # Reset for next sentence
                 o_w, idx_data, cw_data = [], [], []
                 viterbi result.append("\n")
                 continue
             idx, word = line.strip().split("\t")[:2]
             idx data.append(idx)
             cw_data.append(word)
             # Replace out-of-vocabulary words with <unk>
             o_w.append(word if word in vocabulary else "<unk>")
         # Write Viterbi output to file
         with open(output_name, "w") as viterbi_file:
             viterbi file.writelines(viterbi result)
         print("The output has been written into the file viterbi_dev.out")
```

The output has been written into the file viterbi dev.out

```
In [11]: !python eval.py -p viterbi_dev.out -g ./data/dev

'1\tThat\tDT' '38\t.\t.' 131751
total: 131751, correct: 124384, accuracy: 94.41%
```

```
In [12]: # If 'states' was originally defined as a set, convert it to a list
         states = list(states)
         # Determine the output file based on the data type
         output name = "viterbi.out"
         viterbi_result = []
         idx_data, cw_data, o_w = [], [], []
         for line in test_data:
             # Sentence boundary
             if len(line.strip()) == 0:
                 if o_w:
                      # Perform Viterbi decoding on observed words
                     wots = viterbi(o_w, states, e_probabilities, t_probabilities)
                      # Append decoded tags along with word indices and original wo
                      viterbi_result.extend(f"{idx}\t{word}\t{tag}\n" for idx, word
                 # Reset for next sentence
                 o_w, idx_data, cw_data = [], [], []
                 viterbi result.append("\n")
                 continue
             idx, word = line.strip().split("\t")[:2]
             idx data.append(idx)
             cw_data.append(word)
             # Replace out-of-vocabulary words with <unk>
             o_w.append(word if word in vocabulary else "<unk>")
         # Write Viterbi output to file
         with open(output_name, "w") as viterbi_file:
             viterbi_file.writelines(viterbi_result)
         print("The output has been written into the file viterbi.out")
```

The output has been written into the file viterbi.out