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```
import pandas as pd

df = pd.read_csv('/content/Twitter_Data.csv')
display(df.head())
```

	clean_text	category	grid icon
0	when modi promised “minimum government maximum...	-1.0	
1	talk all the nonsense and continue all the dra...	0.0	
2	what did just say vote for modi welcome bjp t...	1.0	
3	asking his supporters prefix chowkidar their n...	1.0	
4	answer who among these the most powerful world...	1.0	

```
import re

def preprocess_text(text):
    # Ensure the input is a string
    if not isinstance(text, str):
        return '' # Or handle as appropriate, e.g., str(text)
    # Remove URLs
    text = re.sub(r'http\S+|www\S+|https\S+', '', text, flags=re.MULTILINE)
    # Remove mentions
    text = re.sub(r'@\w+', '', text)
    return text

# Convert the column to string type and fill NaN values before applying the function
df['clean_text'] = df['clean_text'].astype(str).apply(preprocess_text)
display(df.head())
```

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	clean_text	category
0	when modi promised “minimum government maximum... talk all the nonsense and continue all the dra...	-1.0
1	what did just say vote for modi welcome bjp t...	0.0
2	asking his supporters prefix chowkidar their n...	1.0
3	answer who among these the most powerful world...	1.0

```
import nltk

# Download necessary NLTK data
nltk.download('punkt')
nltk.download('averaged_perceptron_tagger')

[nltk_data] Downloading package punkt to /root/nltk_data...
[nltk_data]  Unzipping tokenizers/punkt.zip.
[nltk_data] Downloading package averaged_perceptron_tagger to
[nltk_data]      /root/nltk_data...
[nltk_data]  Unzipping taggers/averaged_perceptron_tagger.zip.
True
```

```
import nltk

# Download necessary NLTK data
nltk.download('punkt')
nltk.download('averaged_perceptron_tagger')
nltk.download('punkt_tab')
nltk.download('averaged_perceptron_tagger_eng')

[nltk_data] Downloading package punkt to /root/nltk_data...
[nltk_data]  Package punkt is already up-to-date!
[nltk_data] Downloading package averaged_perceptron_tagger to
[nltk_data]      /root/nltk_data...
[nltk_data]  Package averaged_perceptron_tagger is already up-to-
[nltk_data]      date!
[nltk_data] Downloading package punkt_tab to /root/nltk_data...
```

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[nltk_data] Package punkt_tab is already up-to-date!
[nltk_data] Downloading package averaged_perceptron_tagger_eng to
[nltk_data]     /root/nltk_data...
[nltk_data] Unzipping taggers/averaged_perceptron_tagger_eng.zip.
True
```

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```
from nltk.tokenize import word_tokenize
from nltk.tag import pos_tag

def pos_tag_text(text):
    if not isinstance(text, str):
        return [] # Return empty list for non-string inputs
    tokens = word_tokenize(text)
    return pos_tag(tokens)

df['pos_tags'] = df['clean_text'].apply(pos_tag_text)
display(df.head())
```

	clean_text	category	pos_tags
0	when modi promised “minimum government maximum... talk all the nonsense and continue all the dra...	-1.0 0.0	[(when, WRB), (modi, NN), (promised, VBD), (“,... [(talk, NN), (all, PDT), (the, DT), (nonsense,...
2	what did just say vote for modi welcome bjp t...	1.0	[(what, WP), (did, VBD), (just, RB), (say, VB)...
3	asking his supporters prefix chowkidar their n...	1.0	[(asking, VBG), (his, PRP\$), (supporters, NNS)...
4	answer who among these the most powerful world...	1.0	[(answer, NN), (who, WP), (among, IN), (these,...

Start coding or generate with AI.

```
from collections import defaultdict

# Initialize dictionaries for counts
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tag_counts = defaultdict(lambda: defaultdict(int))
word_tag_counts = defaultdict(lambda: defaultdict(int))
trans_counts = defaultdict(lambda: defaultdict(int))

# Iterate through each tweet's POS tags to build counts
for tags_list in df['pos_tags']:
    if not tags_list:
        continue # Skip empty tag lists

    # Count initial tags (for sentence start probability, if needed, though not explicitly asked here)
    # We'll focus on emission and transition for now.

    previous_tag = None
    for word, current_tag in tags_list:
        # Emission counts: P(word|tag)
        word_tag_counts[current_tag][word] += 1
        tag_counts[current_tag]['total'] += 1 # Total count for the tag

        # Transition counts: P(tag_i | tag_i-1)
        if previous_tag:
            trans_counts[previous_tag][current_tag] += 1
        previous_tag = current_tag

    # Calculate probabilities (without smoothing for now)
    emission_probabilities = defaultdict(lambda: defaultdict(float))
    transition_probabilities = defaultdict(lambda: defaultdict(float))

    # Calculate Emission Probabilities P(word|tag)
    for tag, word_counts in word_tag_counts.items():
        total_words_for_tag = tag_counts[tag]['total']
        if total_words_for_tag > 0:
            for word, count in word_counts.items():
                emission_probabilities[tag][word] = count / total_words_for_tag

    # Calculate Transition Probabilities P(tag_i | tag_i-1)
    for prev_tag, next_tag_counts in trans_counts.items():
        total_next_tags_from_prev = sum(next_tag_counts.values())
        if total_next_tags_from_prev > 0:
            for next_tag, count in next_tag_counts.items():
                transition_probabilities[prev_tag][next_tag] = count / total_next_tags_from_prev
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transition_probabilities[prev_tag][next_tag] = count / total_next_tags from prev
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print("--- Sample Emission Probabilities (P(word|tag)) ---")
# Display a few examples for common tags
example_tags = ['NN', 'VB', 'JJ']
for tag in example_tags:
    if tag in emission_probabilities:
        print(f"Top words for tag '{tag}': {sorted(emission_probabilities[tag].items(), key=lambda x: x[1], reverse=True)[:5]}")
    else:
        print(f"No emission probabilities found for tag '{tag}'")

print("\n--- Sample Transition Probabilities (P(tag_i | tag_i-1)) ---")
# Display a few examples for common tags
example_prev_tags = ['NN', 'DT', 'VB']
for prev_tag in example_prev_tags:
    if prev_tag in transition_probabilities:
        print(f"Top next tags for previous tag '{prev_tag}': {sorted(transition_probabilities[prev_tag].items(), key=lambda x: x[1], reverse=True)[:5]}")
    else:
        print(f"No transition probabilities found for tag '{prev_tag}'")

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--- Sample Emission Probabilities (P(word|tag)) ---
Top words for tag 'NN': [('modi', 0.10320215229946092), ('india', 0.016277776092079577), ('congress', 0.0107574)
Top words for tag 'VB': [('have', 0.04313074462485322), ('modi', 0.04215896667611451), ('get', 0.03151799813742)
Top words for tag 'JJ': [('modi', 0.04614891259984841), ('indian', 0.017589750928914825), ('india', 0.014936844)

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--- Sample Transition Probabilities (P(tag_i | tag_i-1)) ---
Top next tags for previous tag 'NN': [('NN', 0.3595183710397132), ('IN', 0.09323072283437303), ('VBD', 0.060803)
Top next tags for previous tag 'DT': [('NN', 0.5083480729583272), ('JJ', 0.19831461307076145), ('NNS', 0.117847)
Top next tags for previous tag 'VB': [('NN', 0.19055963294986075), ('DT', 0.14863160389736818), ('JJ', 0.118715)

```

```

index = 5
selected_tweet_text = df.loc[index, 'clean_text']
original_pos_tags = df.loc[index, 'pos_tags']

print(f"Selected Tweet Text: {selected_tweet_text}")
print(f"Original POS Tags: {original_pos_tags}")

```

Selected Tweet Text: kiya tho refresh maarkefir comment karo
 Original POS Tags: [('kiya', 'NN'), ('tho', 'NN'), ('refresh', 'JJ'), ('maarkefir', 'NN'), ('comment', 'NN')], (

```
from nltk.tokenize import word_tokenize

# 1. Tokenize the selected_tweet_text
tweet_words = word_tokenize(selected_tweet_text)

# 2. Create a list of all unique possible POS tags
all_pos_tags = list(emission_probabilities.keys())

# 3. Define a small constant value for smoothing
default_prob = 1e-6

print(f"Tokenized Tweet Words: {tweet_words}")
print(f"Total Unique POS Tags: {len(all_pos_tags)}")
print(f"Sample Unique POS Tags: {all_pos_tags[:10]}")
print(f"Default Smoothing Probability: {default_prob}")
```

Tokenized Tweet Words: ['kiya', 'tho', 'refresh', 'maarkefir', 'comment', 'karo']
 Total Unique POS Tags: 38
 Sample Unique POS Tags: ['WRB', 'NN', 'VBD', 'NNP', 'JJ', 'PRP', 'VB', 'DT', 'VBG', 'VBZ']
 Default Smoothing Probability: 1e-06

```
def viterbi(words, all_tags, emission_prob, transition_prob, default_prob):
    # Initialize Viterbi path and score matrices
    # Viterbi matrix: stores the maximum probability of a tag sequence ending at a specific tag for a specific word
    viterbi_matrix = [[{}]] # Stores probability
    # Path matrix: stores the backpointer (previous tag) for reconstructing the best path
    path_matrix = [[{}]] # Stores backpointer

    # Initialize for the first word
    first_word = words[0]
    for tag in all_tags:
        # P(tag | start) - assuming uniform probability for starting tags or a start tag distribution if available
        # For simplicity, we'll assume a uniform start probability for all tags, or 1 if no 'start' tag is provided
        # and focus on emission probability for the first word.
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# If a start tag distribution is available, it would be used here.  
# For now, let's treat the first word's tag probability as just its emis: Toggle Gemini  
  
# Emission probability for the first word and current tag  
emit_prob = emission_prob.get(tag, {}).get(first_word, default_prob)  
viterbi_matrix[0][tag] = emit_prob  
path_matrix[0][tag] = None # No previous tag for the first word  
  
# Iterate through the rest of the words  
for t in range(1, len(words)):  
    viterbi_matrix.append({}) # New dict for current word's probabilities  
    path_matrix.append({}) # New dict for current word's paths  
    current_word = words[t]  
  
    for current_tag in all_tags:  
        max_prob = 0  
        best_prev_tag = None  
  
        # Emission probability for the current word and current tag  
        emit_prob = emission_prob.get(current_tag, {}).get(current_word, default_prob)  
  
        for prev_tag in all_tags:  
            # Transition probability from previous tag to current tag  
            trans_prob = transition_prob.get(prev_tag, {}).get(current_tag, default_prob)  
  
            # Calculate the probability of the path ending with prev_tag -> current_tag  
            # multiplied by emission of current_word by current_tag  
            current_path_prob = viterbi_matrix[t-1].get(prev_tag, 0) * trans_prob * emit_prob  
  
            if current_path_prob > max_prob:  
                max_prob = current_path_prob  
                best_prev_tag = prev_tag  
  
        viterbi_matrix[t][current_tag] = max_prob  
        path_matrix[t][current_tag] = best_prev_tag  
  
# Backtrack to find the best path  
best_path = []  
# Find the tag with the highest probability for the last word
```

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last_word_idx = len(words) - 1
if not viterbi_matrix[last_word_idx]: # Handle case where no probabilities were found
    return []

max_final_prob = 0
last_best_tag = None
for tag, prob in viterbi_matrix[last_word_idx].items():
    if prob > max_final_prob:
        max_final_prob = prob
        last_best_tag = tag

if last_best_tag is None: # If no tag was found, return empty
    return []

best_path.append(last_best_tag)
current_best_tag = last_best_tag

# Iterate backwards through the path_matrix
for t in range(len(words) - 1, 0, -1):
    current_best_tag = path_matrix[t][current_best_tag]
    best_path.insert(0, current_best_tag) # Insert at the beginning to maintain order

return list(zip(words, best_path))

# Apply the Viterbi algorithm
predicted_pos_tags = viterbi(tweet_words, all_pos_tags, emission_probabilities, transition_probabilities, default_tag='NN')

```

```

print(f"Selected Tweet: {selected_tweet_text}")
print(f"Tokenized Words: {tweet_words}")
print(f"Predicted POS Tags: {predicted_pos_tags}")
print(f"Original POS Tags: {original_pos_tags}")

```

Selected Tweet: kiya tho refresh maarkefir comment karo
 Tokenized Words: ['kiya', 'tho', 'refresh', 'maarkefir', 'comment', 'karo']
 Predicted POS Tags: [('kiya', 'POS'), ('tho', 'NN'), ('refresh', 'VB'), ('maarkefir', 'JJ'), ('comment', 'NN'),
 Original POS Tags: [('kiya', 'NN'), ('tho', 'NN'), ('refresh', 'JJ'), ('maarkefir', 'JJ'), ('comment', 'NN'), (

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