

CORPUS

```

D1 = "I enjoy reading books on history"
D2 = "The pizza was cold and tasteless"
D3 = "She received a promotion at work"
D4 = "Heavy rain caused flooding downtown"
D5 = "The movie was surprisingly good"
D6 = "My laptop battery dies quickly"
D7 = "The concert tickets sold out fast"
D8 = "He adopted a puppy from the shelter"
D9 = "The exam was more difficult than expected"
D10 = "Scientists discovered a new species of frog"

```

UNI-GRAM COUNTS

```

import collections

# Combine the text from the text documents
combined_text = f"{D1} {D2} {D3} {D4} {D5} {D6} {D7} {D8} {D9} {D10}"

# Tokenize the combined text into words and convert to lowercase
words = combined_text.lower().split()

# Calculate unigram counts
unigram_counts = collections.Counter(words)

# Print the unigram counts
print("Unigram Counts:")
for word, count in unigram_counts.most_common():
    print(f"{word}: {count}")
#Vocabulary size is length of unigrams
V=len(unigram_counts)
print("Vocabulary Size=",V)

```

```

Unigram Counts:
the: 5
was: 3
a: 3
i: 1
enjoy: 1
reading: 1
books: 1
on: 1
history: 1
pizza: 1
cold: 1
and: 1
tasteless: 1
she: 1
received: 1
promotion: 1
at: 1
work: 1
heavy: 1
rain: 1
caused: 1
flooding: 1
downtown: 1
movie: 1
surprisingly: 1
good: 1
my: 1
laptop: 1
battery: 1
dies: 1
quickly: 1
concert: 1
tickets: 1
sold: 1
out: 1
fast: 1
he: 1
adopted: 1
puppy: 1
from: 1
shelter: 1

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exam: 1
more: 1
difficult: 1
than: 1
expected: 1
scientists: 1
discovered: 1
new: 1
species: 1
of: 1
frog: 1
Vocabulary Size= 52

```

BI-GRAM COUNTS

```

import collections
combined_text = f"{D1} {D2} {D3} {D4} {D5} {D6} {D7} {D8} {D9} {D10}"
words = combined_text.lower().split()
# Generate bigrams
bigrams = []
for i in range(len(words) - 1):
    bigrams.append((words[i], words[i+1]))
# Calculate bigram counts
bigram_counts = collections.Counter(bigrams)
# Print the bigram counts
print("\nBigram Counts:")
for bigram, count in bigram_counts.most_common():
    print(f"{bigram[0]} {bigram[1]}: {count}")

```

```

enjoy reading: 1
reading books: 1
books on: 1
on history: 1
history the: 1
the pizza: 1
pizza was: 1
was cold: 1
cold and: 1
and tasteless: 1
tasteless she: 1
she received: 1
received a: 1
a promotion: 1
promotion at: 1
at work: 1
work heavy: 1
heavy rain: 1
rain caused: 1
caused flooding: 1
flooding downtown: 1
downtown the: 1
the movie: 1
movie was: 1
was surprisingly: 1
surprisingly good: 1
good my: 1
my laptop: 1
laptop battery: 1
battery dies: 1
dies quickly: 1
quickly the: 1
the concert: 1
concert tickets: 1
tickets sold: 1
sold out: 1
out fast: 1
fast he: 1
he adopted: 1
adopted a: 1
a puppy: 1
puppy from: 1
from the: 1
the shelter: 1
shelter the: 1
the exam: 1
exam was: 1
was more: 1
more difficult: 1

```

```

scientists discovered: 1
discovered a: 1
a new: 1
new species: 1
species of: 1
of frog: 1

```

TRI-GRAM COUNTS

```

import collections
combined_text = f"{D1} {D2} {D3} {D4} {D5} {D6} {D7} {D8} {D9} {D10}"
words = combined_text.lower().split()
# Generate bigrams
Trigrams = []
for i in range(len(words) - 2):
    Trigrams.append((words[i], words[i+1], words[i+2]))
# Calculate bigram counts
Trigrams_counts = collections.Counter(Trigrams)
# Print the bigram counts
print("\nTrigrams Counts:")
for Trigrams, count in Trigrams_counts.most_common():
    print(f"{Trigrams[0]} {Trigrams[1]} {Trigrams[2]}: {count}")

```

```

Trigrams Counts:
i enjoy reading: 1
enjoy reading books: 1
reading books on: 1
books on history: 1
on history the: 1
history the pizza: 1
the pizza was: 1
pizza was cold: 1
was cold and: 1
cold and tasteless: 1
and tasteless she: 1
tasteless she received: 1
she received a: 1
received a promotion: 1
a promotion at: 1
promotion at work: 1
at work heavy: 1
work heavy rain: 1
heavy rain caused: 1
rain caused flooding: 1
caused flooding downtown: 1
flooding downtown the: 1
downtown the movie: 1
the movie was: 1
movie was surprisingly: 1
was surprisingly good: 1
surprisingly good my: 1
good my laptop: 1
my laptop battery: 1
laptop battery dies: 1
battery dies quickly: 1
dies quickly the: 1
quickly the concert: 1
the concert tickets: 1
concert tickets sold: 1
tickets sold out: 1
sold out fast: 1
out fast he: 1
fast he adopted: 1
he adopted a: 1
adopted a puppy: 1
a puppy from: 1
puppy from the: 1
from the shelter: 1
the shelter the: 1
shelter the exam: 1
the exam was: 1
exam was more: 1
was more difficult: 1
more difficult than: 1
difficult than expected: 1
than expected scientists: 1
expected scientists discovered: 1
scientists discovered a: 1
discovered a new: 1
a new species: 1

```

NEXT WORD PREDICTION USING BI-GRAM MODEL

```
def predict_next_word_bigram(word_sequence, bigram_counts, unigram_counts):
    # Tokenize the input sequence and get the last word
    words_in_sequence = word_sequence.lower().split()
    if not words_in_sequence:
        return "Please provide a word sequence."
    last_word = words_in_sequence[-1]

    # Find potential next words based on bigrams starting with last_word
    potential_next_words = {}
    for (w1, w2), count in bigram_counts.items():
        if w1 == last_word:
            potential_next_words[w2] = count

    if not potential_next_words:
        return f"No bigram found starting with '{last_word}'."

    # Calculate probabilities for potential next words
    #  $P(w_2 | w_1) = \text{Count}(w_1, w_2) / \text{Count}(w_1)$ 
    last_word_unigram_count = unigram_counts.get(last_word, 0)
    if last_word_unigram_count == 0:
        return f"'{last_word}' not found in unigram counts. Cannot predict next word."

    predicted_word = None
    max_probability = -1

    for next_word, bigram_count in potential_next_words.items():
        probability = bigram_count / last_word_unigram_count
        print("probability of " + f'{next_word}' + " is ", probability)
        if probability > max_probability:
            max_probability = probability
            predicted_word = next_word

    return predicted_word

# Example usage:
# Ensure bigram_counts and unigram_counts are defined from previous cells
# (They are present in the kernel state.)

# Try with a word sequence
sequence1 = "a puppy"
next_word1 = predict_next_word_bigram(sequence1, bigram_counts, unigram_counts)
print(f"Given sequence: '{sequence1}', predicted next word: '{next_word1}'")

sequence2 = "my laptop"
next_word2 = predict_next_word_bigram(sequence2, bigram_counts, unigram_counts)
print(f"Given sequence: '{sequence2}', predicted next word: '{next_word2}'")

sequence3 = "I am"
next_word3 = predict_next_word_bigram(sequence3, bigram_counts, unigram_counts)
print(f"Given sequence: '{sequence3}', predicted next word: '{next_word3}'")

sequence4 = "nonexistent word"
next_word4 = predict_next_word_bigram(sequence4, bigram_counts, unigram_counts)
print(f"Given sequence: '{sequence4}', predicted next word: '{next_word4}'")

probability of from is 1.0
Given sequence: 'a puppy', predicted next word: 'from'
probability of battery is 1.0
Given sequence: 'my laptop', predicted next word: 'battery'
Given sequence: 'I am', predicted next word: 'No bigram found starting with 'am'.'
Given sequence: 'nonexistent word', predicted next word: 'No bigram found starting with 'word'.'
```

DEPLOYMENT OF BI-GRAM MODEL

```
ip_text=input("Enter text : ")
next_word1 = predict_next_word_bigram(ip_text, bigram_counts, unigram_counts)
print(f"Given sequence: '{ip_text}', predicted next word: '{next_word1}'")

Enter text : My name
Given sequence: 'My name', predicted next word: 'No bigram found starting with 'name'.'
```

NEXT WORD PREDICTION USING TRI-GRAM MODEL

```

def predict_next_word_trigram(word_sequence, Trigrams_counts, bigram_counts):
    # Tokenize the input sequence
    words_in_sequence = word_sequence.lower().split()
    if not words_in_sequence:
        return "Please provide a word sequence."

    # Ensure at least two words for trigram prediction
    if len(words_in_sequence) < 2:
        return "Sequence must contain at least two words for trigram prediction."

    # Get the last two words as a tuple
    last_two_words_tuple = tuple(words_in_sequence[-2:])

    # Find potential next words based on trigrams starting with the last two words
    potential_next_words = {}
    for (w1, w2, w3), count in Trigrams_counts.items():
        if (w1, w2) == last_two_words_tuple:
            potential_next_words[w3] = count

    if not potential_next_words:
        return f"No trigram found starting with '{' '.join(last_two_words_tuple)}'."

    # Calculate probabilities for potential next words
    #  $P(w_3 | w_1, w_2) = \text{Count}(w_1, w_2, w_3) / \text{Count}(w_1, w_2)$ 
    # The denominator should be the count of the bigram (w1, w2)
    last_two_words_bigram_count = bigram_counts.get(last_two_words_tuple, 0)
    if last_two_words_bigram_count == 0:
        return f"'{' '.join(last_two_words_tuple)}' not found as a bigram. Cannot predict next word."

    predicted_word = None
    max_probability = -1

    for next_word, trigram_count in potential_next_words.items():
        probability = trigram_count / last_two_words_bigram_count
        print("probability of " + f'{next_word}' + " is ", probability)
        if probability > max_probability:
            max_probability = probability
            predicted_word = next_word

    return predicted_word

# Example usage:
# Ensure bigram_counts, unigram_counts, and Trigrams_counts are defined from previous cells
# (They are present in the kernel state.)

# Try with a word sequence
sequence1 = "the movie was"
next_word1 = predict_next_word_trigram(sequence1, Trigrams_counts, bigram_counts)
print(f"Given sequence: '{sequence1}', predicted next word: '{next_word1}'")

sequence2 = "the pizza was"
next_word2 = predict_next_word_trigram(sequence2, Trigrams_counts, bigram_counts)
print(f"Given sequence: '{sequence2}', predicted next word: '{next_word2}'")

probability of was is 1.0
Given sequence: 'the movie', predicted next word: 'was'
probability of was is 1.0
Given sequence: 'the pizza', predicted next word: 'was'

```

DEPLOYMENT OF TRI-GRAM MODEL

```

ip_text=input("Enter text : ")
next_word1 = predict_next_word_trigram(ip_text, Trigrams_counts, bigram_counts)
print(f"Given sequence: '{ip_text}', predicted next word: '{next_word1}'")

Enter text : a puppy from
probability of the is 1.0
Given sequence: 'a puppy from', predicted next word: 'the'

```

NEXT WORD PREDICTION USING BI-GRAM MODEL WITH LAPLACE SMOOTHENING

```

def predict_next_word_bigram_Laplace(word_sequence, bigram_counts, unigram_counts):
    # Tokenize the input sequence and get the last word
    words_in_sequence = word_sequence.lower().split()

```

```

if not words_in_sequence:
    return "Please provide a word sequence."
last_word = words_in_sequence[-1]

# Find potential next words based on bigrams starting with last_word
potential_next_words = {}
for (w1, w2), count in bigram_counts.items():
    if w1 == last_word:
        potential_next_words[w2] = count

if not potential_next_words:
    return f"No bigram found starting with '{last_word}'."

# Calculate probabilities for potential next words
#  $P(w_2 | w_1) = \text{Count}(w_1, w_2) / \text{Count}(w_1)$ 
last_word_unigram_count = unigram_counts.get(last_word, 0)
if last_word_unigram_count == 0:
    return f"'{last_word}' not found in unigram counts. Cannot predict next word."

predicted_word = None
max_probability = -1

for next_word, bigram_count in potential_next_words.items():
    probability = (bigram_count+1) / (last_word_unigram_count+V)
    print("probability of " + next_word + " is ", probability)
    if probability > max_probability:
        max_probability = probability
        predicted_word = next_word

return predicted_word

# Example usage:
# Ensure bigram_counts and unigram_counts are defined from previous cells
# (They are present in the kernel state.)

# Try with a word sequence
sequence1 = "a puppy"
next_word1 = predict_next_word_bigram_Laplace(sequence1, bigram_counts, unigram_counts)
print(f"Given sequence: '{sequence1}', predicted next word: '{next_word1}'")

sequence2 = "the movie"
next_word2 = predict_next_word_bigram_Laplace(sequence2, bigram_counts, unigram_counts)
print(f"Given sequence: '{sequence2}', predicted next word: '{next_word2}'")

sequence3 = "my laptop"
next_word3 = predict_next_word_bigram_Laplace(sequence3, bigram_counts, unigram_counts)
print(f"Given sequence: '{sequence3}', predicted next word: '{next_word3}'")

sequence4 = "how are"
next_word4 = predict_next_word_bigram_Laplace(sequence4, bigram_counts, unigram_counts)
print(f"Given sequence: '{sequence4}', predicted next word: '{next_word4}'")

```

```

probability of from is 0.03773584905660377
Given sequence: 'a puppy', predicted next word: 'from'
probability of was is 0.03773584905660377
Given sequence: 'the movie', predicted next word: 'was'
probability of battery is 0.03773584905660377
Given sequence: 'my laptop', predicted next word: 'battery'
Given sequence: 'how are', predicted next word: 'No bigram found starting with 'are'.'

```

DEPLOYMENT OF LAPLACE SMOOTHENING BASED ON BI-GRAM MODEL

```

ip_text=input("Enter text : ")
next_word1 = predict_next_word_bigram_Laplace(ip_text, bigram_counts, unigram_counts)
print(f"Given sequence: '{ip_text}', predicted next word: '{next_word1}'")

```

```

Enter text : the rain
probability of caused is 0.03773584905660377
Given sequence: 'the rain', predicted next word: 'caused'

```

NEXT WORD PREDICTION USING TRI-GRAM MODEL BASED ON LAPLACE SMOOTHENING

```

def predict_next_word_trigram_Laplace(word_sequence, Trigrams_counts, bigram_counts):
    # Tokenize the input sequence
    words_in_sequence = word_sequence.lower().split()
    if not words_in_sequence:

```

```

    return "Please provide a word sequence."

# Ensure at least two words for trigram prediction
if len(words_in_sequence) < 2:
    return "Sequence must contain at least two words for trigram prediction."

# Get the last two words as a tuple
last_two_words_tuple = tuple(words_in_sequence[-2:])

# Find potential next words based on trigrams starting with the last two words
potential_next_words = {}
for (w1, w2, w3), count in Trigrams_counts.items():
    if (w1, w2) == last_two_words_tuple:
        potential_next_words[w3] = count

if not potential_next_words:
    return f"No trigram found starting with '{' '.join(last_two_words_tuple)}'."

# Calculate probabilities for potential next words
#  $P(w_3 | w_1, w_2) = \text{Count}(w_1, w_2, w_3) / \text{Count}(w_1, w_2)$ 
# The denominator should be the count of the bigram (w1, w2)
last_two_words_bigram_count = bigram_counts.get(last_two_words_tuple, 0)
if last_two_words_bigram_count == 0:
    return f"'{' '.join(last_two_words_tuple)}' not found as a bigram. Cannot predict next word."

predicted_word = None
max_probability = -1

for next_word, trigram_count in potential_next_words.items():
    probability = (trigram_count+1) / (last_two_words_bigram_count+V)
    print("probability of " + f'{next_word}' + " is ", probability)
    if probability > max_probability:
        max_probability = probability
        predicted_word = next_word

return predicted_word

# Example usage:
# Ensure bigram_counts, unigram_counts, and Trigrams_counts are defined from previous cells
# (They are present in the kernel state.)

# Try with a word sequence
sequence1 = "the pizza was"
next_word1 = predict_next_word_trigram_Laplace(sequence1, Trigrams_counts, bigram_counts)
print(f"Given sequence: '{sequence1}', predicted next word: '{next_word1}'")

sequence2 = "heavy rain is"
next_word2 = predict_next_word_trigram_Laplace(sequence2, Trigrams_counts, bigram_counts)
print(f"Given sequence: '{sequence2}', predicted next word: '{next_word2}'")

probability of cold is 0.03773584905660377
Given sequence: 'the pizza was', predicted next word: 'cold'
Given sequence: 'heavy rain is', predicted next word: 'No trigram found starting with 'rain is'.'

```

DEPLOYMENT OF LAPLACE SMOOTHENING BASED ON TRI-GRAM MODEL

```

ip_text=input("Enter text : ")
next_word1 = predict_next_word_trigram_Laplace(ip_text, Trigrams_counts, bigram_counts)
print(f"Given sequence: '{ip_text}', predicted next word: '{next_word1}'")

```

```

Enter text : the exam was
probability of more is 0.03773584905660377
Given sequence: 'the exam was', predicted next word: 'more'

```

NEXT WORD PREDICTION USING BI-GRAM MODEL ADD - K SMOOTHENING

```

def predict_next_word_bigram_K(word_sequence, bigram_counts, unigram_counts, K): #K=0.5-0.01
    # Tokenize the input sequence and get the last word
    words_in_sequence = word_sequence.lower().split()
    if not words_in_sequence:
        return "Please provide a word sequence."
    last_word = words_in_sequence[-1]

    # Find potential next words based on bigrams starting with last_word
    potential_next_words = {}

```

```

for (w1, w2), count in bigram_counts.items():
    if w1 == last_word:
        potential_next_words[w2] = count

if not potential_next_words:
    return f"No bigram found starting with '{last_word}'."

# Calculate probabilities for potential next words
#  $P(w_2 | w_1) = \text{Count}(w_1, w_2) / \text{Count}(w_1)$ 
last_word_unigram_count = unigram_counts.get(last_word, 0)
if last_word_unigram_count == 0:
    return f"'{last_word}' not found in unigram counts. Cannot predict next word."

predicted_word = None
max_probability = -1

for next_word, bigram_count in potential_next_words.items():
    probability = (bigram_count+K) / (last_word_unigram_count+K*V)
    print("probability of " + f'{next_word}' + " is ", probability)
    if probability > max_probability:
        max_probability = probability
        predicted_word = next_word

return predicted_word

# Example usage:
# Ensure bigram_counts and unigram_counts are defined from previous cells
# (They are present in the kernel state.)

# Try with a word sequence
sequence1 = "heavy rain"
next_word1 = predict_next_word_bigram_K(sequence1, bigram_counts, unigram_counts, 0.5)
print(f"Given sequence: '{sequence1}', predicted next word: '{next_word1}'")

sequence2 = "a puppy"
next_word2 = predict_next_word_bigram_K(sequence2, bigram_counts, unigram_counts, 0.5)
print(f"Given sequence: '{sequence2}', predicted next word: '{next_word2}'")

sequence3 = "the pizza"
next_word3 = predict_next_word_bigram_K(sequence3, bigram_counts, unigram_counts, 0.5)
print(f"Given sequence: '{sequence3}', predicted next word: '{next_word3}'")

sequence4 = "nlp is"
next_word4 = predict_next_word_bigram_K(sequence4, bigram_counts, unigram_counts, 0.5)
print(f"Given sequence: '{sequence4}', predicted next word: '{next_word4}'")

```

```

probability of caused is 0.05555555555555555
Given sequence: 'heavy rain', predicted next word: 'caused'
probability of from is 0.05555555555555555
Given sequence: 'a puppy', predicted next word: 'from'
probability of was is 0.05555555555555555
Given sequence: 'the pizza', predicted next word: 'was'
Given sequence: 'nlp is', predicted next word: 'No bigram found starting with 'is'.'

```

DEPLOYMENT OF ADD-K SMOOTHENING BASED ON BI-GRAM MODEL

```

ip_text=input("Enter text : ")
next_word1 = predict_next_word_bigram_K(ip_text, bigram_counts, unigram_counts, 0.5)
print(f"Given sequence: '{ip_text}', predicted next word: '{next_word1}'")

```

```

Enter text : heavy rain
probability of caused is 0.05555555555555555
Given sequence: 'heavy rain', predicted next word: 'caused'

```

NEXT WORD PREDICTION USING TRI-GRAM MODEL BASED ON ADD-K SMOOTHENING

```

def predict_next_word_trigram_K(word_sequence, Trigrams_counts, bigram_counts, K): #K=0.5-0.01
    # Tokenize the input sequence
    words_in_sequence = word_sequence.lower().split()
    if not words_in_sequence:
        return "Please provide a word sequence."

    # Ensure at least two words for trigram prediction
    if len(words_in_sequence) < 2:
        return "Sequence must contain at least two words for trigram prediction."

```



```

# Get the last two words as a tuple
last_two_words_tuple = tuple(words_in_sequence[-2:])

# Find potential next words based on trigrams starting with the last two words
potential_next_words = {}
for (w1, w2, w3), count in Trigrams_counts.items():
    if (w1, w2) == last_two_words_tuple:
        potential_next_words[w3] = count

if not potential_next_words:
    return f"No trigram found starting with '{' '.join(last_two_words_tuple)}'."

# Calculate probabilities for potential next words
#  $P(w_3 | w_1, w_2) = \text{Count}(w_1, w_2, w_3) / \text{Count}(w_1, w_2)$ 
# The denominator should be the count of the bigram (w1, w2)
last_two_words_bigram_count = bigram_counts.get(last_two_words_tuple, 0)
if last_two_words_bigram_count == 0:
    return f"'{' '.join(last_two_words_tuple)}' not found as a bigram. Cannot predict next word."

predicted_word = None
max_probability = -1

for next_word, trigram_count in potential_next_words.items():
    probability = (trigram_count+K) / (last_two_words_bigram_count+K*V)
    print("probability of " + f'{next_word}' + " is ", probability)
    if probability > max_probability:
        max_probability = probability
        predicted_word = next_word

return predicted_word

# Example usage:
# Ensure bigram_counts, unigram_counts, and Trigrams_counts are defined from previous cells
# (They are present in the kernel state.)

# Try with a word sequence
sequence1 = "the movie was"
next_word1 = predict_next_word_trigram_K(sequence1, Trigrams_counts, bigram_counts, 0.5)
print(f"Given sequence: '{sequence1}', predicted next word: '{next_word1}'")

sequence2 = "I have done"
next_word2 = predict_next_word_trigram_K(sequence2, Trigrams_counts, bigram_counts, 0.5)
print(f"Given sequence: '{sequence2}', predicted next word: '{next_word2}'")

probability of surprisingly is 0.05555555555555555
Given sequence: 'the movie was', predicted next word: 'surprisingly'
Given sequence: 'I have done', predicted next word: 'No trigram found starting with 'have done'.'

```

DEPLOYMENT OF ADD-K SMOOTHENING BASED ON TRI-GRAM MODEL

```

ip_text=input("Enter text : ")
next_word1 = predict_next_word_trigram_K(ip_text, Trigrams_counts, bigram_counts, 0.5)
print(f"Given sequence: '{ip_text}', predicted next word: '{next_word1}'")

Enter text : the movie was
probability of surprisingly is 0.05555555555555555
Given sequence: 'the movie was', predicted next word: 'surprisingly'

```

