```
from collections import deque
def bfs(graph, start_node):
   visited = set()
    queue = deque([start_node])
    print("BFS Traversal Order:")
    while queue:
       node = queue.popleft()
        if node not in visited:
           print(node, end=' ')
           visited.add(node)
           for neighbor in graph[node]:
                if neighbor not in visited:
                    queue.append(neighbor)
graph = {
   'A': ['B', 'C'],
   'B': ['D', 'E'],
   'C': ['F'],
'D': [],
    'E': ['F'],
    'F': []
bfs(graph, 'A')
⇒ BFS Traversal Order:
     ABCDEF
```

```
def dfs(graph, start_node, visited=None):
    if visited is None:
        visited = set()
    visited.add(start_node)
    print(start_node, end=' ')
    for neighbor in graph[start_node]:
        if neighbor not in visited:
            dfs(graph, neighbor, visited)
graph = {
    'A': ['B', 'C'],
'B': ['D', 'E'],
    'C': ['F'],
    'D': [],
    'E': ['F'],
    'F': []
}
print("DFS Traversal Order:")
dfs(graph, 'A')
 → DFS Traversal Order:
     ABDEFC
```

```
def print_board(board):
   for row in board:
       print(" ".join("Q" if col else "." for col in row))
   print()
def is_safe(board, row, col, n):
   for i in range(row):
       if board[i][col]:
           return False
   for i, j in zip(range(row-1, -1, -1), range(col-1, -1, -1)):
       if board[i][j]:
          return False
   for i, j in zip(range(row-1, -1, -1), range(col+1, n)):
       if board[i][j]:
           return False
   return True
def solve_n_queens(board, row, n):
   if row == n:
       print("One valid solution:")
       print_board(board)
       return True
   for col in range(n):
       if is_safe(board, row, col, n):
           board[row][col] = 1
           if solve_n_queens(board, row + 1, n):
               return True
           board[row][col] = 0
   return False
def n_queens(n):
   board = [[0] * n for _ in range(n)]
   if not solve_n_queens(board, 0, n):
       print("No solution exists.")
n = 8
n_queens(n)
→ One valid solution:
    . . . . . . . . Q
```

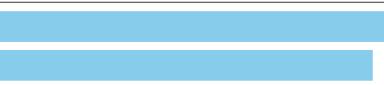
sample_text = "Natural Language Processing (NLP) is a field of artificial intelligence that focuses on the interaction between computers and

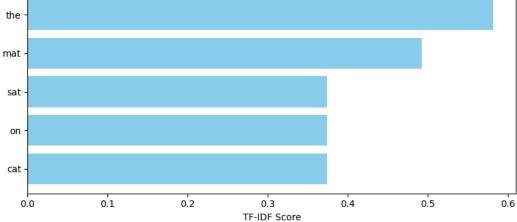
import nltk

```
from nltk.tokenize import word_tokenize, sent_tokenize
from nltk.corpus import stopwords
from nltk.stem import PorterStemmer, WordNetLemmatizer
nltk.download('punkt')
nltk.download('stopwords')
nltk.download('wordnet')
nltk.download('punkt_tab')
words = word_tokenize(sample_text)
sentences = sent_tokenize(sample_text)
stop_words = set(stopwords.words('english'))
filtered_words = [word for word in words if word.lower() not in stop_words]
stemmer = PorterStemmer()
stemmed_words = [stemmer.stem(word) for word in filtered_words]
lemmatizer = WordNetLemmatizer()
lemmatized words = [lemmatizer.lemmatize(word) for word in filtered words]
print("Original Text:", sample_text)
print("\nTokenized Words:", words)
print("\nTokenized Sentences:", sentences)
print("\nStopword Removal:", filtered_words)
print("\nStemmed Words:", stemmed_words)
print("\nLemmatized Words:", lemmatized_words)
print("\n")
🚌 Original Text: Natural Language Processing (NLP) is a field of artificial intelligence that focuses on the interaction between computers
     Tokenized Words: ['Natural', 'Language', 'Processing', '(', 'NLP', ')', 'is', 'a', 'field', 'of', 'artificial', 'intelligence', 'that',
     Tokenized Sentences: ['Natural Language Processing (NLP) is a field of artificial intelligence that focuses on the interaction between c
     Stopword Removal: ['Natural', 'Language', 'Processing', '(', 'NLP', ')', 'field', 'artificial', 'intelligence', 'focuses', 'interaction'
     Stemmed Words: ['natur', 'languag', 'process', '(', 'nlp', ')', 'field', 'artifici', 'intellig', 'focus', 'interact', 'comput', 'human',
     Lemmatized Words: ['Natural', 'Language', 'Processing', '(', 'NLP', ')', 'field', 'artificial', 'intelligence', 'focus', 'interaction',
     [nltk_data] Downloading package punkt to /root/nltk_data...
     [nltk_data]
                   Package punkt is already up-to-date!
     [nltk_data] Downloading package stopwords to /root/nltk_data...
     [nltk_data]
                   Package stopwords is already up-to-date!
     [nltk_data] Downloading package wordnet to /root/nltk_data...
     [nltk_data] Package wordnet is already up-to-date!
     [nltk_data] Downloading package punkt_tab to /root/nltk_data...
     [nltk_data]
                   Package punkt_tab is already up-to-date!
```

```
from sklearn.feature_extraction.text import TfidfVectorizer
import matplotlib.pyplot as plt
import numpy as np
documents = [
    "the cat sat on the mat", \ 
    "the dog sat on the log",
    "the cat chased the dog"
]
vectorizer = TfidfVectorizer()
X = vectorizer.fit_transform(documents)
feature_names = vectorizer.get_feature_names_out()
print("Words:", feature_names)
print("\nTF-IDF Matrix:")
print(X.toarray())
# Function to visualize top TF-IDF words in each document
def visualize_top_words(tfidf_matrix, feature_names, doc_index, top_n=5):
    vector = tfidf_matrix[doc_index].toarray().flatten()
    top_indices = vector.argsort()[-top_n:][::-1]
    top_words = [feature_names[i] for i in top_indices]
    top_scores = vector[top_indices]
    plt.figure(figsize=(8, 4))
    plt.barh(top_words[::-1], top_scores[::-1], color='skyblue')
    plt.xlabel("TF-IDF Score")
    plt.title(f"Top {top_n} words in Document {doc_index + 1}")
    plt.tight_layout()
    plt.show()
for i in range(len(documents)):
    visualize_top_words(X, feature_names, i)
```

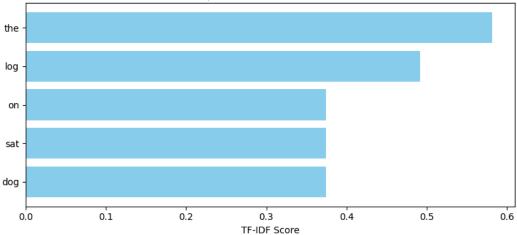
```
Two Words: ['cat' 'chased' 'dog' 'log' 'mat' 'on' 'sat' 'the']
    TF-IDF Matrix:
    [[0.37420726 0.
                                                      0.49203758 0.37420726
                                          0.
      0.37420726 0.58121064]
                 0.
                              0.37420726 0.49203758 0.
                                                                  0.37420726
     0.37420726 0.58121064]
[0.40352536 0.53058735 0.40352536 0.
                  0.62674687]]
```





Top 5 words in Document 1





Top 5 words in Document 3

```
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.feature_extraction.text import TfidfVectorizer
documents = [
    "machine learning models are useful",
    "deep learning is a subset of machine learning",
    "natural language processing uses machine learning techniques"
]
vectorizer = TfidfVectorizer()
X = vectorizer.fit_transform(documents)
feature_names = vectorizer.get_feature_names_out()
df = pd.DataFrame(X.toarray(), columns=feature_names)
print(df)
for i, row in df.iterrows():
    print(f"\nTop words in document {i + 1}:")
    print(row.sort_values(ascending=False).head(3))
print("\nTop terms across all documents:")
print(df.sum().sort_values(ascending=False).head(5))
df.sum().sort_values(ascending=False).head(10).plot(kind='barh')
plt.title("Top TF-IDF Terms")
plt.xlabel("TF-IDF Score")
plt.show()
```

```
are
               deep
                          is language learning
                                                 machine
                                                           models \
0 0.52004 0.000000 0.000000
                               0.00000 0.307144 0.307144 0.52004
1 0.00000
          0.417242 0.417242
                               0.00000 0.492859 0.246430 0.00000
2 0.00000 0.000000 0.000000
                               0.41894 0.247433 0.247433 0.00000
                 of
                    processing
                                  subset techniques
                                                      useful
  natural
                                                                uses
0 0.00000 0.000000
                       0.00000 0.000000
                                            0.00000 0.52004
                                                             0.00000
1 0.00000 0.417242
                       0.00000 0.417242
                                            0.00000 0.00000 0.00000
2 0.41894 0.000000
                       0.41894 0.000000
                                            0.41894 0.00000
                                                             0.41894
Top words in document 1:
are
         0.52004
models
         0.52004
        0.52004
useful
Name: 0, dtype: float64
Top words in document 2:
           0.492859
learning
deep
           0.417242
           0.417242
is
Name: 1, dtype: float64
Top words in document 3:
language
             0.41894
             0.41894
natural
techniques
             0.41894
Name: 2, dtype: float64
Top terms across all documents:
learning
          1.047436
machine
           0.801006
           0.520040
are
models
           0.520040
```

0.520040

useful dtype: float64

```
!pip install scikit-learn
!pip install nltk
    Show hidden output
from sklearn.model_selection import train_test_split
from sklearn.feature_extraction.text import CountVectorizer
from sklearn.naive bayes import MultinomialNB
from sklearn import metrics
# Sample Texts (Spam vs Non-Spam)
documents = [
    "Free money now!",
    "Buy cheap watches",
    "How to make money online",
    "Call me when you're free",
    "Hey, what's up?",
    "Looking forward to our meeting",
    "Earn money fast",
    "Huge discount on watches",
    "Free online course",
    "Let's meet for coffee",
    "I will call you later"
labels = [1, 1, 1, 0, 0, 0, 1, 1, 1, 0, 0] # 1 = Spam, 0 = Non-Spam
X_train, X_test, y_train, y_test = train_test_split(
    documents, labels, test_size=0.3, random_state=42, stratify=labels
)
vectorizer = CountVectorizer()
X_train = vectorizer.fit_transform(X_train)
X_test = vectorizer.transform(X_test)
nb = MultinomialNB()
nb.fit(X_train, y_train)
y pred = nb.predict(X test)
print(f"Accuracy: {metrics.accuracy_score(y_test, y_pred)}")
print(f"Classification Report:\n{metrics.classification_report(y_test, y_pred)}")
Accuracy: 0.75
     Classification Report:
                   precision
                                recall f1-score
                                                   support
                0
                        1.00
                                  0.50
                                            0.67
                1
                        0.67
                                  1.00
                                            0.80
                                                         2
                                            0.75
                                                         4
         accuracy
                                  0.75
                        0.83
        macro avg
                                            0.73
                                                         4
     weighted avg
                        0.83
                                  0.75
                                            0.73
                                                         4
```

```
import random
def print_board(board):
    for row in board:
       print("|".join(row))
print("-" * 5)
def check_winner(board, player):
    # Check rows, columns, and diagonals
    for i in range(3):
        if all([cell == player for cell in board[i]]) or all([board[j][i] == player for j in range(3)]):
    return board[0][0] == board[1][1] == board[2][2] == player or \
           board[0][2] == board[1][1] == board[2][0] == player
def get_empty_positions(board):
    return [(i, j) for i in range(3) for j in range(3) if board[i][j] == ' ']
def ai_move(board):
    # Try to win
    for (i, j) in get_empty_positions(board):
        board[i][j] = '0'
        if check_winner(board, '0'):
            return
        board[i][j] = ' '
    # Try to block X
    for (i, j) in get_empty_positions(board):
        board[i][j] = 'X'
        if check_winner(board, 'X'):
            board[i][j] = '0'
            return
        board[i][j] = ' '
    # Take center if available
    if board[1][1] == ' ':
        board[1][1] = '0'
        return
    # Take any corner or side
    move = random.choice(get_empty_positions(board))
    board[move[0]][move[1]] = '0'
def play_game():
    board = [[' ']*3 for _ in range(3)]
print("Tic-Tac-Toe: You are X, AI is 0")
    print_board(board)
    for _ in range(9):
        # Player move
        while True:
            try:
                row = int(input("Enter row (0-2): "))
                col = int(input("Enter col (0-2): "))
                if board[row][col] == ' ':
                    board[row][col] = 'X'
                    break
                else:
                    print("Cell already taken!")
                print("Invalid input. Try again.")
        print_board(board)
        if check_winner(board, 'X'):
            print("You win!")
            return
        if not get_empty_positions(board):
        print("AI is making a move...")
        ai_move(board)
        print_board(board)
        if check_winner(board, '0'):
            print("AI wins!")
```

```
return
        if not get_empty_positions(board):
            break
    print("It's a draw!")
# Run the game
play_game()
 → Tic-Tac-Toe: You are X, AI is O
      Enter row (0-2): 0
     Enter col (0-2): 0
     X| |
     \perp
     AI is making a move...
     X| |
     0
      Enter row (0-2): 0
Enter col (0-2): 1
     X|X|
      |0|
     AI is making a move...
     x|x|o
     0
     \perp
     Enter row (0-2): 2
     Enter col (0-2): 0
     x|x|o
     0
     x| |
     AI is making a move...
     x|x|o
     0|0|
     X| |
     Enter row (0-2): 1
     Enter col (0-2): 3
     Invalid input. Try again.
     Enter row (0-2): 2
Enter col (0-2): 3
Invalid input. Try again.
```

```
import numpy as np
from sklearn.linear model import Perceptron
from sklearn.linear_model import LogisticRegression
# Sample dataset
X = np.array([
    [0, 0, 1, 0], # free down
    [1, 0, 1, 0], # free down
    [0, 1, 1, 0], # free up
    [1, 1, 1, 1], # blocked all sides
    [0, 0, 0, 1], # free up and down
    [0, 1, 0, 0], # free up, right
y_binary = np.array([1, 1, 1, 0, 1, 1]) # Move or not
y_{multi} = np.array([1, 1, 0, -1, 0, 3]) # Direction: Up = 0, Down = 1, Left = 2, Right = 3
# Train Perceptron (binary classification)
perceptron = Perceptron()
perceptron.fit(X, y_binary)
# Train multi-category model (direction prediction)
X_{multi} = X[y_{multi} != -1]
                                 # Filter out rows with no direction
y_multi_filtered = y_multi[y_multi != -1]
multi_model = LogisticRegression(multi_class='multinomial', solver='lbfgs')
multi_model.fit(X_multi, y_multi_filtered)
# Test input (robot senses obstacles)
test\_input = np.array([[0, \ 0, \ 1, \ 0]]) \ \ \# \ Obstacle \ Left
# Predict movement (1 = move, 0 = stop)
move_decision = perceptron.predict(test_input)[0]
if move_decision == 1:
    direction = multi_model.predict(test_input)[0]
    direction_map = {0: "Up", 1: "Down", 2: "Left", 3: "Right"}
    print("Decision: MOVE")
    print("Direction:", direction_map[direction])
else:
    print("Decision: STOP")
→ Decision: MOVE
     Direction: Down
     /usr/local/lib/python3.11/dist-packages/sklearn/linear_model/_logistic.py:1247: FutureWarning: 'multi_class' was deprecated in version 1
       warnings.warn(
```