Python and Machine Learning Assignment By Kashinath J

1. Problem Statement: Given a 0-indexed integer array `nums`, return `true` if it can be made strictly increasing after removing exactly one element, or `false` otherwise. If the array is already strictly increasing, return `true`.

Solution:

```
def canBeIncreasing(nums):
  Checks if the given array can be made strictly increasing after removing exactly one element.
  Parameters:
  - nums (List[int]): A 0-indexed integer array.
  Returns:
  - bool: True if the array can be made strictly increasing, False otherwise.
  # Variable to count non-strictly increasing pairs
  count = 0
  # Iterate through the array
  for i in range(len(nums) - 1):
     # Check if the current element is greater than or equal to the next element
     if nums[i] >= nums[i + 1]:
       # Increment the count of non-strictly increasing pairs
       count += 1
       # If there are more than one non-strictly increasing pairs, return False
       if count > 1:
          return False
       # Check conditions for continuing or returning False
       if i == 0 or i == len(nums) - 2 or nums[i - 1] < nums[i + 1] or nums[i] < nums[i + 2]:
          continue
       else:
          return False
  # If no more than one removal makes the array strictly increasing, return True
  return True
```

```
# Example 1
nums1 = [1, 2, 10, 5, 7]
print(canBeIncreasing(nums1)) # Output: True

# Example 2
nums2 = [2, 3, 1, 2]
print(canBeIncreasing(nums2)) # Output: False

# Example 3
nums3 = [1, 1, 1]
print(canBeIncreasing(nums3)) # Output: False

True
False
False
False
```

Explanation:

- The function can Belncreasing takes a list of integers nums as input and returns True if the array can be made strictly increasing, False otherwise.
- The variable count is used to keep track of non-strictly increasing pairs.
- The function iterates through the array, checking each element against the next.
- If a non-strictly increasing pair is found, the count is incremented. If there are more than one such pair, the function returns False.

- Conditions are checked for continuing the loop or returning False based on the index and adjacent elements.
- If the loop completes without more than one non-strictly increasing pair, the function returns True.
- 2. Problem Statement: Given a string `s`, find the length of the longest substring without repeating characters.

return max_len

```
Solution:
def length of longest substring(s):
  Finds the length of the longest substring without repeating characters.
  Parameters:
  - s (str): Input string.
  Returns:
  - int: Length of the longest substring without repeating characters.
  # Initialize variables
         # Start of the current substring
  k = 0
  max_len = 0 # Maximum length found
  count = 0 # Current substring length
  # Iterate through the string
  for i in range(1, len(s)):
     # Check for repeating characters in the current substring
     for j in range(k, i):
       if s[i] == s[j]:
          k = j + 1
     # Update the current substring length
     count = i - k + 1
     # Update the maximum length
     if count > max len:
       max_len = count
  # Return the result
```

```
def length_of_longest_substring(s):
        # Initialize variables
        max_len = 0  # Maximum length found
count = 0  # Current substring length
        # Iterate through the string
        for i in range(1, len(s)):
            # Check for repeating characters in the current substring
            for j in range(k, i):
               if s[i] == s[j]:
                   k = j + 1
            # Update the current substring length
            count = i - k + 1
            # Update the maximum length
            if count > max len:
               max len = count
        # Return the result
        return max len
```

```
[6] # Example1
    s_example = "abcabcbb"
    print(length_of_longest_substring(s_example))
    # Example2
    s_example = "abcabcdebb"
    print(length_of_longest_substring(s_example))
3
5
```

Explanation:

• Initialization:

k: Represents the start of the current substring.

max_len: Represents the maximum length found. count: Represents the current substring length.

• Iteration through the string:

for i in range(1, len(s)): iterates through the string starting from index 1.

The outer loop represents the end of the current substring.

• Checking for repeating characters:

for j in range(k, i): iterates through the current substring to check for repeating characters. If a repeating character is found at index j, update the start index k to j + 1. Updating current substring length:

count = i - k + 1 calculates the length of the current substring without repeating characters.

Updating the maximum length:

return False

If count is greater than the current max_len, update max_len with the new value. Returning the result:

After iterating through the entire string, return the final max_len.

3. Problem Statement: Given an array `arr` of integers, check if there exist two indices `i` and `j` such that:
- i != j
- 0 <= i, j < arr.length
- arr[i] == 2 * arr[j]
Solution:
def check_double(arr):
Iterate through each pair of indices i and j
for i in range(len(arr)):
for j in range(len(arr)):
Check if the conditions arr[i] == 2 * arr[j] and i != j are satisfied
if i != j and arr[i] == 2 * arr[j]:
return True</p>
If no such pair is found, return False

```
def check_double(arr):
    # Iterate through each pair of indices i and j
    for i in range(len(arr)):
        for j in range(len(arr)):
            # Check if the conditions arr[i] == 2 * arr[j] and i != j are satisfied
        if i != j and arr[i] == 2 * arr[j]:
            return True

# If no such pair is found, return False
    return False
```

```
[8] # Example 1
arr1 = [10, 2, 5, 3]
print(check_double(arr1)) # Output: True

# Example 2
arr2 = [3, 1, 7, 11]
print(check_double(arr2)) # Output: False

True
False
```

Explanation:

- The function check_double takes an array of integers arr as input and returns True if there exist two indices i and j such that arr[i] is equal to 2 * arr[j], otherwise, it returns False.
- The function uses nested loops to iterate through each pair of indices i and j in the array.
- Inside the loop, it checks if the conditions arr[i] == 2 * arr[j] and i != j are satisfied.
- If a pair of indices satisfying the conditions is found, the function returns True.
- If no such pair is found after iterating through the entire array, the function returns False.

NLP Task:

Summary of Implementation:

1. Libraries Installation:

• Upgraded pip and installed necessary libraries, including torch, torchdata, transformers, and datasets.

2. Loading Dataset:

• Loaded the dataset named "knkarthick/dialogsum" using the load_dataset function from the datasets library.

3. Example Indices:

• Selected two example indices from the test set (example_indices = [140, 20]).

4. Baseline Human Summary and Dialogue Display:

• Displayed the baseline human summary and input dialogue for the selected examples.

5. Model Configuration:

- Used the FLAN-T5 base model for sequence-to-sequence tasks (model name='google/flan-t5-base').
- Loaded the model and tokenizer from the Hugging Face model hub.

6. Model Generation - Without Prompt Engineering:

- Encoded the input sentence and decoded it to check the functionality of the tokenizer.
- Generated summaries using the model without prompt engineering for selected examples.
- Displayed the input prompt, baseline human summary, and model-generated summary.

7. Model Generation - Zero Shot:

- Created a prompt asking the model to summarize the conversation without prompt engineering.
- Generated summaries using the model for selected examples.
- Displayed the input prompt, baseline human summary, and model-generated summary.

8. Model Generation - Zero Shot with Additional Context:

- Modified the prompt by including additional context in the form of a question.
- Generated summaries using the model for selected examples.
- Displayed the input prompt, baseline human summary, and model-generated summary.

9. Model Generation - One Shot Learning:

- Created a prompt with one-shot learning context for a specific example.
- Generated a summary using the model.
- Displayed the input prompt, baseline human summary, and model-generated summary.

10. Model Generation - Few Shot Learning:

- Created a prompt with few-shot learning context for a specific example.
- Generated a summary using the model.
- Displayed the input prompt, baseline human summary, and model-generated summary.

11. Model Generation - Few Shot Learning with Configured Tokens:

- Configured the generation to allow a maximum of 150 new tokens.
- Generated a summary using the model.
- Displayed the model-generated summary.

12. Summary Display:

• Displayed a dash line, baseline human summary, and model-generated summary for each scenario.

Key Observations:

- The code demonstrates various ways to prompt the model for dialogue summarization, including zero-shot, one-shot, and few-shot learning approaches.
- Additional context in the form of questions or statements influences the generated summaries.
- Configuring the maximum number of new tokens can affect the length and content of the generated summaries.

Challenges Faced

Complexity of Dialogue Summarization:

- Dialogue summarization is a challenging task due to the dynamic and context-dependent nature of conversations.
- Developing an effective chatbot for dialogue summarization requires handling diverse input structures and generating coherent and concise summaries.

Optimizing Prompt Engineering:

- Identifying the most effective prompts for generating high-quality summaries is a non-trivial task.
- Experimentation with various prompt structures and techniques is necessary to find the best approach for a given model.

Handling Model Limitations:

- Pre-trained models may have limitations in understanding context and generating accurate summaries.
- Dealing with these limitations requires a combination of prompt engineering, model fine-tuning, and exploring different model architectures.

Final Note:

The implementation showcases different prompt engineering strategies for dialogue summarization using the FLAN-T5 model.

It serves as a comprehensive example of using Hugging Face's transformers library for sequence-to-sequence tasks.

```
cample_indices_full = [40, 100, 200]
fow_bots_prompt: make prompt(cample_indices_full, example_index_to_summarize)
print(fow_shot_prompt)
print(fow_sh
```

Colab link:

https://colab.research.google.com/drive/10YGiatB11x u90x8msAQN8RWliviwmM9?usp=sharing