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1 Backtracking - Generate Permutations

```
Time Complexity: O(m!) where m is the length of the array.
Space Complexity: O(m!) where m is the length of the array.
       def permute(nums: list[int]) -> list[list[int]]:
           """Use backtracking to generate all permutations."""
           # Use Python's implicit "booleannes"
           # `[]` => `False` (generally works for empty sequences)
           if not nums:
               return []
           # The array to be filled
           res: list[list[int]] = []
           # Backtrack and fill the results list
           # NOTE: lists in Python work like pass-by-reference in C++
           backtrack(nums, 0, [], res)
           # Return the result
           return res
      def backtrack(
          nums: list[int], start: int, cur: list[int], res: list[list[int]]
       ) -> None:
           """Perform backtracking on the array of numbers."""
           # Base case: we got one full permutation
           if len(cur) == len(nums):
               # Need to copy since
               res.append(cur[:])
               return
           # Recursive case: generate, recurse, and backtrack
           for i in range(start, len(nums)):
               # Append to the current permutation
               cur.append(nums[i])
               # Place the `i`th element at the start position
               # NOTE: Needed to include previous elements (before `start`)
               nums[i], nums[start] = nums[start], nums[i]
               # Recurse
               backtrack(nums, start + 1, cur, res)
               # Backtrack
               nums[start], nums[i] = nums[i], nums[start]
               cur.pop()
```

2 Balanced Parentheses

Optimal Solution:

Time Complexity: O(m) where m is the length of the string. Space Complexity: O(n) where n is the number parentheses symbols.

The parentheses in a string are balanced if and only if the following two conditions are met:

- 1. The number of "(" and ")" are equal
- 2. Scanning through the string from left to right and counting how many "(" and ")" there are so far, there should never be a time where there are more ")" than "(". We denote the balance of the string as balance = count("(") count(")")

If we only allow for "()" pair of parentheses, then the following will check if the parenthesis are balanced:

```
def is_balanced(string: str) -> bool:
    """Check if a given string has balanced parentheses."""

balance: int = 0
for char in string:
    if char == "(":
        balance += 1
    elif char == ")":
        balance -= 1

if balance < 0:
    return False

return balance == 0</pre>
```

If we allow for more than one pair of parentheses, the following stack-based implementation generalizes well:

```
def is_balanced(string: str) -> bool:
    """Check if a given string has balanced parentheses."""

stack: list[str] = []
    closing: dict[str, str] = {")": "(", "}": "{", "]": "["}
    opening: set[str] = set(closing.values())
    for char in string:
        if char in opening:
            stack.append(char)
        elif char in closing:
            if not stack or closing[char] != stack.pop():
                 return False
```

3 Boyer-Moore Majority Vote Algorithm - Majority Element

4 Flood Fill - Number of Islands

```
Time Complexity: O(m \times n) where m is the number of rows and n is the number of columns. Space Complexity: O(m \times n) where m is the number of rows and n is the number of columns.
```

```
def islands(grid: List[List[str]]) -> int:
    """Counts the number of islands of 1s.
    NOTE: Uses the fact that lists are mutable (will not work in purely
          functional languages such as Haskell).
    m: int = len(grid)
   n: int = len(grid[0])
    if m * n == 0:
        return 0
    islands: int = 0
    for i in range(m):
        for j in range(n):
            if grid[i][j] == "1":
                self.fill(grid, i, j)
                islands += 1
    return islands
def flood_fill(grid: list[list[int]], i: int, j: int) -> bool:
    """Implements the Flood Fill algorithm."""
    # Row index out of bounds
    if not (0 <= i <= len(grid) - 1):
        return False
    # Column index out of bounds
    if not (0 \le j \le len(grid[0]) - 1):
        return False
    # If the current symbol is not 1, no need to continue
    if grid[i][j] != "1":
        return False
    # Mark with a sentinel symbol
    grid[i][j] = "$"
    # Recurse in four directions
    flood_fill(grid, i - 1, j)
    flood_fill(grid, i + 1, j)
    flood_fill(grid, i, j - 1)
    flood_fill(grid, i, j + 1)
```

5 Kadane's Algorithm - Maximum Contiguous Subarray

Optimal Solution:

```
Time Complexity: O(m) where m is the length of the array. Space Complexity: O(1)
```

An implemmentation of the Kadane's algorithm.

```
def max_subarray_sum(arr: list[float]) -> float:
    """Return the maximum subarray sum."""

cur: float = 0
    res: float = float("-inf")
    for num in nums:
        cur = max(cur + num, num)
        res = max(res, cur)
    return res
```

This algorithm is a trivial example of dynamic programming.

6 Levenshtein Distance - Edit Distance

Optimal Solution:

Time Complexity: $O(m \times n)$ where m is the number of rows and n is the number of columns. Space Complexity: $O(m \times n)$ where m is the number of rows and n is the number of columns.

Given two strings word1 and word2, we need to find the optimal/minimum number of operations required to convert word1 to word2. We are permitted the following four operations:

- 1. Insert a character
- 2. Delete a character
- 3. Replace a character
- 4. Keep a character

We take the bottom-up dynamic programming approach:

```
def levenshtein_distance(word1: str, word2: str) -> int:
    """Computes the Levenshtein's distance between two words."""
    # Get the lengths
    m: int = len(word1)
    n: int = len(word2)
    # Prepopulate the DP matrix
    dp: list[list[int]] = [
        [0 for \underline{\ } in range(n + 1)] for \underline{\ } in range(m + 1)
    # Initialize the DP matrix with base cases
    for i in range(1, m + 1):
        dp[i][0] += i
    for j in range(1, n + 1):
        dp[0][j] += j
    # Fill the DP Matrix
    for i in range(1, m + 1):
        for j in range(1, n + 1):
            left: int = dp[i][j - 1]
            top: int = dp[i - 1][j]
            top_left: int = dp[i - 1][j - 1]
            # If characters are equal, just copy the diagonal value
            if word1[i - 1] == word2[j - 1]:
                dp[i][j] = top_left
            # Otherwise, take the minimum of three adjacent cells and
            # increment the result by one
                dp[i][j] = min(left, top, top_left) + 1
    # Return the optimal solution
    return dp[m][n]
```

7 Merge Sort

Optimal Solution:

```
Time Complexity: O(m \times log(m)) where m is the length of the array. Space Complexity: O(m) where m is the length of the array.
```

Merge sort is one of the simplest sorting algorithms that scales very well. It relies on two operations - merge and sort.

```
def merge(arr1: list[int], arr2: list[int]) -> list[int]:
    """Merge two sorted lists."""
    res: list[int] = []
    i: int = 0
    j: int = 0
    while i \le len(arr1) - 1 and j \le len(arr2) - 1:
        if arr1[i] <= arr2[j]:</pre>
            res.append(arr1[i])
            i += 1
        else:
            res.append(arr2[j])
            j += 1
    while i \le len(arr1) - 1:
        res.append(arr1[i])
        i += 1
    while j \le len(arr2) - 1:
        res.append(arr2[j])
        j += 1
    return res
```

We now need to implement the sort function.

```
def sort(arr: list[int]) -> list[int]:
    """Performs a merge sort."""

if len(nums) <= 1:
    return nums

mid: int = len(arr) // 2
    left: list[int] = sort(arr[:mid])
    right: list[int] = sort(arr[mid:])

return merge(left, right)</pre>
```

8 Floyd's Cycle Finding Algorithm - Linked List Cycle Detection

```
Time Complexity: O(m) where m is the number of nodes.
Space Complexity: O(m) where m is the number of nodes.
      from typing import Union
      class Node:
          def __init__(self, val: int = 0, nxt: Node = Union[None, Node]):
              self._val = val
              self._next = nxt
      def floyd_cycle(head: Node) -> bool:
           """An implementation of Floyd's Cycle Finding Algorithm."""
           if not head:
              return False
           # Initialize slow and fast pointers
          slow: Node = head
           fast: Node = head.next
           while slow != fast:
              # No need to check if `fast._next._next`
              # We check only `fast._next` so that `fast = fast._next._next`
              # does not result in an error
              if not fast or not fast._next:
                  return False
              slow = slow._next
              fast = fast._next._next
           return True
```

9 Reverse Linked List

```
Time Complexity: O(m) where m is the number of nodes.
Space Complexity: O(m) where m is the number of nodes.
      from typing import Union
      class Node:
          def __init__(self, val: int = 0, next: Node = Union[None, Node]):
              self._val = val
              self._next = next
      def reverse(head: Node) -> bool:
           """Reverse a linked list."""
           cur: Node = head
          pre: Node = Union[None, Node]
           while cur:
               # Variable `tmp` is needed for moving forward in the loop
              tmp: Node = cur._next
               # Rearrange pointers
              cur._next: Union[None, Node] = pre
              pre = cur
               # Move forward
              cur = tmp
          return pre
```

10 Trie - A Prefix Tree

There are many ways to constuct a trie. We will use a hash map based approach. We first create a Node class.

```
class TrieNode:
    def __init__(self) -> None:
        """Initialize the `TrieNode` class."""

    self._children: dict[str, TrieNode] = {}
    self._isendofword: bool = False
```

We can now design the trie data structure. The initializer will only contain a root of the trie.

```
class Trie:
    def __init__(self) -> None:
        """Initialize the `Trie` class."""
    self._root = TrieNode()
```

Now, the first method to implement is the insertion operation. If we cannot insert a word into a trie, we are not going to be able to do much with it. The algorithm is very straightforward. One iterates over all characters in the word to be inserted and checks if the characters is in the children of the current node. If it is, follow the path of that node. If not, create a new node at that position and once again, follow that road.

```
def insert(self, word: str) -> None:
    """Initialize the `Trie` class."""

ptr: Node = self._root
    for char in word:
        if char not in ptr._children:
            ptr._children[char] = TrieNode()
        ptr = ptr._children[char]

ptr._isendofword = True
```

One of the primary features of the trie is to be able to search for a word. The implementation of this algorithm is shown below.

```
def search(self, word: str) -> bool:
    """Check if a trie contains a given word."""

ptr: Node = self._root
    for char in word:
        if char not in ptr._children:
            return False
        ptr = ptr._children[char]

return ptr._isendofword
```

Similarly, searching a prefix is also very important. It looks almost exactly the same as the code for searching a word.

```
def prefix_search(self, word: str) -> bool:
    """Check if a trie contains a given prefix."""

ptr: Node = self._root
    for char in word:
        if char not in ptr._children:
            return False
        ptr = ptr._children[char]

return True
```