# **Automatic Grading**

All questions in this section are automatically graded. The order of questions and the options are randomized. For fill-in-the-blanks, please follow the format as precisely as possible. Once you move beyond this section, you cannot go back.

1. Check all statements that are **True** with respect to the Single-Source Shortest Paths (SSSP) problems discussed in the Lecture 7b+8a.

(5 marks)

We can easily (in not more than 15 lines) implement the O((|V|+|E|) log |V|) Modified Dijkstra's algorithm in Python because we can use Python's standard library: heapq in a Lazy manner.

The original Dijkstra's cannot implement decreaseKey() operation effectively, while the modified Dijkstra's solve this

while the modified Dijkstra's solve this. We can solve the **unweighted** SSSP problem using O(|V|+|E|) BFS algorithm.

The  $O((|V|+|E|) \log |V|)$  Modified Dijkstra's algorithm can be trapped in an infinite loop if we run it on a graph with negative weight cycle reachable from the source **s**.

We can solve the **unweighted** SSSP problem using  $O((|V|+|E|) \log |V|)$  Modified Dijkstra's algorithm.

Unweighted graph can be solved by Modified Dijkstra's, but BFS is more effective

We can solve the **non-negative weighted** SSSP problem using  $O((|V|+|E|) \log |V|)$  Modified Dijkstra's algorithm.

## All options are correct

2. Check all statements that are **True** with respect to the Traveling-Salesman-Problem (TSP) discussed in the last Lecture 8b.

(5 marks)

The algorithm that tries all permutations of  $\mathbf{N}$  vertices runs in  $O(\mathbf{N}!)$  (or  $O((\mathbf{N}-1)!)$ ) if the first vertex is fixed to vertex 0).

We can solve TSP using the **Modified Dijkstra's** algorithm to find the shortest tour that visits each vertex of the graph once.

Modified Dijkstra's cannot solve TSP effectively, which can only find the single-source shortest path, TSP requires a minimum loop include all nodes For small test cases, e.g., **N** <= 10, we can use

Python's permutations function from itertools to help us try which permutation of **N** vertices yield the shortest cycle/tour.

We can solve TSP using the **Breadth-First Search** algorithm to find the shortest tour that visits each vertex of the graph once.

Same problem as Dijkstra's, BFS can find shortest path between two nodes in a unweighted graph, but cannot find a minimum loop to travel all nodes

Only I and III are true

3. Check all statements that are **True** with respect to these Table ADT data structures discussed in Lecture 5a/5b.

(5 marks)

False

False

(Python custom-)implementation of Hash Table with Separate Chaining collision resolution technique is just a Python list of Python lists.

To avoid ambiguity, we are referring to HashTableDemo.py shown in class.

Hash Table is a list, and the element in Hash Table is also a list (often Doubly Linked List)

There are Table ADT a few additional operations that balanced BST implementation can do that a Hash Table implementation cannot do.

Balanced BST can find minimum/maximum element, check predecessor/successor

If we just need to support the 3 default Table ADT operations: Search(v), Insert(v), and Remove(v) efficiently (i.e., faster than O(**N**) per operation), we can *either* use a good Hash Table implementation (e.g., Separate Chaining) or a *balanced* Binary Search Tree implementation (e.g., self-balancing BST like AVL Tree - an extension of the standard BST that we learned in class). Balanced BST take all 3 default operations as O(logN), Hash Table as O(1)

(Python custom-)implementation of <u>Binary Search Tree is already good enough</u> to be immediately used as a data structure to support Table ADT operations.

To avoid ambiguity, we are referring to BSTDemo.py shown in class (which is not yet an AVL Tree).

For normal BST, in some boundary cases it will like Linked List, operations will finished in O(N). But for most cases, BST is good enough.

All correct, option IV may vary depends on cases

## 4. Fill in the blanks

(5 marks)

You are given the following Python code. Notice the position of the comments.

```
N = 5000
                    # line 1
                                      For overall time complexity of # Line 1-9, the
ans = 0
                    # line 2
                                      outer loop costs O(N), in the loop, the while
for i in range(N//2): # line 3
                                      loop costs O(logN) and for loop costs O(N)
   j = 1
                   # line 4
   while j < N:
j *= 3
                   # line 5
                                      In total, the time complexity is:
                    # line 6
                                      O(N^*(logN + N)) = O(NlogN + N^2)
   for k in range(N): # line 7
       ans += 5
                   # line 8
                                  We keep the largest term only, which is O(N^2)
                    # line 9
print(ans)
```

For all the boxes below, choose one of the following standard time complexities: **1**, log **N**, **N**, **N** log **N**, **N**^2, **N**^2 log **N**, or **N**^3. Do not write "O(" and ")" anymore as <u>O(your-actual-answer)</u> anymore that Big-O notation is already written. Remember that the grading by machine is **very strict**, so do not lose marks because of formatting issues.

The time complexity of # line 1 to 2 (just before entering the outer loop) is O( 1 ).

The time complexity of # line 4 to 6 (the first part of the inner loop) is O( logN).

The time complexity of # line 7 to 8 (the second part of the inner loop) is O(N).

The overall time complexity of # line 1 to 9 (i.e., the whole program) is  $O(N^2)$ .

The final value of ans when it is printed at line 9 is 62500000

Enter the correct answer below.

1		Character Limit: 3
2		Character Limit: 5
3		Character Limit: 5
4		Character Limit: 5
5	Please enter a number for this text box.	Character Limit: 9

**5.** Check all statements that are **True** with respect to the Graph Traversal algorithms: Depth-First Search (DFS) and Breadth-First Search (BFS) discussed in Lecture 6b/7a.

(5 marks)

We can modify either DFS or BFS algorithm in order to find **ALL** topological **False** orderings of a Directed Acyclic Graph (DAG) G = (V, E) in O(|V|+|E|) time. DFS/BFS can find ONE topological ordering in O(V+E), but cannot final ALL in this time We run DFS from a source vertex **s** on a same graph G = (V, E) and it does not visit a vertex x. Therefore, if we run BFS from the same source vertex **s** on the same graph G = (V, E), it will also *not* visit vertex **x**. **False** There is at least one graph traversal application that BFS can do but DFS cannot and vice versa. DFS and BFS may suite for different problems, but they can solve the same problem set Both DFS and BFS run in O(|V|+|E|) when started from the same source vertex s on the same graph G = (V, E) which has been stored in an Adjacency List. Graph Data Structure only influence the Space Complexity Both DFS and BFS visit the same set of vertices when started from the same source vertex s on the same graph G = (V, E). **False** Both DFS and BFS visit the same **sequence** of vertices when started from the same source vertex s on the same graph G = (V, E). DFS is depth-first, BFS is breath-first, they will visit vertex in DIFFERENCE sequence

Only II, III, IV are correct

**6.** Check all statements that are **True** with respect to the Graph Data Structures discussed in Lecture 6a.

(5 marks)

The best graph data structure to be used for ALL graph problems is Adjacency
List.

AL suitable to separate graph (not dense), AM for dense graph

Adjacency List data structure uses O(|V|+|E|) space.

Near complete means there still some pair of edges has no connection, the space complexity still is $O(V+E)$ not equal to $O(V^2)$ , only in complete graph it is $O(V^2)$	
If the graph to be stored is a <b>dense</b> (to be precise, <b>near complete</b> ) weighted	E-la-
graph, then using either Adjacency Matrix or Adjacency List will not be too	False
much different, i.e., both will use O(V^2) space.	
	-
Converting a graph that is currently stored in an Adjacency Matrix into an	False
Adjacency List takes O( V + E ) time. Also take O(V^2) time, to check all blocks for a N*N matrix	
Adjacency Matrix data structure uses O( V + E ) space. Use O(V^2) matrix for a	False

# Only II is correct

7. Match the algorithm names on the left with its tightest worst-case time complexity. Assume that all graphs are simple graphs (no self-loop and no multi-edges).

(5 marks)

Drag the options on the left and drop them into the options on the right.

Depth-First Search on a graph G = (V, E)	O(log V)
O(V+E)	O(1)
Modified Dijkstra's on a non-negative weighted graph G = (V, E)	O((V+E) log V)
O((V+E)logV)	
, , , ,	O(V+E)
Breadth-First Search on a Tree T with	
V vertices	O(E log V)
O(V)	
	O(V)
Sort E weighted edges of an Edge List	
based on non-decreasing weights using Merge Sort	O(V log V)
O(ElogE)	
	O(V^2)
Converting an Adjacency Matrix of a	
graph G = (V, E) into an Edge List	O(V^3)
O(V^2)	3(. 3)
	Reset

# 8. Fill in the blanks

(5 marks)

Steven knows that most IT5003 students (final paper on 4 Dec 2021) will have attempted (parts of) CS2040C final paper on 27 Nov 2021, thus probably have seen the questions about median of 3 sorted arrays (called lists in Python), each array has the same length  $\bf N$ .

You are given the following Python code that is supposed to solve that question, but is it the 'fastest way'? Let's find out:

```
>>> from statistics import median
>>> A, B, C = [1, 2, 3], [1, 5, 9], [2, 7, 8] # just an example
>>> median(A) # should be 2
2
>>> median(B) # should be 5
5
>>> median(C) # should be 7
7
>>> A+B+C # should be unsorted version of the 3 combined lists # question A
[1, 2, 3, 1, 5, 9, 2, 7, 8]
>>> median(A+B+C) # question B
3
>>> sorted(A+B+C) # to convince that 3 is the median of the 3 combined sorted lists
[1, 1, 2, 2, 3, 5, 7, 8, 9]
```

Now, the time complexity of the line A+B+C in terms of N as highlighted in # question A is O( N ). Do not write O() anymore. Remember that len(A) = len(B) = len(C) = N for this problem.

Next, the time complexity of the line median(A+B+C) in terms of **N** as highlighted in # question B is O( N log N ). Do not write O() anymore.

Note that for question B, we need to know what is inside the black-box of function median, so you are given the following information:

```
# Reference, Steven has unpacked the latest statistics.py of Python 3.10 from
# https://github.com/python/cpython/blob/3.10/Lib/statistics.py
# copied verbatim below:
# FIXME: investigate ways to calculate medians without sorting? Quickselect?
def median(data):
    """Return the median (middle value) of numeric data.
   When the number of data points is odd, return the middle data point.
   When the number of data points is even, the median is interpolated by
   taking the average of the two middle values:
   >>> median([1, 3, 5])
   >>> median([1, 3, 5, 7])
   4.0
   data = sorted(data)
   n = len(data)
   if n == 0:
       raise StatisticsError("no median for empty data")
   if n % 2 == 1:
       return data[n // 2]
   else:
       i = n // 2
       return (data[i - 1] + data[i]) / 2
```

Enter the correct ar	nswer below.	
1		Character Limit: 7
2		Character Limit: 7

**9.** Check all statements that are **True** with respect to the various ADT (List/Stack/Queue/Deque/Priority Queue) topics discussed in Lecture 3a/3b/4a/4b.

(5 marks)

```
We can implement Priority Queue ADT using Python heapq {\tt H} efficiently (all operations in O(log {\tt N})) like this:
```

```
from heapq import heappush, heappop
def enqueue(v): heappush(H, v)
def peek(): return H[0]
def dequeue(): heappop(H)
```

## We can implement PriorityQueue ADT using Python

list L efficiently (both enqueue (v) and dequeue () operations are in O(log N)) like the one shown in BinaryHeapDemo.py in class.

For list, operation is in O(N)

We can implement Queue ADT using Python list  $\bot$  efficiently (all operations in O(1)) like this:

```
def enqueue(v): L.append(v)
def peek(): return L[0]
def dequeue(): L.pop(0)
```

dequeue from head of the list need O(N), since it need to move all other elements

We can implement Stack ADT using Python list  $\bot$  efficiently (all operations in O(1)) like this:

We can implement Queue ADT using Python deque  $\mathbb D$  efficiently (all operations in O(1)) like this:

```
from collections import deque
def enqueue(v): D.append(v)
def peek(): return D[0]
def dequeue(): D.popleft()
```

#### Only I and IV are correct

10. Check all statements that are **True** with respect to the Sorting topics discussed in Lecture 2a/2b.

(5 marks)

False

False

False

The underlying implementation of L.sort() to sort a Python list L is probably Timsort (a variant of Merge Sort).

We can sort a Python list L by calling L.sort() or L = sorted(L)

There is no difference between sorting a Python list L using either L.sort() or sorted(L), i.e., list L will be sorted afterwards using either way.

L.sort() will change the original list, sorted(L) will return a new sorted list

The time complexity of sorting a Python list L containing N integers using either L.sort() or L = sorted(L) is O(N log N).

L.sort() is a stable sorting algorithm.

Only I, II, IV and V are correct

# **IQ Tests**

Both questions in this section are a bit challenging. Do not burn too much time. There are two other application sections after this.

11. Steven claims that there is a better way to store a **complete unweighted** (simple) graph **K**<sub>N</sub> with **N** vertices and **N**\* (N-1)/2 edges than the 3 default graph data structures discussed in class, i.e., not using Adjacency Matrix, Adjacency List, nor Edge List.

If you concur, explain the better way and analyze its space complexity or other benefits!

If you disagree, explain why one of the 3 default data structures is still the most appropriate way to store this kind of graph!

(5 marks)

For a completed unweighted graph, every pair of nodes will have an edge. In this case, storing edge is meaningless, since we can assume that every node have an edge to every other node. Hence, we can simply create a variable to store the number of vertices N, which just a number. We can use the vertex number to calculate all edges in graph. The space complexity to store a number is O(1).

Character World: L110000

Steven claims that there is a better way to sort (in ascending order) a nearly sorted list **L** with **N** non-negative and distinct integers than just using O(**N** log **N**)

Python L.sort() or L = sorted(L) that has been asked in the earlier section (that you can now cannot go back to).

PS: Here, nearly sorted is formally defined as follows: Each of the integer is at most **K** indices away from its target sorted position and  $0 \le K \le \min(N, 7)$ . For example, see list L = [3, 1, 2, 4, 5] with N = 5 and K = 2. Notice that integer 3 (at index 0) is just 2 indices away from its sorted location (index 2). Similarly, notice that integer 1 and 2 are both just 1 index away from their sorted locations.

If you concur, explain the better way and analyze its time complexity!

False

If you disagree, explain why just sorting the whole **N** integers in  $O(N \log N)$  using L.sort() or L = sorted(L) is already the best possible way!

(10 marks)

Use min heap (heapq):

- 1. Create a min heap with size K+1 (Since each integer will not far away from the correct index than K steps. Hence, at anytime in a min heap with size K+1, there must have an element that need to sort)
- 2. Put the first K+1 elements into the min heap
- 3. For each remaining element in list:
  - heappop() from the min heap and put it to the result list
  - heappush() this (remaining) element
- 4. heappop() the elements from min heap to the result list

Time Complexity:

Character Wortd: 120000

For step 1, create a min heap require time O(K logK)

For step 2, for each remain element, heappop and heappush use time O(logK)

There are total N elements, so total time complexity is O(K logK), which is better than O(N logN).

The code is on the last page.

# **Application 1 (Second Last Section) - Queueing at Cashier Lines**

A supermarket has **N** Cashiers. For each cashier, we know the number of people currently queuing to be served by that cashier. These people form a line (a queue), waiting to be served by that cashier.

Now, **M** customers will arrive in the next 1 minute (one by one, and during this next 1 specific minute, none of the cashier will finish processing the head of his/her line - so the line length will only increase). Each of these **M** customers (that come one after another in the next 1 minute) uses the same *greedy strategy*: queue at the back of the current shortest line (the line with the fewest number of people) and if ties, choose any such lines (it doesn't matter for this problem).

Your job is to determine the current number of people in the chosen cashier line every time each of the **M** customers join that cashier line.

The input contains two integers  $\mathbf{N}$  and  $\mathbf{M}$  in the first line ( $\mathbf{N}$  and  $\mathbf{M}$  as explained above), followed by  $\mathbf{N}$  integers in the second line, denoting the number of people in each Cashier line initially. For example:

```
3 4
3 2 4
```

Means that there are N = 3 Cashiers (numbered Cashier 0, 1, and 2) and currently they are about to serve [3, 2, 4] customers, respectively.

We can visualize the initial state of the 3 Cashier lines as follows ('X's are customers already in each Cashier line):

```
Cashier Line = 0 | 1 | 2

X | X | X

X | X | X

X | X | X

X | X

X | X
```

Then, **M** = 4 customers come in the next 1 minute (one after another), so the first new customer (i.e., 'A') will greedily select Cashier 1 (**size 2 - we print 2**, now Cashier 1 size is 3), the Cashier lines become:

```
Cashier Line = 0 | 1 | 2
------

X | X | X

X | X | X

X | X | X

X | A | X

| X
```

Next, the second new customer (i.e., 'B') will then greedily select either Cashier 0 or 1 (both **size 3 - we print 3**, suppose he/she goes to Cashier 0, now Cashier 0 size is 4), the Cashier lines become:

```
Cashier Line = 0 | 1 | 2

X | X | X

X | X | X

X | X | X

X | A | X

B | | X
```

Note that even if 'B' goes to Cashier 1, we still print 3 at this stage, so there is no ambiguity.

Next, the third customer (i.e., 'C') will then greedily select the other line that the second customer doesn't select earlier (in this case, Cashier 1, of **size 3 - we print 3**, now Cashier 1 size become 4 too), the Cashier lines become:

```
Cashier Line = 0 | 1 | 2

X | X | X

X | X | X

X | X | X

X | A | X

B | C | X
```

Finally, the last customer (i.e., 'D') will select any of the Cashier line (as all have **size 4 - we print 4**). Therefore, we output the following  $\mathbf{M} = 4$  lines:

```
2
3
3
4
```

Solve this problem in pseudo-code (you will need time to answer the longer last question after this). You can quote the name of some data structure(s) and/or algorithm(s) that we learned in class verbatim in your explanation (if nothing is modified). Analyze the worst-case time complexity of your proposed solution.

Note that you will only get full marks for this question if your proposed solution is correct and runs in O(M log N) --- so that M and N can be up to 50,000 (lots of customers and Cashier lines); partial if your solution is correct but runs in O(MN) --- so that M and N can be up to just 1000, or just very low mercy marks if your solution is actually incorrect.

Model Answer: 36 words and 259 characters only.

(15 marks)

```
1. # Question 13
import heapq
3.
4. def solution(M, cashier_lst):
 5. Create a new min heap H
6.
      For each cashier:
7.
           heappush(H, cashier_queuevalue)
8. For each customer in M:
           queueValue = heappop(H) + 1
9.
           heappush(H, queueValue)
                                                                 Character Wortd: L10000
10.
11. return H
12.
```

# **Application 2 (The Last Section) - Book Translation**

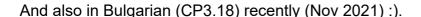
Warning: This is the hardest question in this paper. If you don't spend enough time on this section, you may not be able to complete them on time.

======

As most of you already know, Steven has written a book titled "Competitive Programming 4" (released on July 2020).

It has been translated into Korean (CP3) back in 2017.

In Spanish (CP4) back in early 2021:).



There are a few other book translation projects in the queue, e.g., Chinese, Portuguese, Japanese, and even Indonesian.

Although Steven can code in multiple *programming* languages (e.g., C++, **Python - this course**, Java, MATLAB, etc), he only speaks English and Indonesian (and too lazy to translate his own book to Indonesian).

So, Steven has enlisted (paid) translators to help translate between languages, but obviously they quote different costs. In some cases multiple translations might be needed. For example, if Steven can't find a person who can translate his book from English to Swedish, but have one person who can translate from English to French and another from French to Swedish, then Steven can actually do the translation.

While minimizing the total cost of all these translations is important to you, the most important condition is to minimize each target language's distance (in number of translation steps) from English, since this cuts down on the errors that typically crop up during any translation.

Fortunately, the Method to Solve (MtS) this problem is in Chapter 4 of Steven's book (this statement is True), so you should have no problem in solving this, right?

# 14. Input and potential (graph) data structure(s)

For this problem, the input starts with a line containing two integers  $\mathbf{n}$  and  $\mathbf{m}$  indicating the number of target languages and the number of translators that Steven has contacted  $(1 \le \mathbf{n} \le 10,000, 1 \le \mathbf{m} \le 50,000)$ .

The second line will contain **n** short strings specifying the **n** target languages. Steven wants his book to be translated to *ALL* these **n** target languages.

After this line are  $\mathbf{m}$  lines of the form I1 I2  $\mathbf{c}$  where I1 and I2 are two different languages (two short strings) and  $\mathbf{c}$  (0 <=  $\mathbf{c}$  <= 10,000) (let the unit be 1K SGD) specifying the cost to translate the book between these two languages (in either direction). The languages I1 and I2 are always either "English" (that has 7 characters) or one of the target (real-life) languages (all short language names will be not more than 20 characters), and any pair of languages will appear at most once in the input. The initial book is always written in "English".

### A Sample Input looks like this:

```
4 6
Chinese French Portuguese Swedish
English Chinese 1
English French 1
English Portuguese 5
Chinese Portuguese 1
Portuguese Swedish 5
French Swedish 1
```

Write a simple Python code to handle the Input format as shown above and also show how you are going to store this graph information in a (graph) data structure(s). If you recall, the AM/AL/EL graph data structures that we learned in class only deal with vertices

```
labeled from [0..|V|-1].
                             1 # Question 14
                                def read input():
 (7 marks)
                                    n, m = map(int, input().split())
                                    languages = input().split()
                                    # 创建一个字典来映射语言名称到索引
  Enter your answer here
                              6
                                    language_index = {lang: idx for idx, lang in enumerate(languages, start=1)}
                                    language_index["English"] = 0 # 英语作为索引 0
                              8
                              9
                                   # 初始化邻接列表(AL)
                                    graph = [[] for _ in range(n + 1)] # 加 1 是因为还有英语
                             12
                                    # 读取并存储边信息
                             13
                                    for _ in range(m):
                             15
                                        l1, l2, cost = input().split()
                             16
                                        cost = int(cost)
                                        graph[language_index[l1]].append((language_index[l2], cost))
                             17
                                        graph[language_index[l2]].append((language_index[l1], cost))
                             19
Fill in the blanks 20
                                    return graph, language_index
                             21 print(read_input())
```

# **15**.

(3 marks)

# Output and Sample Test Cases to Confirm Understanding

For each test case, your task is to compute **the minimum cost** to translate Steven's book (from "English") to all the n target languages, subject to the constraints described above, or print "Impossible" (case sensitive, notice capital "I") if it is not possible.

For this Sample Input 1 (shown earlier):

```
4 6
Chinese French Portuguese Swedish
English Chinese 1
English French 1
English Portuguese 5
Chinese Portuguese 1
Portuguese Swedish 5
French Swedish 1
```

### The expected Output is:

Because Steven can translate "English" to "Chinese" for 1K SGD, "English" to "French" also for 1K SGD, then translate the "French" translation one more time into "anathar 1K CCD Einally for "Bhall to "Dantumana"

INCO SWEATSH IOI ANOTHER IN SOD. FINANY, IOI ENGITSH IO FORTUGUESE,

Steven prefers to pay 5K SGD (for translation accuracy) instead of going via "English" -> "Chinese" -> "Portuguese" that "only" costs 2K SGD. Thus, in overall, the total of the translation costs is 1+1+1+5 = 8K SGD. Thus, the output is "8".

======

For this Sample Input 2:

```
2 1
A B
English B 1
```

The expected Output is:

```
Impossible
```

Because there is no suitable direct (or indirect) translator(s) to help us translate Steven's book into language "A".

======

Now the questions that will be automatically graded. For the following Test Case A, B, and C, what should be the output?

#### **Test Case A**

```
5 5

AAA BBB CCC DDD EEE

English AAA 1

BBB English 2

DDD English 3

English EEE 5

English CCC 4
```

The output of Test Case A should be 15 .

======

#### **Test Case B**

```
AAA BBB CCC DDD EEE
English AAA 1
BBB AAA 2
DDD CCC 3
DDD EEE 5
BBB English 4
```

The output of Test Case B should be Impossible

======

#### **Test Case C**

```
2 3
BBB AAA
English AAA 10
BBB AAA 5
English BBB 14
```

	rrect answer below.
1	Please enter a number for this text box.
2	
3	Please enter a number for this text box.
me data stru	d last part, solve this problem in pseudo-code. You can quote the name cture(s) and/or algorithm(s) that we learned in class verbatim in your nothing is modified). Analyze the worst case time complexity of your on
7 marks)	
Enter your answ	31 HOLG
	Character <b>World</b> :
	on code of your previous answer (combining Q14+Q16 answers in Pyth t this part will be graded as 0 (regardless of what you wrote) if your er in Q16 is incorrect. Thus, please explain your algorithm in pseudo-co ad only if you are convinced that your algorithm is correct and fast, then der of this final assessment to code the solution in Python for the last for
evious answe st (in Q16) an	i militari de la como ano constante de la cons
evious answe st (in Q16) an e the remaind arks.	106 words and 725 characters only.
evious answe st (in Q16) an e the remaind arks.	·

16.

**17.** 

```
1. # Question 12
 2. import heapq
 3.
 4. def sort_almost_sorted_list(lst, k):
 5.
       # 初始化最小堆
 6.
       min_heap = []
 7.
       result = []
 8.
 9.
       # 将前 k+1 个元素添加到堆中
10.
       for i in range(min(k+1, len(lst))):
11.
           heapq.heappush(min_heap, lst[i])
12.
13.
       # 遍历列表中剩余的元素
14.
       for i in range(k+1, len(lst)):
15.
           # 提取堆中的最小元素并添加到结果列表
16.
           smallest = heapq.heappop(min_heap)
17.
           result.append(smallest)
18.
           # 将当前元素添加到堆中
19.
20.
           heapq.heappush(min_heap, lst[i])
21.
22.
       # 将堆中剩余的元素添加到结果列表
23.
       while min heap:
24.
           smallest = heapq.heappop(min_heap)
25.
           result.append(smallest)
26.
       return result
27.
```

#### For Question 16, 17

在这个问题中,你需要处理的是图论问题,具体来说是最小生成树(MST)问题的一个变种。下面是如何将问题映射到图论概念和建议的方法:

#### 图表示

、 - 节点:每种语言(包括英语)都是图中的一个节点。 - 边:每对具有翻译成本的语言对构成一条边。成本 c 是边的权重。 - 图类型:这是一个无向图,因为翻译成本在任一方向上都是相同的。

#### 问题解释

- 你需要找到将"英语"节点连接到所有目标语言节点的最小成本。 - 这类似于从"英语"节点开始找到最小生成树(MST)。

与典型的MST问题(任何覆盖所有节点的树都足够)不同的是,这里的树必须包括所有目标语言。如果任何目标语言没有被包 括,结果就是"Impossible"。

#### 建议的算法

- · Prim算法或克鲁斯卡尔算法:两者都适用于在图中找到MST。但是,由于你需要确保包括所有目标语言,你会稍微修改算法。 Prim算法在有特定起始节点的情况下通常更易于实现,这种情况下起始节点为"英语"。 Kruskal's算法可能对于稀疏图更有效,但它本身并不从特定节点开始。

#### 针对目标语言的修改:

- 构建MST后,验证是否包括了所有目标语言。
- 如果任何目标语言缺失,则输出"Impossible"。

实现考虑: 使用优先队列(堆)来有效地选择下一个最小边,适用于Prim算法。 对于Kruskal's算法,根据权重对边进行排序,并使用不相交集数据结构(并查集)进行循环检测。

#### Pseudo Code:

# Question 16

#### 初始化:

- 1. 创建一个图G, 包含所有语言作为节点。
- 2. 设置一个优先队列(最小堆)Q,用于选择最小边。
- 3. 定义一个集合V,用于存储已经访问过的节点。
- 4. 定义一个字典或数组cost,用于存储从英语到每种语言的最小翻译成本。
- 5. 将所有节点的翻译成本初始化为无穷大(除了英语,初始为0)。
- 6. 将所有节点加入优先队列Q(优先级为它们的翻译成本)。

#### Prim 算法:

- 1. 从优先队列中取出成本最低的节点u(初始为英语)。
- 2. 将u加入已访问集合V。
- 3. 对于每一个与u相连的节点v(不在V中):
  - a. 如果u到v的成本小于当前记录的成本:
    - 更新v到英语的最小成本。
    - 更新优先队列中v的优先级。
- 4. 重复步骤1至3, 直到所有目标语言都被访问过或优先队列为空。

#### 检查和输出:

- 1. 检查是否所有目标语言都在V中。
- 2. 如果是,输出V中所有节点的最小成本之和。
- 3. 如果不是,输出"Impossible"。