

Assignment 4: CSE548: Analysis of Algorithms

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Problem 1. Textbook [Kleinberg & Tardos] Chapter 6, page 312, problem #3.

Proof. ...

Input: Ordered Graph, G with nodes $v_1 v_2 \dots v_{n-1} v_n$

Approach for (a): One such example where the algorithm fails is that consider graph G with edges, (v_1, v_2) , (v_1, v_3) , (v_2, v_5) , (v_3, v_4) , (v_4, v_5) , (v_5, v_6) . The algorithm will output 3, where as the desired result is 4. Algorithm starts with node, v_1 and considers v_2 as it is shortest to v_1 ignoring v_3 which reaches v_6 with as depth as possible. While finding the shortest path, we should consider all possible sub problems and proceed with the maximum path length to the next step. The best solution is dynamic approach.

Approach for (b): The problem can be broken down and solved incrementally, bottom-up approach as we did in class for the problem, Longest Increasing Subsequence. let $L[i]$ be the longest path from v_1 to v_i and let $L[i] = -\infty$, if there is no path from v_1 to v_i . while constructing the actual longest path, we store the predecessor of i on a longest path from v_1 to v_i while computing, $L[i]$.

LongestPath(G):

```
for i from 2 to n;  
   $L[i] = -\infty$ ;  
  for  $(v_i, v_j)$  belongs to EdgeSet,  $E$ ;  
    if  $1 + L[j] > L[i]$ ;  
      if  $L[i] = 1 + L[j]$ ;
```

Output: To find the maximum distance from v_1 to v_i , output, $L[i]$.

Correctness: As we take bottom up approach by calculating and solving smaller sub problems for every vertex, this algorithm correctness can be achieved by induction hypothesis.

Time Complexity: As there are two nested loops in the worst case the program runs in $O(n^2)$ running time. \square

Problem 2. Textbook [Kleinberg & Tardos] Chapter 6, page 312, problem #5.

Proof. ...

Input: Long string of English letters, $Y = y_1 y_2 \dots y_{n-1} y_n$ and a Blackbox that compute $Quality(x)$, where x is a possible English word.

Approach: Idea is to break the big string to smaller sub problems by segmenting them into prefix string and a character match. i.e. If segmentation of Y into $y_1 y_2 \dots y_{n-1} y_n$

yields the maximum total quality, then $y_1y_2\dots y_{n-1}$ gives the maximum quality for string Y without y_n . Maintaining a score matrix for every sub problem starting from index 1 to n and incrementally adding the calculating scores should give us the maximum quantity at $\text{score}[n][n]$.

Score(Array[0...n-1]):

```
Score = [[0 for i in range(length(Array)+1)]0 for i in range(length(Array)+1)];
for i in range(length(Array)+1);
    for j in range(i, length(Array)+1);
        Score[i][j] = min(Score[i-1][j-1] + quality(Array[j][n-1]);
return Score[n][n]
```

Output: As the problem counts the quality of each character with the rest of the string, the highest quality is computed at $\text{score}[n][n]$

Correctness: As we are calculating incrementally whether the " i th" character place any quality with the rest of the string keeping the computation of $i-1$ value on string, This incremental computation finds the cost correctly at the end of the i th index.

Time Complexity: from the above algorithm implementation, we can conclude it is $O(n^2)$ running time. \square

Problem 3. Textbook [Kleinberg & Tardos] Chapter 6, page 312, problem #13.

Proof. ...

Input: Graph G with vertices $i_1\dots i_k$ and to find a opportunity cycle in it.

Approach: One way to find the cycle is by running Dijkstra's algorithm or Bellman Ford on the given edge list. Recursively by solving for cycles with small subset of vertices, we can find the existence of cycle for a given graph G using Dynamic programming too.

Approach for (b): The problem can be broken down and solved incrementally, bottom-up approach as we did in class for the problem, Longest Increasing Subsequence. let $L[i]$ be the longest path from v_1 to v_i and let $L[i] = -\infty$, if there is no path from v_1 to v_i . while constructing the actual longest path, we store the predecessor of i on a longest path from v_1 to v_i while computing, $L[i]$.

Algorithm

```
for i from 1 to k;
    Distance[i] =  $-\infty$ ;
    Visited[i] = FALSE;
    Previous[i] = NULL;
Current = v : for v in V;
for v in V;;
    for every vw in E;
        Update Distance[v];
        Mark Previous[v] = {w};
        Mark v = Unvisited();
```

Output: If the combined weight of that cycle gives you "opportunity" then it is an opportunity cycle. If it does not, then there is none. Applying them to all vertices one by one we can check for the cycle.

Correctness: Existence of cycle can be found for every vertex using the distance list, which

proves that the results at all levels are identified and captured properly.

Time Complexity: As we are running the algorithm for n nodes, this is $O(n)$ complexity. \square

Problem 4. Textbook [Kleinberg & Tardos] Chapter 6, page 312, problem #19.

Proof. ...

Input: Three strings, x, y, s

Approach: The problem can be break down into simpler and smaller sub problems. Given that, x, y, s strings, if we are to find whether s is formed by interleaving x and y , we start comparing $s[n]$, the last character of s is equal to x or y and if so by extending it to comparing till $s[1]$ with x and y , we say that the original computation is true. 2D matrix, $Output[length(x) + 1][length(y) + 1]$ is used to save the smaller computations, $output[i][j]$ means that, $s[i + j]$ can be formed using $x[i]$ and $y[j]$ characters.

Output(x, y):

```

output = [[0 for i in range(length(x)+1)]0 for j in range(length(y)+1)];
for i in range(length(x)+1);
  for j in range(i, length(y)+1);
    if output[i-1][j] == True and s[i + j - 1] == x[i - 1];
      output[i][j] = True;
    else if output[i][j-1] == True and s[i + j - 1] == y[j - 1];
      output[i][j] = True;
    else;
      output[i][j] = False;
return output[n][n]
```

Output: If $output[n][n]$ is *True* then the string s can be formed by interleaving two inputs, else not.

Correctness: At any point of time, $output[i][j]$ holds value computed from , $s[i + j]$ and $x[i]$ and $y[j]$ characters. So, as this algorithm is incrementally working by dividing into sub problems and combining them, back. By induction on s we can verify the correctness

Time Complexity: from the above algorithm implemenatation, we can conclude that it runs in $O(n^2)$ running time. \square

Problem 5. Textbook [Kleinberg & Tardos] Chapter 6, page 312, problem #24.

Proof. ...

Input: set of n precincts P_1, P_2, \dots, P_n each containing m registered voters. **Output:** Give an algorithm to determine whether a given set of precincts is susceptible to gerrymandering; the running time of your algorithm should be polynomial in n and m .

I have discussed this problem with Abhishek Reddy Y N Approach: Consider precinct n , and assume precinct n has an party-A voters. We can choose either to include precinct n in our district or not. If we do include precinct n , our system is susceptible to gerrymandering if we can come up with a set of $n/2 - 1$ precincts from among the remaining precincts $1, \dots, n - 1$ that has s_{a_k} total party-A voters. if we exclude precinct n , then system can be susceptible to gerrymandering if we can come up with a set of $n/2$ precincts

from among the remaining precincts 1, . . . , n1 that has s total party-A voters.

Gerrymander(n)

```

G[n][n/2][a-nm/4-1];;
for i from 1 to n;
  G[k][0][0] = True;
  for k from 1 to n;
    for l from 1 to n/2;
      for w from 1 to a-nm/4 -1;
        if k == 1, l == 1, w == a1;
          G[k][l][w] = True;
        else;
          if G[k-1][l-1][w-a1];
            G[k][l][w] = True;
          else;
            if G[k-1][l][w];
              G[k][l][w] = True;
            else;
              G[k][l][w] = False;

```

Output: Return True if the given precincts are successful else False.

Time Complexity: As the algorithm travels through three loops, filling the column takes $O(n^3)$ time. The final traversal which is dominated by first traversal by taking overall time of $O(mn^3)$. □