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Cpts 350

HW 9

1. Design an efficient algorithm to count

the number of paths from  $u$  to  $v$ .

Here is the algorithm written in python.

```
def count_paths(graph, u, v):  
    count = [0] * len(graph)  
    count[u] = 1  
    for node in topological_order(graph):  
        for neighbor in graph[node]:  
            count[neighbor] += count[node]  
    return count[v]
```

```
def topological_order(graph):  
    visited = set()  
    order = []  
  
    def dfs(node):  
        visited.add(node)  
        for neighbor in graph[node]:  
            if neighbor not in visited:  
                dfs(neighbor)  
        order.append(node)  
  
    for node in range(len(graph)):  
        if node not in visited:  
            dfs(node)  
  
    return order[::-1]
```

graph = {

0: [1, 2]

1: [3]

2: [3]

3: [4, 5]

4: []

5: []

$u = 0$   $v = 4$

print("# of paths from,  $u$ ,  $v$ ,  $v$ ,  $v$ ,  
count\_paths(graph,  $u$ ,  $v$ )

The algorithm

has a time complexity

of  $O(V+E)$  where  $V$

is the # of vertices and  $E$

is the # of edges in the graph.

2. Design an efficient algorithm to count the number of good paths from  $u$  to  $v$ . also, in python

```
def count-good-paths(graph, colors, u, v):
    green-count = [0] * len(graph)
    yellow-count = [0] * len(graph)
    for node in topological-order(graph):
        for neighbor in graph[node]:
            if colors[neighbor] == 'green':
                green-count[neighbor] += green-count[node] + 1
                yellow-count[neighbor] += yellow-count[node]
            elif colors[neighbor] == 'yellow':
                green-count[neighbor] += green-count[node]
                yellow-count[neighbor] += yellow-count[node] + 1
    good_paths = 0
    for node in range(len(graph)):
        if green-count[node] > yellow-count[node]:
            good_paths += 1
    return good_paths
```

```
def topological-order(graph):
```

```
    visited = set()
```

```
    order = []
```

```
    def dfs(node):
```

```
        visited.add(node)
```

```
        for neighbor in graph[node]:
```

```
            if neighbor not in visited:
```

```
                dfs(neighbor)
```

```
    order.append(node)
```

```
    for node in range(len(graph)):
```

```
        if node not in visited:
```

```
            dfs(node)
```

```
    return order[::-1]
```

```
graph =
```

```
0 1 2
1 3
2 3
3 4 5
4 5
5 []
```

```
colors =
```

```
0 green
1 yellow
2 green
3 green
4 yellow
5 green
```

```
u = 0
```

```
v = 4
```

```
print(count-good-paths(
    graph, colors, u, v))
```

3. Design an efficient algorithm to count the number of ugly paths from  $u$  to  $v$ .

- \* For each SCC that contains a cycle:
- Check if the color sequence satisfies the regular expression  $\gamma$
  - If it does, return  $\infty$ .
  - If no SCC with a cycle satisfies the regular expression  $\gamma$ , proceed to calculate the count of finite ugly paths
  - Use probability, 1 to count the # of paths that satisfy the regular expression  $\gamma$  from  $u$  to  $v$ .

Define:

- $dp(u, c)$  as the number of paths from node  $u$  to  $v$  that end with color  $c$  and satisfy the regular expression  $\gamma$ .
- $next(u, c)$  as the set of nodes that can be reached from node  $u$  with an edge labeled with color  $c$ .

$$dp(u, c) = \begin{cases} 1 & \text{if } u = v \text{ and } c \text{ satisfies the regex } \gamma \\ 0 & \text{if } u = v \text{ and } c \text{ does not satisfy regex } \gamma \\ \sum_{v' \in next(u, c)} dp(v', c') & \text{if } u \neq v \text{ and } c \text{ satisfies regex } \gamma \\ 0 & \text{if } u \neq v \text{ and } c \text{ does not satisfy regex } \gamma \end{cases}$$

If there exist any infinite paths, the count is infinite; otherwise we use dynamic programming to calculate the count of finite paths.

4.

1. let  $A_1$  and  $A_2$  denote the adjacency matrices of the two graphs respectively
2. Perron's Theorem guarantees the existence of a unique largest real eigenvalue for non-negative irreducible matrices, which can represent strongly connected graphs
3. if  $\lambda_1 > \lambda_2$ , it suggests that, on average, there are more paths in the first graph compared to second graph
4. if  $\lambda_1 < \lambda_2$ , there are more paths in the second graph compared to the first graph.
- if  $\lambda_1 = \lambda_2$  both graphs have a similar number of paths

Graph 1

$$A_1 = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix}$$

$$A_2 = \begin{bmatrix} 0 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 0 \end{bmatrix}$$

The Perron eigenvalue is approx 1.414 for graph 1 and 2 for graph 2



## 5. Mini paper: Utilizing Control Flow Diagram for testing C-program

### 1. path Enumeration:

for each node  $u$  in the Control Flow Diagram, its descendant you can compute the total number of  $C(u)$  of paths from the root of the diagram to  $u$ , while some paths may be infinite due to loops in the programs, the count for each path can be obtained.

### 2. Maximal paths:

By sorting the  $C(u)$ s for all  $u$ , you can identify a node  $u^*$  with maximal  $C(u)$ . This node represents a critical point in the program's control flow, as it is reached by the largest number of paths from the root.

### 3. testing

the maximum path  $u^*$ , you can design targeted test cases to cover various scenarios and branches around this node. By focusing our testing efforts on this crucial path.

$$6. (0+1+10)^*(1101)^*$$

let  $DP[i]$  denote the # of binary strings of length  $i$  that satisfy the given expression

$$DP[i] = DP[i-1] + DP[i-2] + DP[i-3] + DP[i-4]$$

The base cases are  $DP[0] = 1$  (empty string) and  $DP[i] = 0$  for  $i < 0$ .

It has a time complexity of  $O(n)$ ,