

# Asymptotic and Graphs

Julius January 17, 2024

## 1 Asymptotic Notation

### 1.1 Upper Bound, Lower Bound, Tight Bound

**Def.**  $O(g(n)) = \{f(n) | \exists C, n_0 \in \mathbb{R}^+, s.t. \forall n \geq n_0, 0 \leq f(n) \leq Cg(n)\}$

- Every Quadratic function of  $n$  is in  $O(n^2)$
- Every Linear function of  $n$  is in  $O(n^2)$
- Every Cubic function of  $n$  is not in  $O(n^2)$

**Def.**  $\Omega(g(n)) = \{f(n) | \exists C, n_0 \in \mathbb{R}^+ s.t. \forall n \geq n_0, f(n) \geq Cg(n)\}$

- Every Quadratic function of  $n$  is in  $\Omega(n^2)$
- Every Linear function of  $n$  is not in  $\Omega(n^2)$
- Every Cubic function of  $n$  is in  $\Omega(n^2)$

**Def.**  $\Theta(g(n)) = \{f(n) | \exists C_1, C_2, n_0 \in \mathbb{R}^+ s.t. \forall n \geq n_0, 0 \leq C_1g(n) \leq f(n) \leq C_2g(n)\}$

- Every Quadratic function of  $n$  is in  $\Theta(n^2)$
- Every Linear function of  $n$  is not in  $\Theta(n^2)$
- Every Cubic function of  $n$  is not in  $\Theta(n^2)$

**Def.**  $f(n) = o(g(n)) \leftrightarrow f(n) = O(g(n)), f(n) \neq \Theta(g(n))$      $f(n) = \omega(g(n)) \leftrightarrow f(n) = \Omega(g(n)), f(n) \neq \Theta(g(n))$

|                | Worst Case                                     | Best Case                                      |
|----------------|--|--|
| Linear Search  | $O(n), \Omega(n), \Theta(n)$                   | $O(1), \Omega(1), \Theta(1)$                   |
| Binary Search  | $O(\lg n), \Omega(\lg n), \Theta(\lg n)$       | $O(1), \Omega(1), \Theta(1)$                   |
| Insertion Sort | $O(n^2), \Omega(n^2), \Theta(n^2)$             | $O(n), \Omega(n), \Theta(n)$                   |
| Merge Sort     | $O(n \lg n), \Omega(n \lg n), \Theta(n \lg n)$ | $O(n \lg n), \Omega(n \lg n), \Theta(n \lg n)$ |

Comparing growth of functions:  $f_1(n) = 3^n n^2 \lg^5 n$      $f_2(n) = 2^n n^8 \lg n$ . Which function grows faster?

Because exponential component is fastest growing, we can ignore the polynomial and logarithmic components. Two algorithms A and B that solve the same problem have the following worst-case runtime complexities:

Algorithm A:  $O(3^n n^2 \lg^5 n)$ . Algorithm B:  $O(2^n n^8 \lg n)$ .

Which algorithm runs faster? It's hard to define because the O is just an upper bound, and it's hard to define the average case. There are some cases we need to create **hybrid algorithms** to solve the problem.

### 1.2 Average Case Analysis

Why don't we perform average case performance analysis?

Given different environments, the average case performance of an algorithm is not a constant. It really depends on what data set you are using the algorithm to.

## 2 Graphs

### 2.1 BFS and DFS

What are we searching for?

1. To find out if there is a path from node A to node B
2. To find out all nodes that can be reached from A

We turn to use Adjacency List  $int\ h[N], e[M], w[M], ne[M], idx$  to represent the sparse graph and Adjacency Matrix  $int\ g[N][N]$  to represent dense graph. We use **BFS** with **queue** and **distance array** to find the shortest path, while we use **DFS** with **label array**, **recursion** and **backtrace**.

```

1 // Find the Distance from node u to node n
2 int h[N], e[M], ne[M], idx; // adjacency list for sparse graph
3 int d[N]; memset(d, -1, sizeof d); // Initialize -1 for all v in V to represent distance
4 int n, q[N], hh = 0, tt = -1; // array to represent queue
5 int bfs(int u){
6     q[++tt] = u;
7     d[u] = 0;
8     while(hh <= tt){
9         int s = q[hh++];
10        for(int i = h[s]; i != -1; i = ne[i]){
11            int j = e[i]; // j is the node connected to s
12            if(d[j] == -1){
13                d[j] = d[s] + 1;
14                st[j] = true; // mark j as visited
15                q[++tt] = j;
16            }
17        }
18    }
19    return d[n]; // return the distance from node u to node n
20 }
21
22 // Find the Full Permutation of 1 to n
23 int n, path[N];
24 bool st[N];
25 void dfs(int u){
26     if(u == n){
27         for(int i = 0; i < n; i++) printf("%d ", path[i]);
28         puts("");
29     }else{
30         for(int i = 1; i <= n; i++){ // try all possible number nodes
31             if(!st[i]){
32                 path[u] = i;
33                 st[i] = true;
34                 dfs(u+1);
35                 st[i] = false;
36             }
37         }
38     }
39 }

```

## 2.2 Bipartite

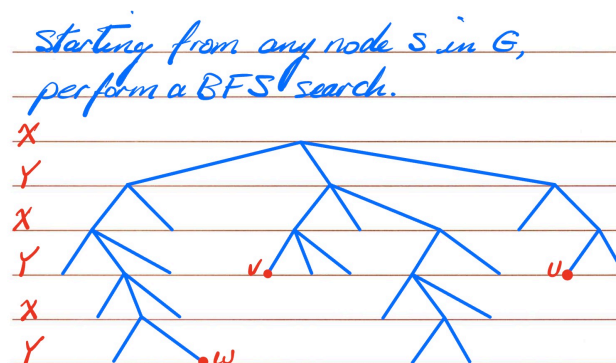
Given an undirected connected graph  $G$ , how do we determine if  $G$  is bipartite, where you could split nodes into two sets and all edges go between two sets?

A graph is bipartite  $\iff$  There is no odd cycle in a graph  $\iff$  There is no contradiction during staining

### BFS Tree for Bipartite:

Run BFS start from any node, say s. Label each node X or Y depending on whether they appear at an odd or even level in BFS tree.  $O(|V| + |E|)$

Then, go through all edges and exam all the labels at the two end of the edge. If all edge ends have different labels, then the graph is bipartite. Otherwise, it's not.



**Facts:** Not possible to have an edge between  $(v, w)$  in  $G$ , since they are more than one level apart in BFS tree.

**Facts:** Edges in  $G$  that end up with the same set  $X$  or  $Y$ , must have both ends at the same level in BFS tree;

```

1  int h[N], e[M], ne[M], idx; // Adjacency List
2  int color[N]; // 0: without color 1: black 2: red
3  bool dfs(int x, int c){ // return true if no contradiction during staining x with color c
4      color[x] = c;
5      for(int i = h[x]; i != -1; i = ne[i]){ // color all the nodes connected with x
6          int j = e[i];
7          if(!color[j]){ // only when without color, call dfs
8              if(!dfs(j, 3 - c)) return false;
9          }else if(color[j] == c){ // with color but conflict
10             return false;
11         }
12     }
13     return true;
14 }
15 void add(int a, int b){ e[idx] = b, ne[idx] = h[a], h[a] = idx++;}
16 int main(){
17     scanf("%d%d", &n, &m); // n: number of nodes, m: number of edges
18     memset(h, -1, sizeof h);
19     while (m -- ){
20         int a, b;
21         scanf("%d%d", &a, &b);
22         add(a, b), add(b, a); // undirected graph
23     }
24     bool flag = true; // true: no conflict
25     for(int i = 1; i <= n; i ++){
26         if(!color[i]){ // only when without color, color black
27             if(!dfs(i, 1)){
28                 flag = false;
29                 break;
30             }
31         }
32     }
33 }

```

**Weakly connected** if for any pair of nodes (v, u), there is a path by ignoring edge directions.

**Connected Graph** if for any pair of nodes (v, u), there is a either a directed path from v to u or from u to v.

**Strongly connected** if for any pair of nodes (v, u), there is a directed path from v to u and a path from u to v.

**Not connected** if a directed graph is not connected in any of the above 3 ways.

**How can we determine if a directed graph G is strongly connected?**

Brute Force Solution  $O(n(n + m)) = O(n^2 + nm)$ : Run BFS/DFS from every node in G. If we can reach all nodes from every node, then the graph is strongly connected.

Better Solution  $O(n + m)$ : First run BFS/DFS from an arbitrary node S see if it can reach all other nodes in G. Then we get the  $G^T$  (reverse all the edges in  $O(n + m)$ ) and run BFS/DFS from node S again to see if it can reach all other nodes in  $G^T$ . If both BFS/DFS can reach all nodes, then the graph is strongly connected, because all the nodes are mutually reachable through node S.

## 2.3 Topological Order

**Topological Order** is the linear ordering of vertices of a graph, such that for every directed edge (u,v), vertex u comes before v in the ordering.

**How to find a topological order in a Directed Acyclic Graph?**

**In-degree** of a node is the number of edges pointing to it.

**Out-degree** of a node is the number of edges pointing from it.

**Kahn's Algorithm**  $O(n + m)$ : First find a node with in-degree 0, then remove it and all edges pointing from it. Decrease the in-degree of the end nodes of the removed edges by 1. Repeat this process until all nodes are removed. The order of removal is the topological order.

```

1  int h[N], e[N], ne[N], idx; // directed graph don't need 2 * N edges
2  int q[N], d[N], hh, tt = -1; // d[N] is the indegree of all nodes
3  void toposort(){
4      for (int i = 1; i <= n; i ++ ){ // find all nodes in d[N] with indegree 0
5          if(!d[i]) q[++tt] = i;
6      }
7      while(hh <= tt){
8          int t = q[hh++];
9          for(int i = h[t]; i != -1; i = ne[i]){
10             int j = e[i];
11             d[j] --; // delete the edge t->j
12             if(!d[j]) q[++tt] = j;
13         }
14     }
15     if(tt == n-1){ // if all nodes are in the queue, then there is a topological order

```

```

16     for(int i = 0; i < n; i++) printf("%d ", q[i]);
17 }else{
18     puts("-1");
19 }
20 }
21 int main(){
22     scanf("%d%d", &n, &m);
23     memset(h, -1, sizeof h);
24     while (m -- ){
25         int a, b;
26         scanf("%d%d", &a, &b);
27         add(a, b);
28         d[b] ++; // remember to add the indegree of node b
29     }
30     toposort();
31     return 0;
32 }

```

### 3 Discussion

**3.1** Arrange the following functions in increasing order of growth rate with  $g(n)$  following  $f(n)$  in your list if and only if  $f(n) = O(g(n))$ .

$$\log n^n, n^2, n^{\log n}, n \log(\log n), 2^{\log n}, \log^2 n, n^{\sqrt{2}}$$

$$\log^2 n \leq 2^{\log n} = n \leq n \log(\log n) \leq \log n^n = n \log n \leq n^{\sqrt{2}} \leq n^2 \leq n^{\log n}$$

**3.2** Suppose that  $f(n)$  and  $g(n)$  are two positive non-decreasing functions such that  $f(n) = O(g(n))$ . Is it true that  $2^{f(n)} = O(2^{g(n)})$ ?

False.  $f(n) = 2n, g(n) = n$ , then  $2^{f(n)} = 2^{2n}, 2^{g(n)} = 2^n, (2^n)^2 \neq O(2^n)$

**3.3** Find an upper bound Big O on the worst case run time of the following code segment. Carefully examine to see if this is a tight upper bound Big O

```

1 void bigOh1(int[] L, int n)
2 while(n > 0)
3     find max(L, n); // finds the max in L[0..n-1]
4     n = n/4;

```

There are  $\log_4 n$  iterations, and each iteration takes  $O(n)$  time. So the upper bound is  $O(n \log n)$ . While for tight upper bound, we have to add up the number of operations. Let's say the first iteration cost  $c \cdot n$ , then  $c \cdot n/4$ , ..., we get  $\sum_{i=0}^{\infty} \frac{cn}{4^i} = 4cn/3$ , we can see actually the tight upper bound is  $O(n)$ , but the operation addition is not always easy to do, so we prefer O to  $\Theta$ .

**3.4** Find an lower bound Big  $\Omega$  on the best case run time of the following code segment. Carefully examine to see if this is a tight lower bound Big  $\Theta$

```

1 string bigOh2(int n)
2 if(n == 0) return "a";
3 string str = bigOh2(n-1);
4 return str + str;

```

There are  $n$  recursion calls and each call takes  $O(1)$  time. So the lower bound is  $\Omega(n)$ . While for tight lower bound, we have to add up the number of operations. There is a string concatenation in each recursion call, and  $\text{bigOh2}(n)$  return  $2^n$  "a", so the tight lower bound is  $\sum_{i=0}^n 2^i = 2^{n+1} = \Omega(2^n)$ , but the operation addition is not always easy to do, so we prefer  $\Omega$  to  $\Theta$ .

**3.5** Pual Erdos himself has a number of Zero. Anyone who wrote a paper with him has a number of one, anyone who wrote a paper with someone who wrote a paper with him has a number of two, and so on. Suppose we have a database of all papers ever written along with their authors.

- Explain how to represent this data as a graph.
- Explain how to find the Erdos number of a given author.
- Explain how to find all the researchers with Erdos number at most two.

Use an undirected graph because they are coauthors. Each node represents a researcher, and each edge represents a paper. Run BFS from Erdos to find all the Erdos number of all researchers, while stop BFS at level two

for researchers with distance at most two.

**3.6 Suppose we are interested in finding the longest simple path in a Directed Acyclic Graph. In particular, we are interested in finding a path that visits all vertices. Given a DAG, give a linear-time algorithm to determine if there is a simple path that visits all vertices.**

Just find the topological order of the graph. If there is a topological order, and there is a simple path follows that order to visit all vertices, then required simple path exists. If not, such path does not exist.

Find a longest path is not an easy thing, but if the graph happens to be acyclic, then we can use topological order to find it.