Analysis of Graph Algorithms

We first analyze *Breadth First Search* with a *rough analysis* of the algorithm in order to develop some intuition. We then build on this analysis to provide a more accurate estimate.

Breadth First Search Rough Analysis

Here is the pseudocode for the algorithm along with the estimated time complexity for each line:

procedure	BFS(G,	src):
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T				
1	for $v \in V$:	V times		
2	unmark v		O(1)	
3	reset parent		O(1)	
4	reset distance		O(1)	
5	reset src distance	O(1)		
6	mark <i>src</i>	O(1)		
7	$Q \leftarrow \texttt{Make-Queue}(src)$	O(1)		
8	while $Q \neq \emptyset$:	V times		
9	$v \leftarrow Pop(Q)$		O(1)	
10	for $u \in Adjacent[v]$:		E_i times	
11	if u is unmarked:		U	O(1)
12	update u			O(1)
13	$\operatorname{mark} u$			O(1)
14	Add(Q , u)			O(1)

The time complexity estimates in the pseudocode above come from the following observations:

- First consider the complexity of the *Queue* operations. If we use a *Linked List* with pointer to the *tail node* the **Queue** operations MakeQueue, Add, and Pop can be implemented efficiently in O(1).
- O(V), Lines 1-5, Init: Initialization is O(V) since the loop is executed once per vertex and we do constant amount of work per vertex.
- O(1), Lines 6-7, Setup: This is O(1) given the previous note about *LinkedList* as a choice for *Queue*.

• Line 8, While Loop: The while loop is executed V times. This may not be clear immediately, but it follows from the fact that each vertex will enter the Queue *exactly once* and will leave the Queue *exactly once*. This is ensured by the marking strategy – once a vertex enters the Queue it is marked which prevents it from entering the Queue twice.

- O(V), Line 9, Finish: Line 9, Pop(Q), which is O(1), is executed V times by the while loop (once per vertex) after which the Queue is empty.
- $O(V \cdot E)$, Lines 10-14, Explore:

The for loop in Line 10 will execute at most E times. The for loop simply looks at the adjacent edges of v, so at most we may have to examine all edges in the graph. The work inside the for loop is O(1) and since these lines are repeated V times by the **while** loop, the total is $O(V \cdot E)$.

Overall the time complexity is:

 $Init(Lines \ 1:5) + Setup(Lines \ 6:7) + Finish(Line \ 9) + Explore(Lines \ 10:14)$

 $O(V) + O(1) + O(V) + O(V \cdot E) = O(V \cdot E)$

Our estimate of $O(V \cdot E)$ suggests that the algorithm is impractical for *dense* graphs. If the graph is fully connected, i.e. every vertex is connected to every other vertex, then we can estimate that $E \approx V \cdot V$ (actually $E = V \cdot (V-1)/2$), which implies that BFS is $O(V \cdot E) = O(V^3)$, i.e. not practical for large graphs.

We now consider a more accurate analysis of *BFS*. The overestimate in our analysis is in *Line* 10. Clearly, we do not need to explore all edges in the graph for each vertex. Instead, for each vertex v we only explore the adjacent edges for this vertex which is some number $Adj_v = E_v$.

The precise analysis breaks Explore(Lines 10: 14) by looking at the time spent to process each vertex during its *Finish* and *Explore* steps. Here is the while loop, shown as the individual V cycles along with work per vertex:

	Work per Adjacent
	Lines 11-14
E_{v_1}	O(1)
E_{v_2}	O(1)
E_{v_3}	O(1)
•••	
E_{v_V}	O(1)
$\sum E_{v_i} \cdot O(1)$	

What remains is to see if we can provide an estimate for $\sum E_{v_i}$. We claim that

$$E_{v_1} + E_{v_2} + E_{v_3} + \dots + E_{v_V} = 2E$$

Even though we do not know the individual terms in the above summation we actually know the overall value of the summation itself. This value is just 2E, since every time we look at an adjacent vertex we effectively look at one of the edges (a, b). Also, each edge (a, b) is looked at *twice* — once from the point of view of vertex *a* and once from the point of view of vertex *b* (we are assuming undirected graph).

Finally, we get

Explore(Lines 10:14) =
$$\sum E_{v_i} \cdot O(1) = 2E \cdot O(1) = O(E)$$

Note that we do not multiply by V for the while loop, since the individual cycles of the while loop are already taken into account in the rows of the table.

Replacing in the earlier analysis, we get:

$$Init(Lines \ 1:5) + Setup(Lines \ 6:7) + Finish(Line \ 9) + Explore(Lines \ 10:14)$$
$$O(V) + O(1) + O(V) + O(E) = O(V + E)$$

This is much better than the first estimate. If the graph is fully connected, in which case $E \approx V^2$, we get that BFS is $O(V + V^2)$ or $O(V^2)$, not $O(V^3)$.

The analysis of Prim's algorithm is almost identical to the analysis of *BFS*. The only difference comes from the fact that we use *PriorityQueue*, so the complexity of the operations is no longer O(1).

There are two operations to consider: **Pop-Min** and **Decrease-Key**. The second operation should rearrange the *PriorityQueue* after the *comparison value/key* is updated. In *Prim* this value is the **distance** field of a vertex.

Here are a few possibilities for *PriorityQueue* and the complexity of operations:

	Make-PQ	Pop-Min	Decrease-Key
BST	$O(n\log n)$	$O(\log n)$	$O(\log n)$ but need to Remove+Add
Heap	O(n)	$O(\log n)$	$O(\log n)$ only Push-Up
Fibonacci Heap	O(n)	$O(\log n)$	O(1)
LinkedList	O(n)	O(n)	O(1)

The *LinkedList* is given as an exercise. In principle, *PriorityQueue* may manage the items by storing them in a *sorted* or *unsorted* linked list, although this may not be ideal implementation in most cases.

The analysis below will be based on the *Heap* option. Effectively, this means that in the algorithm analysis, which is essentially the analysis of BFS, we need to multiply by a factor of log V.

proc 1 2 3	cedure $Prim(G, src)$: for $v \in V$: unmark v reset parent	V times	O(1) O(1)	
4	reset distance		O(1)	
5	reset <i>src</i> distance	O(1)	0(1)	
6	$PQ \gets \texttt{Make-PQ}(V(G))$	O(V)		
7	$T \gets \texttt{Make-Graph}(V(G))$	O(1)		
8	while $Q \neq \emptyset$:	V times		
9	$v \leftarrow \texttt{Pop-Min}(Q)$		$O(\log V)$	
10	mark v		O(1)	
11	Add(T , $edge(u,v)$)		O(1)	
$12 \\ 13 \\ 14 \\ 15$	for $u \in Adjacent[v]$: if u is unmarked and $dist[u] > weight(v \rightarrow u)$: $dist[u] \leftarrow weight(v \rightarrow u)$, i.e. Decrease-Key update parent		E times	$O(1) \\ O(\log V) \\ O(1)$

The analysis is as follows:

- O(V), Lines 1-5, Init: Same as in BFS.
- O(V), Lines 6-7, Setup: Building a *Heap* with all vertices is O(V) and building an empty *Graph* is O(1).
- Line 8, While Loop: Executes V times. The *PriorityQueue* starts with all vertices and one vertex is popped per cycle.
- $O(V \log V)$, Lines 9-11, Finish Same as BFS but removing from the *PriorityQueue* is $O(\log V)$.
- $O(E \log V)$, Lines 9-11, Finish Similar to BFS in the table below.

	Popped	#Adjacent	Work per Adjacent, Lines 13-15
			dominated by
			Line 14:Decrease-Key
_	v_1	E_{v_1}	$O(\log V)$
	v_2	E_{v_2}	$O(\log V)$
	v_3	E_{v_3}	$O(\log V)$
		•••	
	v_V	E_{v_V}	$O(\log V)$
	Total	$\sum E_{v_i} \cdot O(\log V)$	

Finally, we get

$$Explore(Lines \ 12:15) = \sum E_{v_i} \cdot O(\log V) = 2E \cdot O(\lg V) = O(E \log V)$$

and overall

$$Init(Lines \ 1:5) + Setup(Lines \ 6:7) + Finish(Lines \ 9:11) + Explore(Lines \ 12:15)$$

$$O(V) + O(V) + O(V \log V) + O(E \log V) = O((V + E) \log V)$$

The same analysis also applies to *Dijkstra's Algorithm*, since the two algorithms only differ in what value they compute for the dist[] field. Thus, both algorithms are $O((V + E) \log V)$ if *Heap* is used.

Note: Line 14 requires special attention, since it depends on the choice of *PriorityQueue*. As an exercise repeat the analysis for the various *PriorityQueue* options.