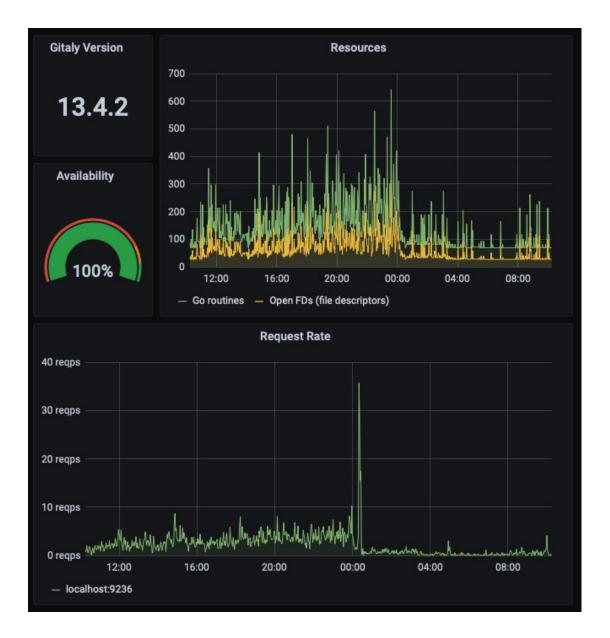


Graphs

Prof. Darrell Long
CSE 13S

Is this a graph?

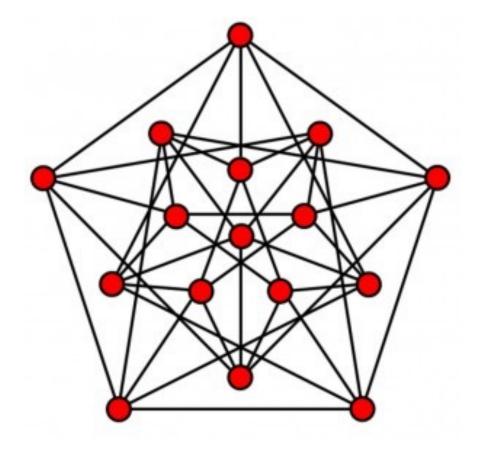
- Yes, this is commonly called a graph, but it is not what we are talking about here...
- We will call these depictions of numerical quantities *plots*.

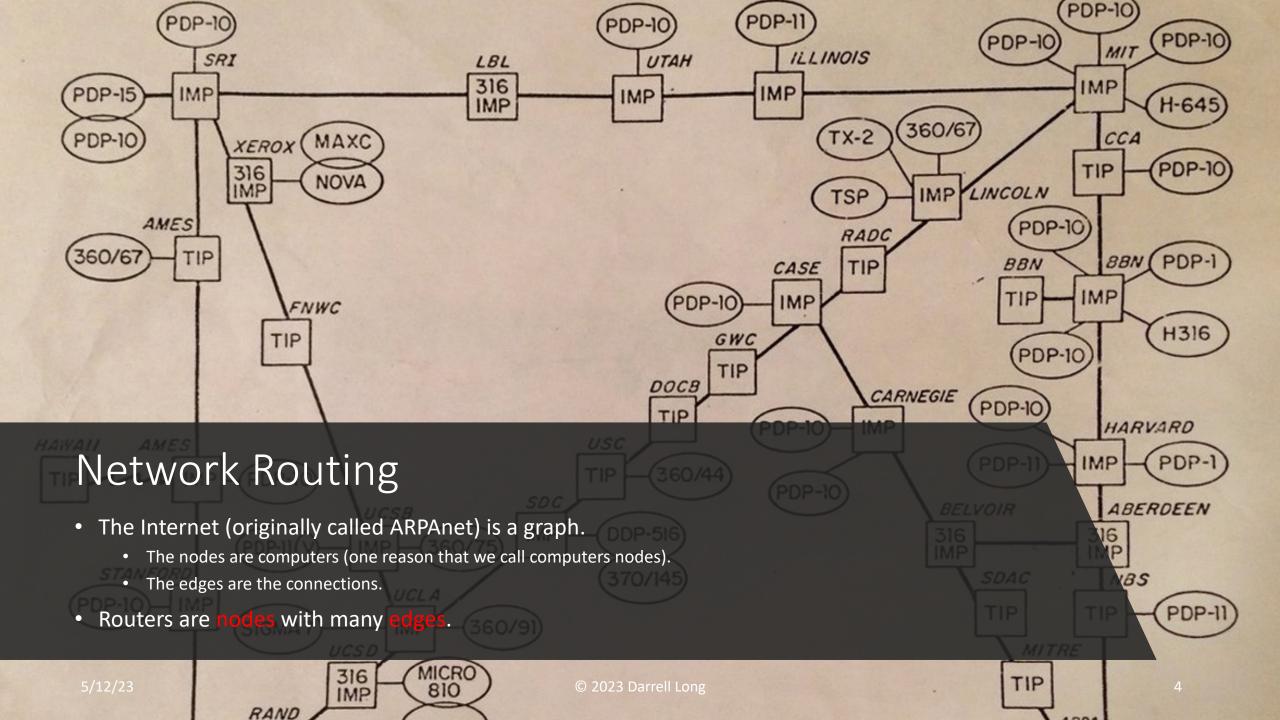


Is this a graph?

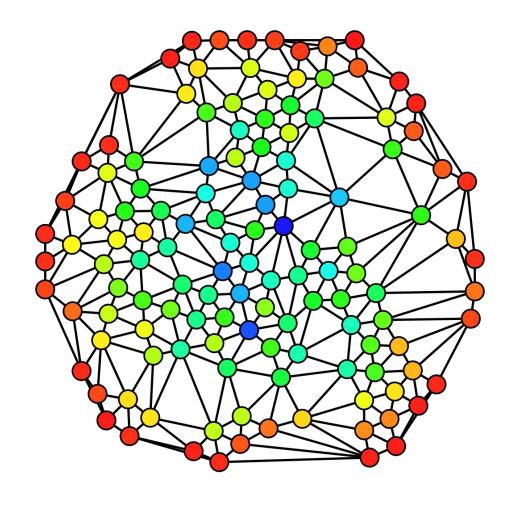
- This is an example of what we will call a *graph*.
- There is an entire branch of mathematics dedicated to *graph* theory.
- Graphs of this type are used in mathematics, computer science, physics, and even sociology.

Vertices and Edges

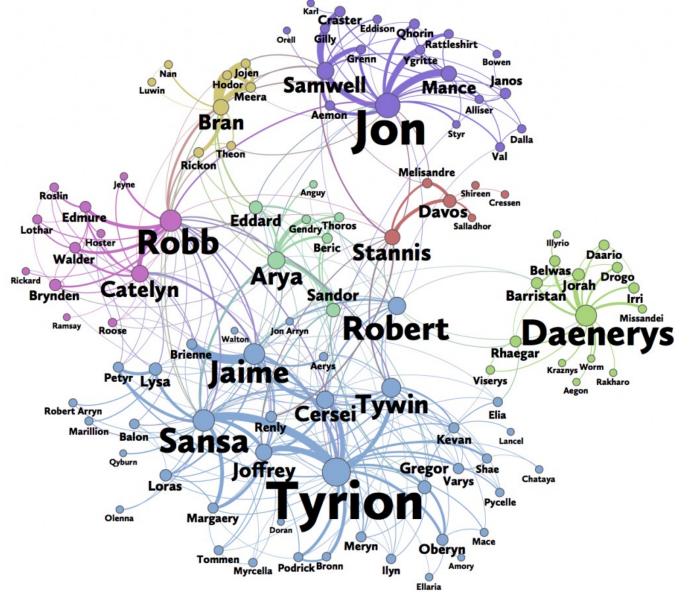




Social Networks



Network of Throne



A Formal Definition

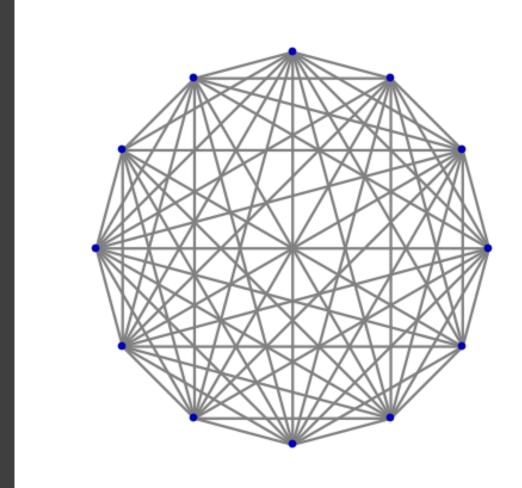
- We formally define a graph *G* as follows:
 - $G = \langle V, E \rangle$
 - A graph is defined by its vertices and edges.
- V is the set of vertices.
 - $V = \{v_1, v_2, \dots, v_n\}$
- *E* is the set of edges.
 - Each edge is a tuple of vertices.
 - $E = \{ \langle v_i, v_j \rangle, \langle v_p, v_q \rangle, \cdots, \langle v_s, v_t \rangle \}$



- Edges may have a direction, $n_1 \rightarrow n_2$, and we call that a directed graph.
- Edges may have no direction (or both directions), $n_1 \leftrightarrow n_2$, and we call that an undirected graph.
- The edges may have weights, which represent capacity, strength, or cost.

Representing a graph

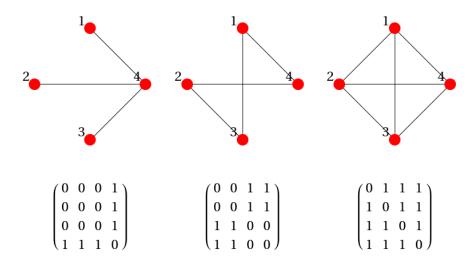
- Adjacency Matrix
 - n × n matrix
 - Binary: edges present or absent
 - Weighted: $n \neq 0$
- Adjacency List
 - Column array for nodes
 - Linked list of edges from each node
 - May contain weights



Computed by Wolfram|Alpha

Adjacency Matrix (Used in Assignment 4)

- A non-zero entry in $M_{i,j}$ means there is an edge $n_i \rightarrow n_j$.
- A matrix that is symmetric around the diagonal represents an *undirected* graph.
- The entry can specify not only the existence of an edge, but also its weight.
- Requires $O(n^2)$ space.
 - Sparse matrix techniques can improve it.



Computed by Wolfram|Alpha

A graph in C

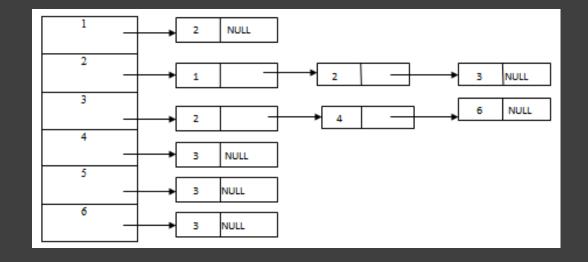
- Uses dynamically sized adjacency matrix.
- Adding an edge is O(1).
- Checking for the existence of an edge is also O(1).

```
typedef struct {
   uint32_t vertices; // Number of vertices in graph.
   uint32_t **matrix; // Adjacency matrix.
} Graph;
// Constructor for a graph with dynamically sized adjacency matrix.
// Performs no error checking with memory allocation.
Graph *graph create(uint32 t vertices) {
    Graph *g = (Graph *)malloc(sizeof(Graph));
   g->vertices = vertices;
   g->matrix = (uint32_t **)calloc(vertices, sizeof(uint32_t *));
    for (uint32_t i = 0; i < vertices; i += 1) {</pre>
        g->matrix[i] = (uint32_t *)calloc(vertices, sizeof(uint32_t));
   return g;
// Adds an edge <i, j> of weight k.
// Does not check if vertices are valid.
void graph_add_edge(Graph *g, uint32_t i, uint32_t j, uint32_t k) {
   g->matrix[i][j] = k;
// Checks if edge <i, j> exists.
// Does not check if vertices are valid.
bool graph_has_edge(Graph *g, uint32_t i, uint32_t j) {
    return g->matrix[i][j] > 0;
```

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Adjacency List

- Each node is represented as an entry in a column vector.
 - Each entry is the head of a linked list.
- The list elements contain:
 - The destination node, and
 - The *weight* of the edge.
- Why would you prefer this over an adjacency matrix?
 - An adjacency matrix is $O(n^2)$ space,
 - An adjacency list will be more space efficient for sparse graphs.



Another graph in **C**

- Uses adjacency lists.
 - Implemented using linked lists.
- Adding an edge is O(1).
- Checking for the existence of an edge is O(n).
 - Must traverse entire list!

```
typedef struct {
   uint32 t vertices; // Number of vertices in graph.
   LinkedList **lists; // Adjacency lists.
 Graph:
Graph *graph_create(uint32_t vertices) {
   Graph *g = (Graph *)malloc(sizeof(Graph));
   g->vertices = vertices;
   g->lists = (LinkedList **)calloc(vertices, sizeof(LinkedList *));
    for (uint32_t i = 0; i < vertices; i += 1) {</pre>
       g->lists[i] = ll create(); // Initialize each to be empty adjacency list.
   return g;
void graph_add_edge(Graph *g, uint32_t i, uint32_t j, uint32_t k) {
   ll_insert(g->lists[i], j, k);
// Does not check if vertices are valid.
bool graph_has_edge(Graph *g, uint32_t i, uint32_t j) {
   // Searches the list. looking for a node that contains j as the vertex.
    return ll_find(g->lists[i], j);
```

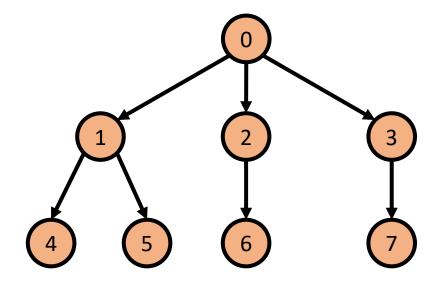
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Basic Graph Algorithms

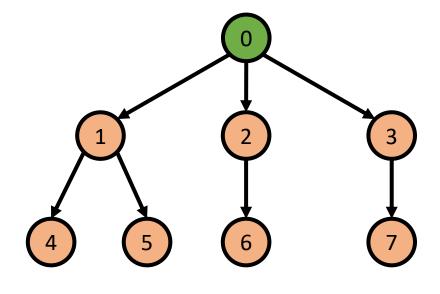
- We often want to find something in a graph.
- Two ways of searching a graph:
 - 1. Breadth-first search
 - Uses a queue.
 - Explore the set of vertices immediately reachable.
 - Then repeat the process for each vertex in the set.
 - Also known as "level-order" traversal.
 - 2. Depth-first search
 - Uses recursion or a stack.
 - Search as far as possible before backing up.
 - We will showcase *iterative* DFS using a stack.

Depth-First Search

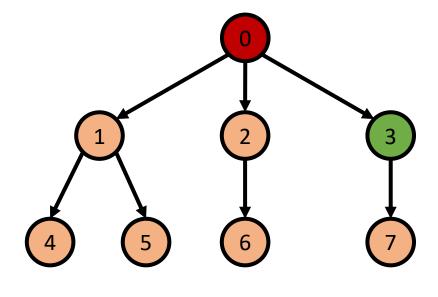
- With a Tree
 - Basic recursive traversal
 - Visit all of the nodes
- With a Graph
 - Can be recursive
 - But need to mark which nodes have been visited
 - The recursive algorithm's "Base Case"



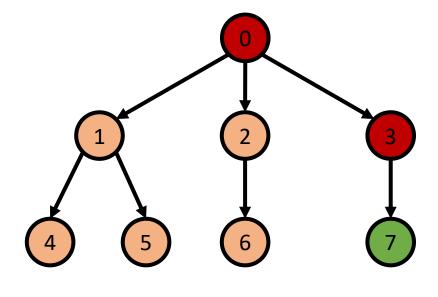
Stack: 0



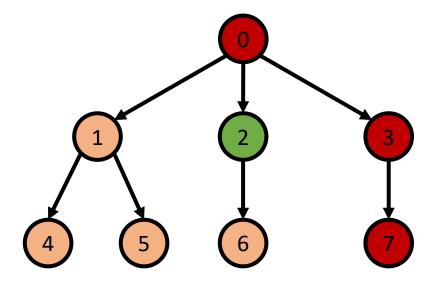
Stack: 1, 2, 3



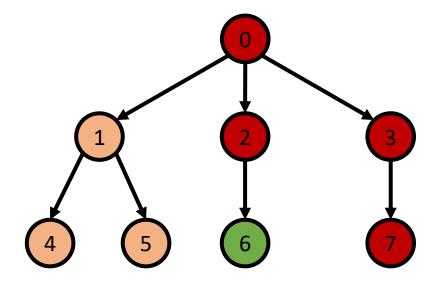
Stack: 1, 2, 7



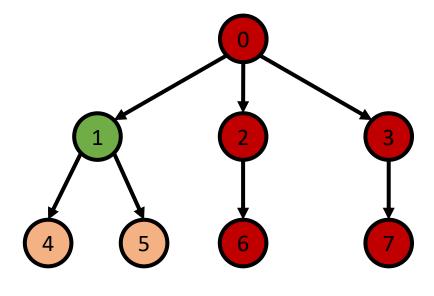
Stack: 1, 2



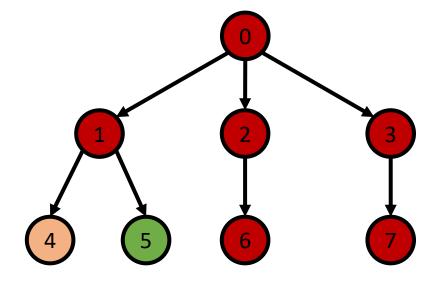
Stack: 1, 6



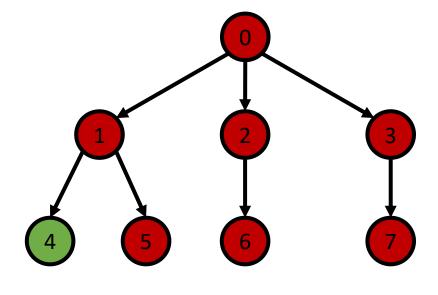
Stack: 1



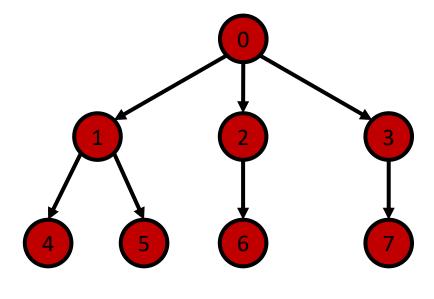
Stack: 4, 5



Stack: 4



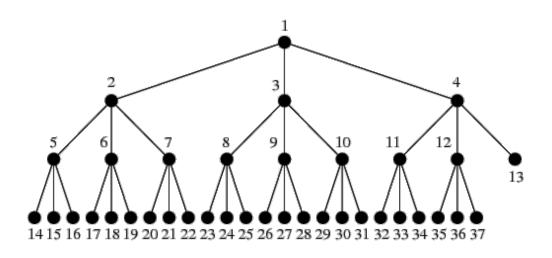
Stack: 4



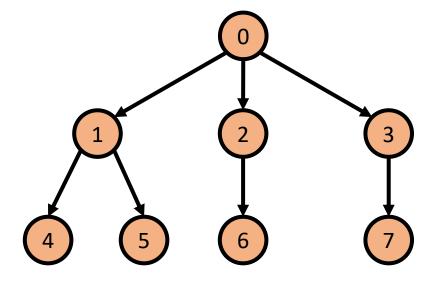
Stack:

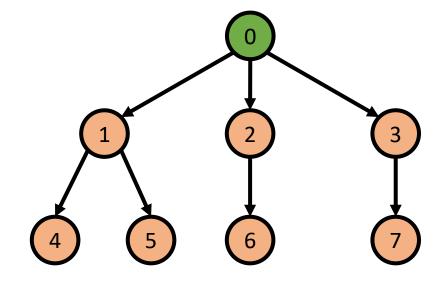
Trees

- Trees are form of acyclic graph
 - Acyclic means that if you follow the edges, there are *no loops* (cycles).
- You will often hear the term *DAG*, which stands for:
 - <u>D</u>irected
 - <u>A</u>cyclic
 - <u>G</u>raph

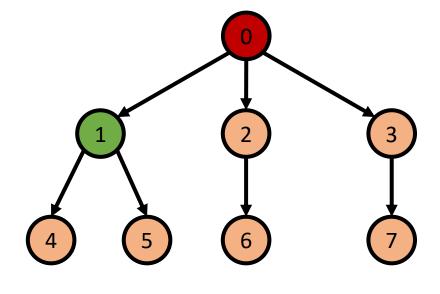


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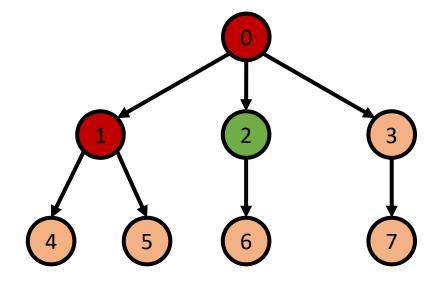




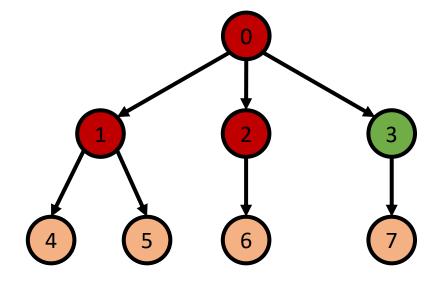
Queue: 1, 2, 3



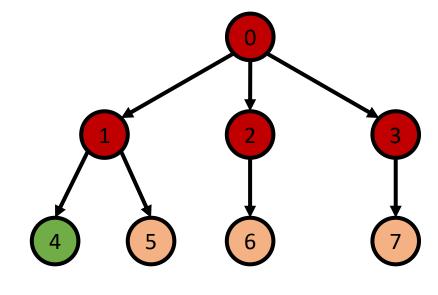
Queue: 2, 3, 4, 5



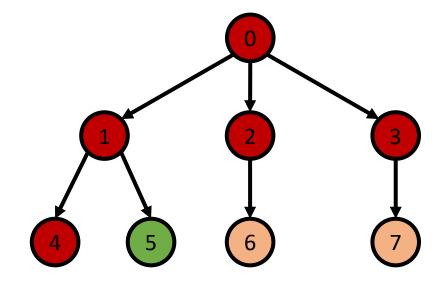
Queue: 3, 4, 5, 6



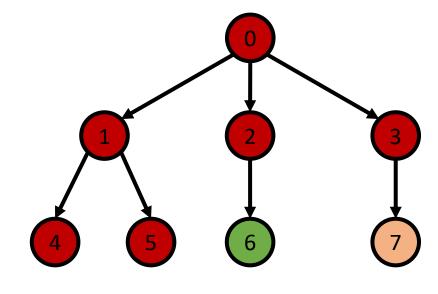
Queue: 4, 5, 6, 7

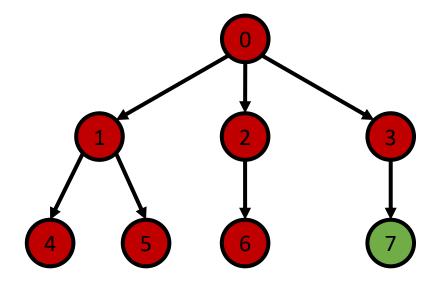


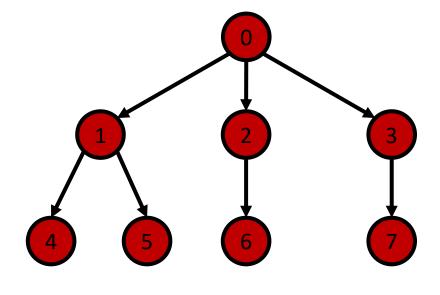
Queue: 5, 6, 7



Queue: 6, 7

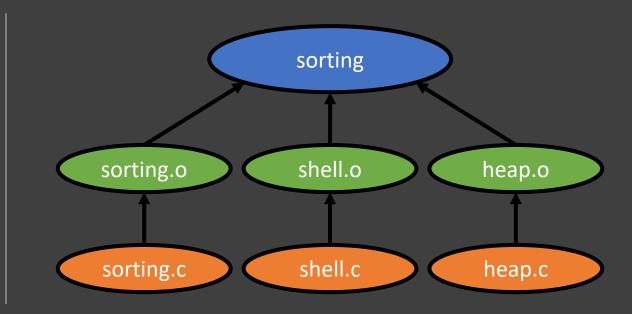






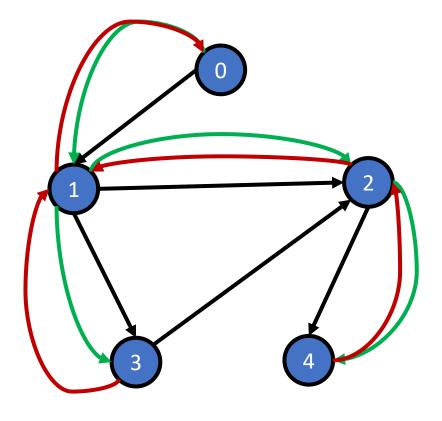
Topological Sort

- Linearly orders vertices in a DAG such that, for any directed edge <u,v>, u appears before v.
- Typically used to indicate ordering of dependencies.
 - Such as in a Makefile!
- Implemented using DFS:
 - Whenever a vertex is finished, it is prepended to a list.
 - Finished vertex: no more exploration possible from that vertex.
- Topological orderings are *not* unique.
 - Could be more than one valid topological ordering.



Modified DFS to Topologically Sort (Recursive)

```
void toposort(Graph *g, uint32_t v, Set *visited, LinkedList *order) {
   *visited = set_insert(*visited, v);
    for (uint32_t u = 0; u < graph_vertices(g); u += 1) {</pre>
       if (!set_member(*visited, u) && graph_has_edge(g, v, u)) {
           toposort(g, u, visited, order);
   ll_insert(order, v);
int main(void) {
   Graph *g = graph_create(8);
   graph_add_edge(g, 0, 1, 5); // <0, 1>, weight = 5
   graph_add_edge(g, 1, 2, 2); // <0, 1>, weight = 2
   graph_add_edge(g, 1, 3, 2); // <0, 1>, weight = 2
   graph_add_edge(g, 3, 2, 3); // <0, 1>, weight = 3
   graph_add_edge(g, 2, 4, 4); // <0, 1>, weight = 4
   Set visited = set_empty();
   LinkedList *order = ll_create();
   toposort(g, 0, &visited, order);
   return 0;
```



Order: $0 \longrightarrow 1 \longrightarrow 3 \longrightarrow 2 \longrightarrow 4$

Kahn's Algorithm

Input: A directed acyclic graph $G = \langle V, E \rangle$. **Output:** A list L containing a topological ordering of G.

```
KAHN(G)
 1 L \leftarrow [] // List storing topological ordering.
 2 S \leftarrow \{\} // Set of vertices with no incoming edges.
 3 for v \in V
           if \nu has no incoming edges
                 S \leftarrow S \cup \{v\}
 5
     while |S| > 0
           remove some v \in S
           L \leftarrow L + [v]
           for each u \in V where e is \langle v, u \rangle and e \in E
                 E \leftarrow E - \{e\}
10
                 if u has no incoming edges
11
                      S \leftarrow S \cup \{u\}
12
13
     return L
```

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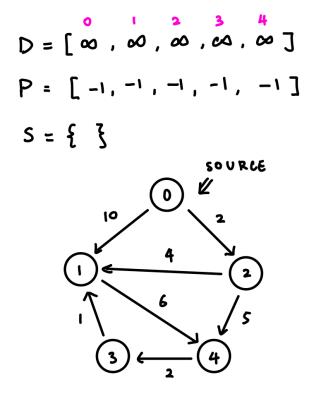
Single-Source Shortest Paths (SSSP)

- Assume some graph $G = \langle V, E \rangle$ and source vertex $s \in V$.
 - We want the shortest path from s to any $v \in V$.
- SSSP algorithms:
 - Bellman-Ford
 - Dijkstra's
 - We will focus on Dijkstra's algorithm.



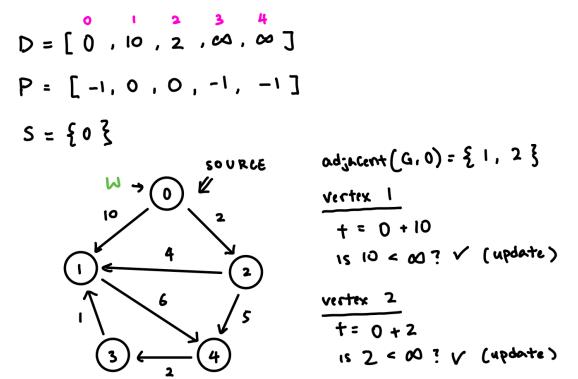
- 1. Distances from s for each vertex $v \in V$.
- 2. Predecessors for each $v \in V$ in shortest paths from s.

```
DIJKSTRA(G, s)
 1 S \leftarrow \{s\}
 2 for v \in G
                         // Distance from source is infinite.
                          // Predecessor of vertex is unknown.
 5 D[s] \leftarrow 0
                           // Distance from s to s.
     while S \neq V
          w \leftarrow \text{vertex in V} - S \text{ such that D}[w] \text{ is minimum}
          S \leftarrow S \cup \{w\}
          for each v \in ADJACENT(G, w)
               t \leftarrow D[w] + WEIGHT(G, w, v)
10
11
               if t < D[v]
                                     // If new distance from s to \nu is shorter.
12
                     D[v] \leftarrow t
                                    // Update distance.
13
                     P[v] \leftarrow w
                                     // Update predecessor.
14 return D, P
```



- 1. Distances from s for each vertex $v \in V$.
- 2. Predecessors for each $v \in V$ in shortest paths from s.

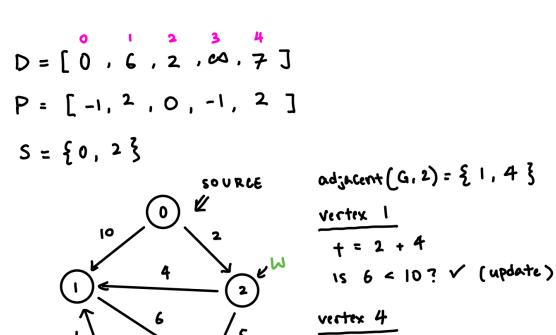
```
DIJKSTRA(G, s)
 1 S \leftarrow \{s\}
 2 for v \in G
                           // Distance from source is infinite.
           P[v] \leftarrow -1
                           // Predecessor of vertex is unknown.
    D[s] \leftarrow 0
                           // Distance from s to s.
     while S \neq V
          w \leftarrow \text{vertex in V} - S \text{ such that D}[w] \text{ is minimum}
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12
                     D[v] \leftarrow t
                                     // Update distance.
13
                      P[v] \leftarrow w
                                      // Update predecessor.
14 return D, P
```



Input: A directed graph $G = \langle V, E \rangle$ and source vertex $s \in V$. **Output:**

- 1. Distances from s for each vertex $v \in V$.
- 2. Predecessors for each $v \in V$ in shortest paths from s.

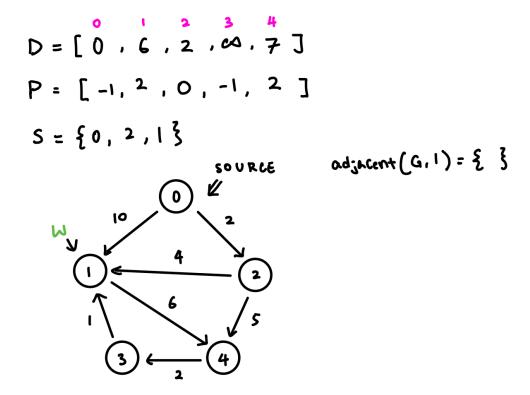
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14 return D, P
```



15 7 < 00 ? / (update)

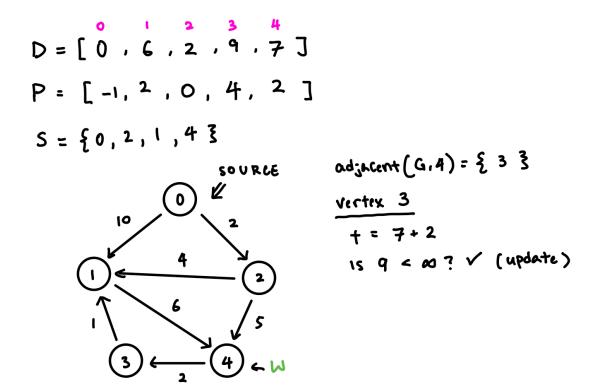
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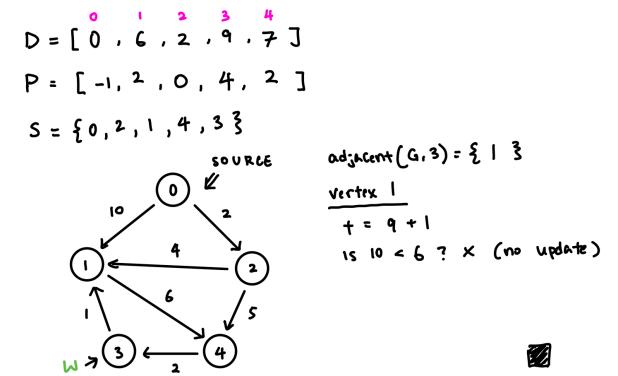
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13
                     P[v] \leftarrow w
                                     // Update predecessor.
14 return D, P
```



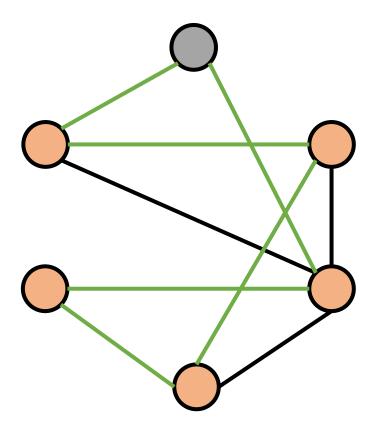
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           for each v \in ADJACENT(G, w)
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10
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12
                      D[v] \leftarrow t
                                      // Update distance.
13
                      P[v] \leftarrow w
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14 return D, P
```



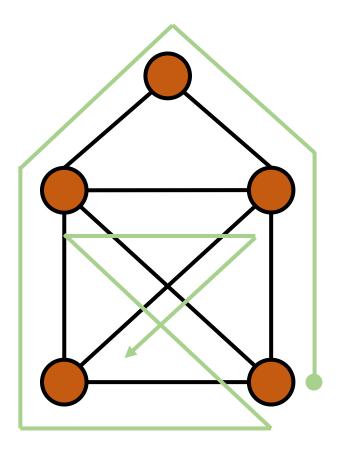
Hamiltonian Path

- A path in an undirected or directed graph that visits each *vertex* exactly once.
 - Must start from an origin vertex and end up back at the origin.



Eulerian Path

- A path in an undirected or directed graph that visits each edge exactly once.
 - Must start from an origin vertex and end up back at the origin.



Summary

- Graphs pervade computer science.
 - Shortest path finding
 - Graph coloring
 - Network flow
 - Dependency ordering
 - ... and so much more!
- Come in undirected and directed forms.
- Used generally to indicate relationships between entities.
- Can be represented using either an adjacency matrix or using adjacency lists.