A common application of graphs is to model some sort of transportation network (airports, roads, etc.).

If we are currently at a location s and wish to travel to a location t, then we may want to find a path which starts at s and ends at t. There may be many such paths, and in the application, some paths may be much more expensive to follow than others.

We can assign a weight to each edge $\{u, v\}$ of the graph which represents the cost of moving from location u to location v. Now consider some path p from s to t. The weight of the path w(p) is the sum of the edge weights along the path.

In this setting, we would be interested in computing a **shortest path** from s to t. A shortest path is a path of minimum weight from s to t. Let $\delta(u, v)$ denote the weight of a shortest path between any two vertices u and v in the graph $(\delta(u, v) = \infty)$ if there are no paths from u to v).

In some applications we may want to have negative weights on an edge. Note that if there is a negativeweight cycle, then some shortest paths may not exist.

5.

The following theorem is crucial in the computation of shortest paths:

Theorem
A subjant of a shortest path is a shortest path.

Shorker path from u to v.

If not, replace this path 2 with a shortest path from u to v to obtain a shorter path from s to E, a Contradiction.

The following theorem is known as the *triangle in*equality and is also important in the computation of shortest paths:

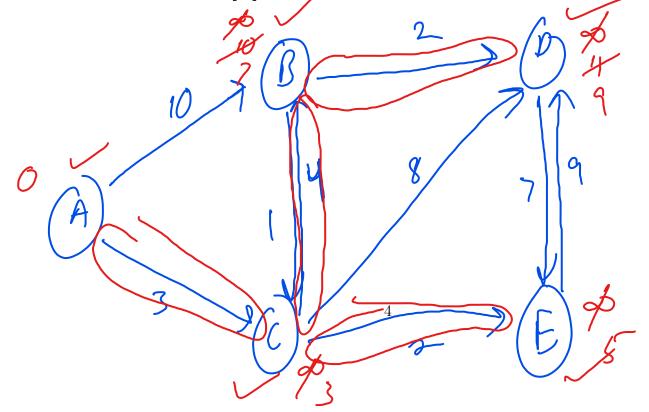
Theorem For all $u, v, x \neq V$ we have $\int (u, v) \leq \int (u, x) + \int (x, v)$ $\int (u, v) \leq \int (u, x) + \int (x, v)$ $\int (u, v) \leq \int (u, x) + \int (x, v)$ $\int (u, v) \leq \int (u, x) + \int (x, v)$

We will now consider the *single-source shortest path* problem in which we are given a graph with a designated source vertex s, and we wish to compute the shortest path weights $\delta(s, v)$ for each $v \in V$.

We will assume all edge weights are nonnegative so that we will not have any negative weight cycles.

Dijkstra's Algorithm. Idea: Greedy. For each vertex v, we maintain an upper bound d[v] on $\delta(s, v)$.

- 1. Maintain a set S of vertices whose shortest path weights from s are known, that is $d[v] = \delta(s, v)$.
- 2. At each step, add the vertex $v \in V \setminus S$ whose d[v] is minimal.
- 3. Update d[u] for any vertex u adjacent to v.



Example of Dijkstra's Algorithm:

Theorem: (i) For all $v \in S : d[v] = \delta(s, v)$. (ii) For all $v \notin S : d[v]$ is the weight of a shortest path from s to v that uses only vertices in S (besides v itself).

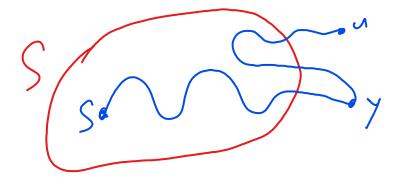
The implication of this theorem is that Dijkstra's algorithm terminates with $d[v] = \delta(s, v)$ for each $v \in V$ (because each S = V at the end of the algorithm).

The proof is by induction. It is clearly true in the base case $(d[s] = 0 \text{ and } d[v] = \infty \text{ for all } v \neq s)$. Assume (i) and (ii) are true before an iteration, and we will show it remains true after another iteration.

Let u be the vertex added to S in this iteration. So $d[u] \leq d[v]$ for all $v \in V \setminus S$.

(i) We need to Show d[vi] = f(Siu).

Suppose it is not. Then there is a path p from 5 to 4 with w(p) < d[4]. So there must be a variex YEVS on the path p.



We should have picked y instead of u, so such a part does not exist. Thus I[u] = f(s,u).

(ii) Let Ut S. Let p be a shortest path from 5 to V
that only were vertices in S (besides v itself).

2 cases:

1) p does not contain 4. Then (ii) is true by inductive hypothers

2) p contains u. (5) > 0 > 0 > 0 > 0 > 0

So (ii) is true.

The running time of Dijkstra's algorithm is $O(m \log n)$ when maintaining $V \setminus S$ as a priority queue.

Now consider the *unweighted case* in which we want to find a path with the smallest number of edges.

Certainly Dijkstra's Algorithm can still work (we can just set the weight of each edge to be 1). Can we do better?

Idea: do a modified BFS search starting from s. Recall BFS traverses the graph in "layers" where each layer will be the same distance from s.

So far we have only been computing the *length* of a shortest path. What if we want to know what the shortest path actually is?

We can build a shortest path tree similarly to how we built MST, BFS, and DFS trees (we remember the "predecessor" for each vertex during the computation).