Minimum Spanning Trees II: Kruskal's algorithm COMS20010 (Algorithms II)

John Lapinskas, University of Bristol

But we already have a good algorithm: Prim's runs in $O(|E|\log |E|)$ time, and we can't beat O(|E|) time since we need to read the input.

But we already have a good algorithm: Prim's runs in $O(|E|\log |E|)$ time, and we can't beat O(|E|) time since we need to read the input.

So why am I bothering to teach you Kruskal's algorithm as well?

• It has slightly better constant factors (debatably);

But we already have a good algorithm: Prim's runs in $O(|E|\log |E|)$ time, and we can't beat O(|E|) time since we need to read the input.

- It has slightly better constant factors (debatably);
- It's an application of a cool and useful data structure;

But we already have a good algorithm: Prim's runs in $O(|E|\log |E|)$ time, and we can't beat O(|E|) time since we need to read the input.

- It has slightly better constant factors (debatably);
- It's an application of a cool and useful data structure;
- The **really** good algorithms use ideas from both Prim and Kruskal. (More on this next week!)

But we already have a good algorithm: Prim's runs in $O(|E|\log |E|)$ time, and we can't beat O(|E|) time since we need to read the input.

- It has slightly better constant factors (debatably);
- It's an application of a cool and useful data structure;
- The **really** good algorithms use ideas from both Prim and Kruskal. (More on this next week!)
- Interviewers might expect you to know it...

But we already have a good algorithm: Prim's runs in $O(|E|\log |E|)$ time, and we can't beat O(|E|) time since we need to read the input.

- It has slightly better constant factors (debatably);
- It's an application of a cool and useful data structure;
- The **really** good algorithms use ideas from both Prim and Kruskal. (More on this next week!)
- Interviewers might expect you to know it... **

Input: A connected weighted graph G = ((V, E), w). Output: A minimum spanning tree of G.

A minimum spanning tree is a subtree T of G covering all of G's vertices, whose total weight $\sum_{e \in E(T)} w(e)$ is as small as possible.

Input: A connected weighted graph G = ((V, E), w). Output: A minimum spanning tree of G.

A minimum spanning tree is a subtree T of G covering all of G's vertices, whose total weight $\sum_{e \in E(T)} w(e)$ is as small as possible.

We are even more greedy than in Prim's algorithm.



Input: A connected weighted graph G = ((V, E), w). Output: A minimum spanning tree of G.

A minimum spanning tree is a subtree T of G covering all of G's vertices, whose total weight $\sum_{e \in E(T)} w(e)$ is as small as possible.

We are even more greedy than in Prim's algorithm.



Input: A connected weighted graph G = ((V, E), w). Output: A minimum spanning tree of G.

A minimum spanning tree is a subtree T of G covering all of G's vertices, whose total weight $\sum_{e \in E(T)} w(e)$ is as small as possible.

We are even more greedy than in Prim's algorithm.



Input: A connected weighted graph G = ((V, E), w). Output: A minimum spanning tree of G.

A minimum spanning tree is a subtree T of G covering all of G's vertices, whose total weight $\sum_{e \in E(T)} w(e)$ is as small as possible.

We are even more greedy than in Prim's algorithm.



Input: A connected weighted graph G = ((V, E), w). Output: A minimum spanning tree of G.

A minimum spanning tree is a subtree T of G covering all of G's vertices, whose total weight $\sum_{e \in E(T)} w(e)$ is as small as possible.

We are even more greedy than in Prim's algorithm.



Input: A connected weighted graph G = ((V, E), w). Output: A minimum spanning tree of G.

A minimum spanning tree is a subtree T of G covering all of G's vertices, whose total weight $\sum_{e \in E(T)} w(e)$ is as small as possible.

We are even more greedy than in Prim's algorithm.



Input: A connected weighted graph G = ((V, E), w). Output: A minimum spanning tree of G.

A minimum spanning tree is a subtree T of G covering all of G's vertices, whose total weight $\sum_{e \in E(T)} w(e)$ is as small as possible.

We are even more greedy than in Prim's algorithm.



Input: A connected weighted graph G = ((V, E), w). Output: A minimum spanning tree of G.

A minimum spanning tree is a subtree T of G covering all of G's vertices, whose total weight $\sum_{e \in E(T)} w(e)$ is as small as possible.

We are even more greedy than in Prim's algorithm.



Input: A connected weighted graph G = ((V, E), w). Output: A minimum spanning tree of G.

A minimum spanning tree is a subtree T of G covering all of G's vertices, whose total weight $\sum_{e \in E(T)} w(e)$ is as small as possible.

We are even more greedy than in Prim's algorithm.



Input: A connected weighted graph G = ((V, E), w). Output: A minimum spanning tree of G.

A minimum spanning tree is a subtree T of G covering all of G's vertices, whose total weight $\sum_{e \in E(T)} w(e)$ is as small as possible.

We are even more greedy than in Prim's algorithm.



Input: A connected weighted graph G = ((V, E), w). Output: A minimum spanning tree of G.

A minimum spanning tree is a subtree T of G covering all of G's vertices, whose total weight $\sum_{e \in E(T)} w(e)$ is as small as possible.

We are even more greedy than in Prim's algorithm.



Input: A connected weighted graph G = ((V, E), w). Output: A minimum spanning tree of G.

A minimum spanning tree is a subtree T of G covering all of G's vertices, whose total weight $\sum_{e \in E(T)} w(e)$ is as small as possible.

We are even more greedy than in Prim's algorithm.



Input: A connected weighted graph G = ((V, E), w). Output: A minimum spanning tree of G.

A minimum spanning tree is a subtree T of G covering all of G's vertices, whose total weight $\sum_{e \in E(T)} w(e)$ is as small as possible.

We are even more greedy than in Prim's algorithm.



Input: A connected weighted graph G = ((V, E), w). Output: A minimum spanning tree of G.

A minimum spanning tree is a subtree T of G covering all of G's vertices, whose total weight $\sum_{e \in E(T)} w(e)$ is as small as possible.

We are even more greedy than in Prim's algorithm.



Input: A connected weighted graph G = ((V, E), w). Output: A minimum spanning tree of G.

A minimum spanning tree is a subtree T of G covering all of G's vertices, whose total weight $\sum_{e \in E(T)} w(e)$ is as small as possible.

We are even more greedy than in Prim's algorithm.



Input: A connected weighted graph G = ((V, E), w). Output: A minimum spanning tree of G.

A minimum spanning tree is a subtree T of G covering all of G's vertices, whose total weight $\sum_{e \in E(T)} w(e)$ is as small as possible.

We are even more greedy than in Prim's algorithm.



Input: A connected weighted graph G = ((V, E), w). Output: A minimum spanning tree of G.

A minimum spanning tree is a subtree T of G covering all of G's vertices, whose total weight $\sum_{e \in E(T)} w(e)$ is as small as possible.

We are even more greedy than in Prim's algorithm.



Input: A connected weighted graph G = ((V, E), w). Output: A minimum spanning tree of G.

A minimum spanning tree is a subtree T of G covering all of G's vertices, whose total weight $\sum_{e \in E(T)} w(e)$ is as small as possible.

We are even more greedy than in Prim's algorithm.



Input: A connected weighted graph G = ((V, E), w). Output: A minimum spanning tree of G.

A minimum spanning tree is a subtree T of G covering all of G's vertices, whose total weight $\sum_{e \in E(T)} w(e)$ is as small as possible.

We are even more greedy than in Prim's algorithm.



Input: A connected weighted graph G = ((V, E), w). Output: A minimum spanning tree of G.

A minimum spanning tree is a subtree T of G covering all of G's vertices, whose total weight $\sum_{e \in E(T)} w(e)$ is as small as possible.

We are even more greedy than in Prim's algorithm.



Input: A connected weighted graph G = ((V, E), w). Output: A minimum spanning tree of G.

A minimum spanning tree is a subtree T of G covering all of G's vertices, whose total weight $\sum_{e \in E(T)} w(e)$ is as small as possible.

We are even more greedy than in Prim's algorithm.



Input: A connected weighted graph G = ((V, E), w). Output: A minimum spanning tree of G.

A minimum spanning tree is a subtree T of G covering all of G's vertices, whose total weight $\sum_{e \in E(T)} w(e)$ is as small as possible.

We are even more greedy than in Prim's algorithm.



Input: A connected weighted graph G = ((V, E), w). Output: A minimum spanning tree of G.

A minimum spanning tree is a subtree T of G covering all of G's vertices, whose total weight $\sum_{e \in E(T)} w(e)$ is as small as possible.

We are even more greedy than in Prim's algorithm.



Input: A connected weighted graph G = ((V, E), w). Output: A minimum spanning tree of G.

A minimum spanning tree is a subtree T of G covering all of G's vertices, whose total weight $\sum_{e \in E(T)} w(e)$ is as small as possible.

We are even more greedy than in Prim's algorithm.



Input: A connected weighted graph G = ((V, E), w). Output: A minimum spanning tree of G.

A minimum spanning tree is a subtree T of G covering all of G's vertices, whose total weight $\sum_{e \in E(T)} w(e)$ is as small as possible.

We are even more greedy than in Prim's algorithm.



Kruskal's algorithm: Formal version and correctness

Input: A connected weighted graph G = ((V, E), w). **Output:** A minimum spanning tree of G.

A minimum spanning tree is a subtree T of G covering all of G's vertices,

whose total weight $\sum_{e \in E(T)} w(e)$ is as small as possible.

Kruskal's algorithm: Formal version and correctness

Input: A connected weighted graph G = ((V, E), w). Output: A minimum spanning tree of G.

A minimum spanning tree is a subtree T of G covering all of G's vertices,

whose total weight $\sum_{e \in E(T)} w(e)$ is as small as possible.

Formally: Let e_1, \ldots, e_m be the edges of G, with $w(e_1) \leq \cdots \leq w(e_m)$.

Let $T_0 = (V, \emptyset)$ be the empty graph on V. Given T_i , let $T_{i+1} = T_i + e_{i+1}$ if this is a forest, or T_i otherwise.

Kruskal's algorithm is to calculate and return T_m . Why does this work?

Kruskal's algorithm: Formal version and correctness

Input: A connected weighted graph G = ((V, E), w). Output: A minimum spanning tree of G.

A minimum spanning tree is a subtree T of G covering all of G's vertices, whose total weight $\sum_{e \in E(T)} w(e)$ is as small as possible.

Formally: Let e_1, \ldots, e_m be the edges of G, with $w(e_1) \leq \cdots \leq w(e_m)$.

Let $T_0 = (V, \emptyset)$ be the empty graph on V. Given T_i , let $T_{i+1} = T_i + e_{i+1}$ if this is a forest, or T_i otherwise.

Kruskal's algorithm is to calculate and return T_m . Why does this work?

T_m is a spanning tree: Suppose not, for a contradiction.

By construction, T_m has no cycles and $V(T_m) = V$, so T_m must have at least two components C_1 and C_2 (both of which are trees).

Since G is connected, it must contain an edge e_i between C_1 and C_2 . $T_m + e_i$ contains no cycles since C_1 and C_2 are trees, so nor does $T_{i-1} + e_i$, so we should have $e_i \in E(T_i)$ — a contradiction.

 T_m is minimum: Again we will use an exchange argument.

Let S be a minimum spanning tree with $S \neq T_m$. We will turn S into a tree S^+ with one more edge in common with T_m , and $w(S^+) \leq w(S)$.

By repeating the process, we prove: $w(S) \ge w(S^+) \ge \cdots \ge w(T_m)$, and we're done.

Goal: Turn an arbitrary minimum spanning tree S into a new tree S^+ , with one more edge in common with T_m and with $w(S^+) \le w(S)$.

Goal: Turn an arbitrary minimum spanning tree S into a new tree S^+ , with one more edge in common with T_m and with $w(S^+) \le w(S)$.

Key fact: If we add an edge to *S*, we create exactly one cycle *C*. If we then remove any other edge from *C*, the result is a tree. (See problem sheet.)

Goal: Turn an arbitrary minimum spanning tree S into a new tree S^+ , with one more edge in common with T_m and with $w(S^+) \le w(S)$.

Key fact: If we add an edge to *S*, we create exactly one cycle *C*. If we then remove any other edge from *C*, the result is a tree. (See problem sheet.)

Since $T_m \neq S$ and both have |V| - 1 edges by the FLoT, there must be an edge $e \in E(T_m) \setminus E(S)$. Let C be the unique cycle in S + e.



Goal: Turn an arbitrary minimum spanning tree S into a new tree S^+ , with one more edge in common with T_m and with $w(S^+) \le w(S)$.

Key fact: If we add an edge to *S*, we create exactly one cycle *C*. If we then remove any other edge from *C*, the result is a tree. (See problem sheet.)

Since $T_m \neq S$ and both have |V| - 1 edges by the FLoT, there must be an edge $e \in E(T_m) \setminus E(S)$. Let C be the unique cycle in S + e.



Goal: Turn an arbitrary minimum spanning tree S into a new tree S^+ , with one more edge in common with T_m and with $w(S^+) \le w(S)$.

Key fact: If we add an edge to *S*, we create exactly one cycle *C*. If we then remove any other edge from *C*, the result is a tree. (See problem sheet.)

Since $T_m \neq S$ and both have |V| - 1 edges by the FLoT, there must be an edge $e \in E(T_m) \setminus E(S)$. Let C be the unique cycle in S + e.



Since T_m has no cycles, there must be some edge $f \in E(C) \setminus E(T_m)$.

Goal: Turn an arbitrary minimum spanning tree S into a new tree S^+ , with one more edge in common with T_m and with $w(S^+) \le w(S)$.

Key fact: If we add an edge to *S*, we create exactly one cycle *C*. If we then remove any other edge from *C*, the result is a tree. (See problem sheet.)

Since $T_m \neq S$ and both have |V| - 1 edges by the FLoT, there must be an edge $e \in E(T_m) \setminus E(S)$. Let C be the unique cycle in S + e.



Since T_m has no cycles, there must be some edge $f \in E(C) \setminus E(T_m)$.

Goal: Turn an arbitrary minimum spanning tree S into a new tree S^+ , with one more edge in common with T_m and with $w(S^+) \le w(S)$.

Key fact: If we add an edge to *S*, we create exactly one cycle *C*. If we then remove any other edge from *C*, the result is a tree. (See problem sheet.)

Since $T_m \neq S$ and both have |V| - 1 edges by the FLoT, there must be an edge $e \in E(T_m) \setminus E(S)$. Let C be the unique cycle in S + e.



Since T_m has no cycles, there must be some edge $f \in E(C) \setminus E(T_m)$. Since Kruskal's algorithm added e instead of f, we have $w(e) \le w(f)$. We therefore take $S^+ = S - f + e$.

John Lapinskas

Goal: Turn an arbitrary minimum spanning tree S into a new tree S^+ , with one more edge in common with T_m and with $w(S^+) \le w(S)$.

Key fact: If we add an edge to *S*, we create exactly one cycle *C*. If we then remove any other edge from *C*, the result is a tree. (See problem sheet.)

Since $T_m \neq S$ and both have |V| - 1 edges by the FLoT, there must be an edge $e \in E(T_m) \setminus E(S)$. Let C be the unique cycle in S + e.



Since T_m has no cycles, there must be some edge $f \in E(C) \setminus E(T_m)$. Since Kruskal's algorithm added e instead of f, we have $w(e) \le w(f)$. We therefore take $S^+ = S - f + e$.

John Lapinskas

Goal: Turn an arbitrary minimum spanning tree S into a new tree S^+ , with one more edge in common with T_m and with $w(S^+) \le w(S)$.

Key fact: If we add an edge to *S*, we create exactly one cycle *C*. If we then remove any other edge from *C*, the result is a tree. (See problem sheet.)

Since $T_m \neq S$ and both have |V| - 1 edges by the FLoT, there must be an edge $e \in E(T_m) \setminus E(S)$. Let C be the unique cycle in S + e.



Since T_m has no cycles, there must be some edge $f \in E(C) \setminus E(T_m)$. Since Kruskal's algorithm added e instead of f, we have $w(e) \le w(f)$. We therefore take $S^+ = S - f + e$.