

Minimum Spanning Tree Problems

AU4606: Network Optimization

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- Global minimum cut problems
 - MA ordering
 - Randomization algorithms

- 1 Minimum spanning tree problem: formulation
- 2 Algorithms for minimum spanning tree problems
- 3 Matroids and greedy algorithms

- 1 Minimum spanning tree problem: formulation
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What is a minimum spanning tree problem

Minimum spanning tree problem

Given an undirected connected graph $G = (N, A)$ and costs c_{ij} for all arcs $(i, j) \in A$, find a spanning tree $T = (N, A')$ of G such that

$$c(T) = \sum_{(i,j) \in T} c_{ij}$$

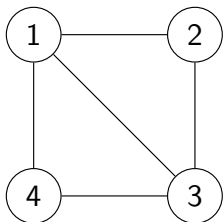
is minimized.

Spanning tree $T = (N, A')$

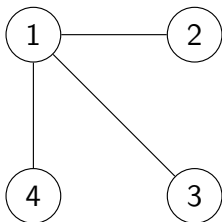
- $A' \subset A$
- T is a connected acyclic graph (tree)

Minimum spanning tree is not necessarily unique (e.g., a graph with uniform arc weights)

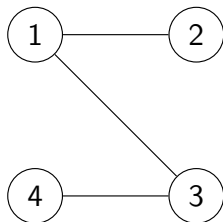
Minimum spanning tree problem: examples



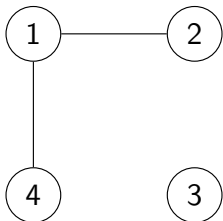
(a) Graph G



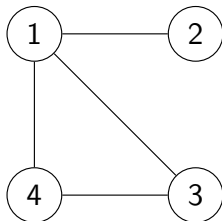
(b) Spanning tree



(c) Spanning tree



(d) Not a spanning tree



(e) Not a spanning tree

- 1 Minimum spanning tree problem: formulation
- 2 Algorithms for minimum spanning tree problems
- 3 Matroids and greedy algorithms

Optimality condition

Cut optimality conditions

A spanning tree T^* is a minimum spanning tree if and only if for every tree arc $(i, j) \in T^*$, $c_{ij} \leq c_{kl}$ for every arc (k, ℓ) contained in the cut formed by deleting (i, j) from T^* .

Every arc in an MST is a min cost arc across the cut defined by removing it

Property of minimum spanning tree

Let F be a subset of arcs of some minimum spanning tree. Let S be a set of nodes of some component of F . Then some minimum spanning tree contains all arcs in F and a minimum cost arc $(i, j) \in (S, \bar{S})$.

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Path optimality conditions

A spanning tree T^* is a minimum spanning tree if and only if for every nontree arc $(k, \ell) \notin T^*$, $c_{ij} \leq c_{kl}$ for every arc (k, ℓ) contained in the path in T^* connecting nodes k and ℓ .

Kruskal's algorithm: procedure

- Start from a spanning tree, check path optimality condition, swap arcs

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Algorithm Kruskal's algorithm

```
1: Sort the arcs by their costs  $(e_1, e_2, \dots, e_m)$ 
2:  $L \leftarrow \emptyset, k \leftarrow 1$ 
3: while  $|L| < n - 1$  do
4:   if Arcs in  $L \cup \{e_k\}$  do not form a cycle then
5:      $L \leftarrow L \cup \{e_k\}$ 
6:   end if
7:    $k \leftarrow k + 1$ 
8: end while
```

Correctness of Kruskal's algorithm

Kruskal's algorithm finds a minimum spanning tree for undirected graphs.

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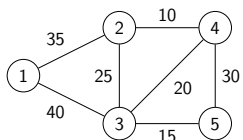
Correctness of Kruskal's algorithm

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Detecting cycles

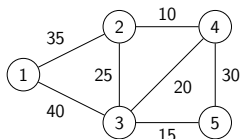
- Maintain a collection of node sets N_1, N_2, \dots
- For (k, ℓ) , if both k and ℓ belong to same set, then we have a cycle

Kruskal's algorithm: example

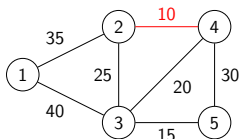


(a) Graph G

Kruskal's algorithm: example

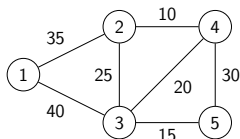


(a) Graph G

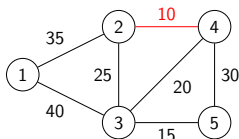


(b) Iteration 1

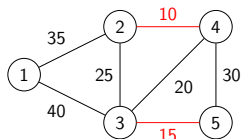
Kruskal's algorithm: example



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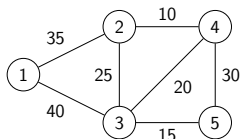


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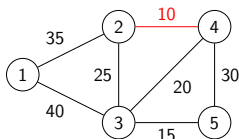


(c) Iteration 2

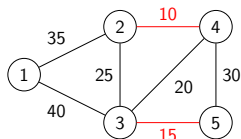
Kruskal's algorithm: example



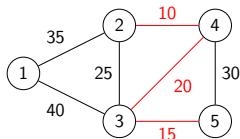
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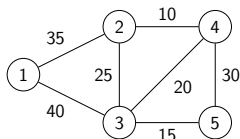


(c) Iteration 2

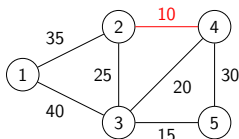


(d) Iteration 3

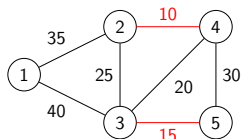
Kruskal's algorithm: example



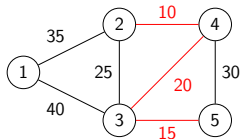
(a) Graph G



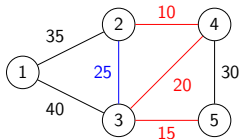
(b) Iteration 1



(c) Iteration 2

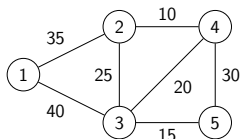


(d) Iteration 3

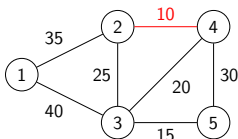


(e) Iteration 4

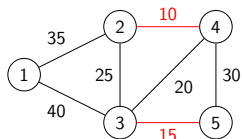
Kruskal's algorithm: example



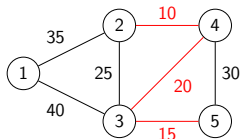
(a) Graph G



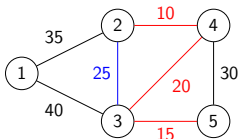
(b) Iteration 1



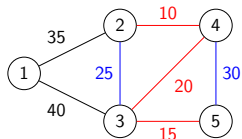
(c) Iteration 2



(d) Iteration 3

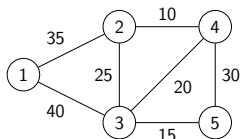


(e) Iteration 4

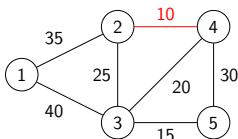


(f) Iteration 5

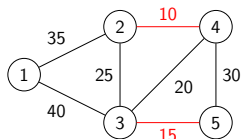
Kruskal's algorithm: example



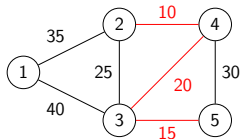
(a) Graph G



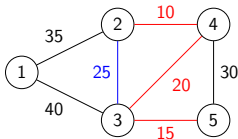
(b) Iteration 1



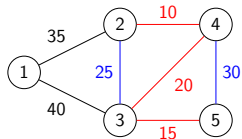
(c) Iteration 2



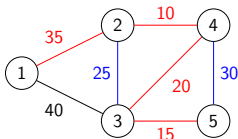
(d) Iteration 3



(e) Iteration 4



(f) Iteration 5



(g) Iteration 6

Prim's algorithm: procedure

Property of minimum spanning tree

Let F be a subset of arcs of some minimum spanning tree. Let S be a set of nodes of some component of F . Then some minimum spanning tree contains all arcs in F and a minimum cost arc $(i, j) \in (S, \bar{S})$.

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Algorithm Prim's algorithm

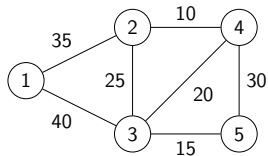
- 1: $S \leftarrow \emptyset$
 - 2: $T \leftarrow \emptyset$
 - 3: Add an arbitrary node to S
 - 4: **while** $|S| < n$ **do**
 - 5: $(i^*, j^*) \in \operatorname{argmin}_{(i,j) \in (S, \bar{S})} c_{ij}$
 - 6: $S \leftarrow S \cup j^*$
 - 7: $T \leftarrow T \cup \{(i^*, j^*)\}$
 - 8: **end while**
-

- Maintain a tree T on S , add arc with min cost to T from (S, \bar{S})

Correctness of Prim's algorithm

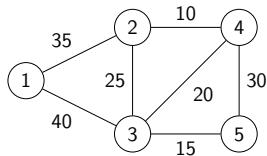
Prim's algorithm finds a minimum spanning tree for undirected graphs.

Prim's algorithm: example

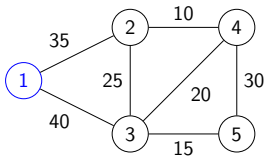


(a) Graph G

Prim's algorithm: example

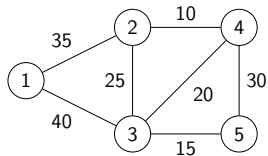


(a) Graph G

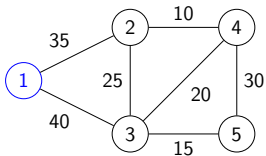


(b) Initialization

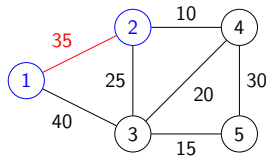
Prim's algorithm: example



(a) Graph G

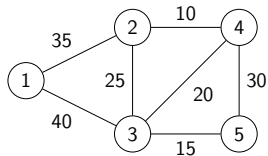


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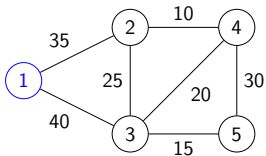


(c) Iteration 1

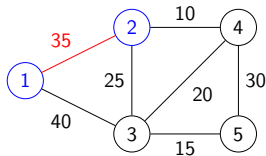
Prim's algorithm: example



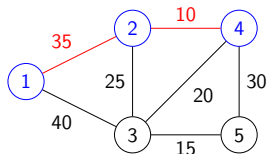
(a) Graph G



(b) Initialization

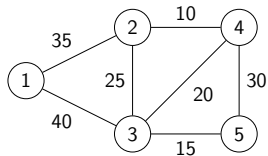


(c) Iteration 1

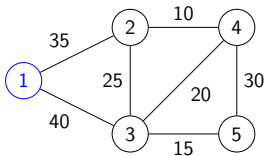


(d) Iteration 2

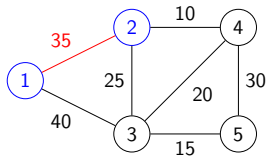
Prim's algorithm: example



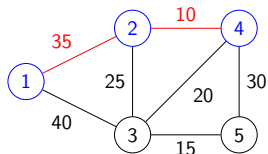
(a) Graph G



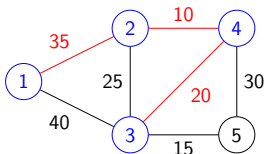
(b) Initialization



(c) Iteration 1

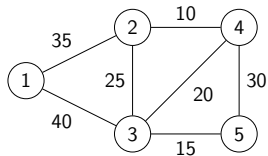


(d) Iteration 2

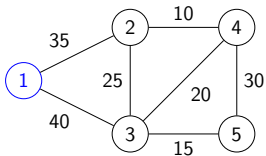


(e) Iteration 3

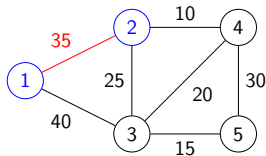
Prim's algorithm: example



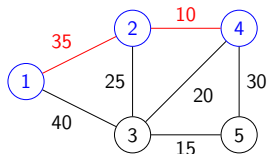
(a) Graph G



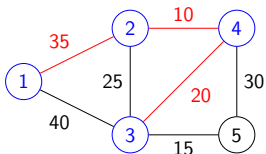
(b) Initialization



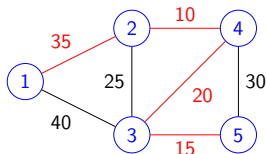
(c) Iteration 1



(d) Iteration 2



(e) Iteration 3



(f) Iteration 4

Algorithm Sollin's algorithm

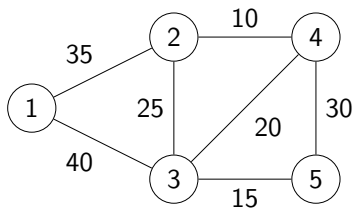
```
1:  $N_i \leftarrow \{i\}$  for  $i \in N$ 
2:  $T \leftarrow \emptyset$ 
3: while  $|T| < n - 1$  do
4:   for each  $N_k$  do
5:      $(i_k, j_k) \in \arg \min_{(i,j) \in (N_k, \bar{N}_k)} C_{ij}$ 
6:   end for
7:   for each  $N_k$  do
8:     if  $i_k$  and  $j_k$  belong to different trees then
9:       Merge two trees and  $T \leftarrow T \cup \{(i_k, j_k)\}$ 
10:    end if
11:  end for
12: end while
```

- Maintain multiple trees and merge them until getting a spanning tree

Correctness of Sollin's algorithm

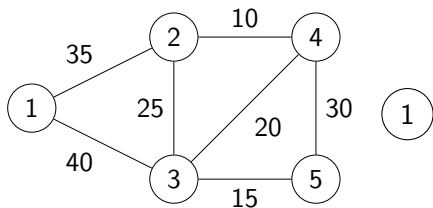
Sollin's algorithm finds a minimum spanning tree for undirected graphs.

Sollin's algorithm: example

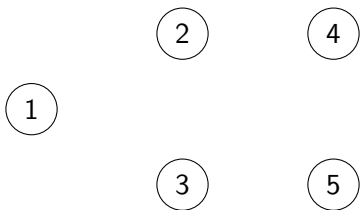


(a) Graph G

Sollin's algorithm: example

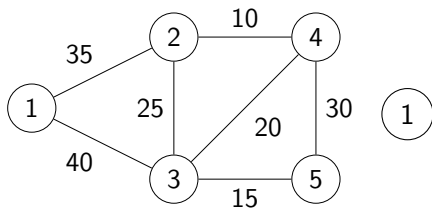


(a) Graph G

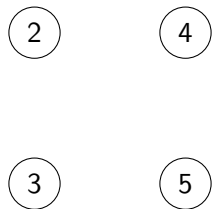


(b) Initialization

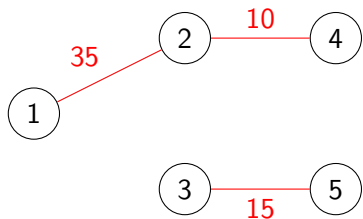
Sollin's algorithm: example



(a) Graph G

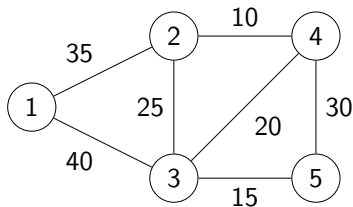


(b) Initialization

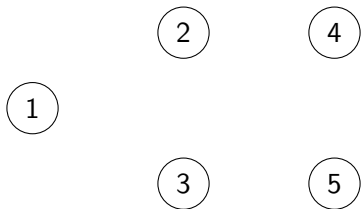


(c) Iteration 1

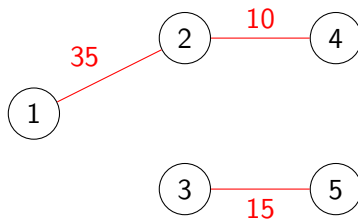
Sollin's algorithm: example



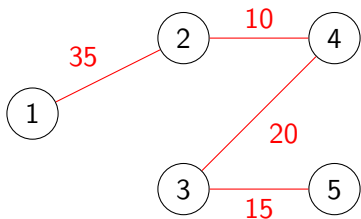
(a) Graph G



(b) Initialization



(c) Iteration 1



(d) Iteration 2

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- 3 Matroids and greedy algorithms**

Kruskal's algorithm is "greedy"

Algorithm Kruskal's algorithm

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1: Sort the arcs by their costs  $(e_1, e_2, \dots, e_m)$ 
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6:   end if
7:    $k \leftarrow k + 1$ 
8: end while
```

- Pick an arc with minimum cost at each iteration
- Check if picked arcs constitute a cycle

When does greedy algorithm work?

Matroid

A matroid is an ordered pair (E, \mathcal{I}) where E is a finite set and \mathcal{I} is a collection of subsets of E that satisfies

- 1 $\emptyset \in \mathcal{I}$;
- 2 for any $I_1 \subset I_2$, if $I_2 \in \mathcal{I}$ then $I_1 \in \mathcal{I}$;
- 3 if $I_1 \in \mathcal{I}$ and $I_2 \in \mathcal{I}$ and $|I_1| < |I_2|$, then there exists $e \in I_2 \setminus I_1$ such that $I_1 \cup e \in \mathcal{I}$.

- The sets in \mathcal{I} are called independent sets
- A maximal independent set is an independent set with max cardinality

Examples of matroids:

- Matric matroid: let M be a real-valued matrix
 - 1 E is the set of columns of M
 - 2 \mathcal{I} is the collection of subsets of linearly independent columns
- Graphic matroid: let G be an undirected graph
 - 1 E is the set of arcs
 - 2 \mathcal{I} is the collection of subsets of arcs that contain no cycle (forest)

Greedy works for matroid optimization

An optimization problem over matroids

Let (E, \mathcal{I}) be a matroid and $w : E \rightarrow \mathbb{R}$ be a function that assigns a cost to each element of E . Define $w(X) = \sum_{x \in X} w(x)$ for any nonempty subset $X \subset E$.

Find a maximal independent set that has the minimum cost.

Algorithm Greedy algorithm

- 1: Sort the elements of $E = \{e_1, \dots, e_n\}$ so that $w(e_1) \leq \dots \leq w(e_n)$
 - 2: $L \leftarrow \emptyset$,
 - 3: **for** $i = 1 : N$ **do**
 - 4: **if** $L \cup \{e_i\}$ is independent **then**
 - 5: $L \leftarrow L \cup \{e_i\}$
 - 6: **end if**
 - 7: **end for**
-

Greedy works

If (E, \mathcal{I}) is a matroid, then greedy algorithm finds a maximal independent set that has the minimum cost.

Greedy works for matroid optimization

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If (E, \mathcal{I}) is a matroid, then greedy algorithm finds a maximal independent set that has the minimum cost.

Greedy and only greedy works

Let \mathcal{I} be a collection of subsets of a set E . Then (E, \mathcal{I}) is a matroid if and only if \mathcal{I} has the following properties

- 1 $\emptyset \in \mathcal{I}$;
- 2 If $I \in \mathcal{I}$ and $I' \subset I$, then $I' \in \mathcal{I}$;
- 3 For all weight functions $w : E \rightarrow \mathbb{R}$, the greedy algorithm finds a maximal independent set that has the minimum cost.

A unit-time task scheduling problem

Problem statement

Given

- 1 a set $S = \{a_1, \dots, a_n\}$ of n unit-time tasks;
- 2 a set of n integer deadlines $d(a_1), \dots, d(a_n)$;
- 3 a set of nonnegative penalties $w(a_1), \dots, w(a_n)$;

Find a schedule for S that minimizes total penalties of missed deadlines.

Given a schedule, a task is *early* if finished before ddl and *late* otherwise

Canonical form a task schedule

An arbitrary task schedule can be transformed into the canonical form where early tasks precede late tasks.

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Search for a task schedule reduces to search for a set of early tasks!

A unit-time task scheduling problem

Independent set

A set of tasks is independent if there exists a schedule such that these tasks are no late.

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Independence can be checked efficiently

A set of tasks A is independent if and only if $N_t(A) \leq t$ where $N_t(A)$ is the number of tasks whose deadline is t or earlier for $t = 0, 1, \dots, n$.

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Matroid of task scheduling

If S is a set of unit-time tasks with deadlines, and \mathcal{I} is the set of all independent sets of tasks, then (S, \mathcal{I}) is a matroid.

A unit-time task scheduling problem

Minimizing penalty of late tasks = maximizing penalty of early ones!

Algorithm Greedy algorithm

- 1: Sort the tasks $S = \{a_1, \dots, a_n\}$ so that $w(a_1) \geq \dots \geq w(a_n)$
 - 2: $T \leftarrow \emptyset$,
 - 3: **for** $i = 1 : N$ **do**
 - 4: **if** $T \cup \{a_i\}$ is independent **then**
 - 5: $T \leftarrow T \cup \{a_i\}$
 - 6: **end if**
 - 7: **end for**
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Upcoming

Week 1-8 (AU4606 & AI4702):

- Introduction (1 lecture)
- Preparations (3 lectures)
 - basics of graph theory
 - algorithm complexity and data structure
 - graph search algorithm
- Shortest path problems (3 lectures)
- Maximum flow problems (5 lectures)
- Minimum cost flow problems (3 lectures)
- Introduction to multi-agent systems (1 lecture)
- Introduction to cloud networks (1 lecture)

Week 9-16 (AU4606):

- Simplex and network simplex methods (1 lecture)
- Global minimum cut problems (1.5 lectures)
- **Minimum spanning tree problems (1.5 lectures)**
- Submodular function optimization (2 lectures)
- Optimal assignments and matching (2 lectures)