# Modeling, Analysis and Optimization of Networks (Part 2: design) Lecture 3

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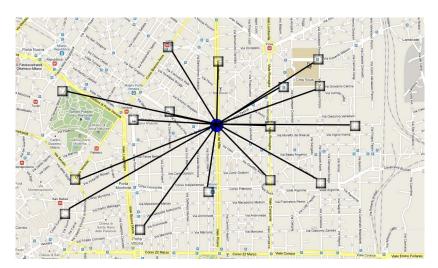


























## The problem

#### Given

- a set of terminal nodes,
- a set of bridge nodes,
- a set of potential links connecting them,

#### I want to

- decide how to link nodes,
- in such a way that transmissions can be performed between each pair of terminal nodes,
- minimizing the network cost.



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- Find a **tree** in G of minimum total cost
- ... containg **all** terminals  $(i \in T)$  and **any subset** of the bridges  $(i \in B)$ .

It is called the Steiner Tree Problem (STP) (Gauss, 1777-1855).

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#### Some considerations:

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# Solving the STP

#### Some considerations:

- Is it like a Minimum Spanning Tree?
- ...(it's not): MST is polynomially solvable, STP is NP-Hard.
- We'll see how to approximate it.





## **Approximation**

### What does it mean approximation?

Exact algorithms	a-priori guarantee of global optimality
Heuristics	no quality guarantee
Upper and lower bounds	a-posteriori quality guarantee
Approximation algorithms	a-priori quality guarantee

An  $\alpha$ -approx algorithm **always** gives a solution of cost **at most**  $\alpha$  times worse than the optimum.





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- Good news: easy to compute.
- Bad news: such a tree might not be optimal (on the whiteboard, AA page 28).
- Question: what if an element of T exists having no neighbors in T?





## The metric STP

The **metric** STP is a STP whose edge costs satisfy the **triangle inequality**: given three vertices  $i, j, k \in V$ 

$$c_{ij} \leq c_{ik} + c_{kj}$$

**Theorem:** there is an approximation factor preserving reduction from the STP to the metric STP (proof on the whiteboard, AA page 27).



# Approximating the metric STP

Metric STPs have a better structure:

**Theorem:** (for the metric STP), the cost of an MST on *T* is within 2-OPT (proof on the whiteboard, AA page 28).





# Approximating the STP

A 2-approx algorithm for the STP is the following:

- given a STP instance on a graph G, build an (equivalent) instance of the metric STP on a graph G'
- find a MST on terminals in graph G'
- map edges of G' in this MST to edges in G





## Steiner forests

Let us generalize the STP as follows: given

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- A set of **connection requests** between terminals: for each pair of terminals  $s, t \in T$ , coefficients  $r_{st} = 1$  if s and t must be connected,  $r_{st} = 0$  otherwise.





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- A set of **connection requests** between terminals: for each pair of terminals  $s, t \in T$ , coefficients  $r_{st} = 1$  if s and t must be connected,  $r_{st} = 0$  otherwise.
- Find a forest in G of minimum total cost
- ... containing at least one path connecting each pair of terminals s and t having r<sub>st</sub> = 1.



# Mathematical programming models

Let us consider a *cut function*: for each  $S \subseteq V$ 

$$f(S) = egin{cases} 1 & ext{if } S ext{ contains } s ext{ and } V \setminus S ext{ contains } t ext{ such that } r_{st} = 1 \ 0 & ext{otherwise} \end{cases}$$

Let us consider *crossing sets*: for each  $S \subseteq V$ 

$$\delta(S) = \text{set of edges crossing the cut } (S, V \setminus S)$$





# Mathematical programming models

#### Primal:

$$\begin{array}{ll} \text{minimize } \sum_{(i,j) \in E} c_{ij} x_{ij} \\ \\ \text{subject to } \sum_{(i,j) \in \delta(S)} x_{ij} \geq f(S) \\ \\ x_{ij} \in \{0,1\} \end{array} \qquad \forall S \subseteq V \\ \\ \forall (i,j) \in E \end{array}$$





# Mathematical programming models

#### Primal:

minimize 
$$\sum_{(i,j)\in E} c_{ij}x_{ij}$$
  
subject to  $\sum_{(i,j)\in \delta(S)} x_{ij} \geq f(S)$   $\forall S\subseteq V$   
 $x_{ij}\geq 0$   $\forall (i,j)\in E$ 

#### Dual:

maximize 
$$\sum_{S\subseteq V} f(S)y_S$$
 subject to  $\sum_{S:(i,j)\in\delta(S)} y_S \leq c_{ij}$   $\forall (i,j)\in E$   $y_S\geq 0$   $\forall S\subseteq V$ 

# Figurative terminology

- set S has been raised if  $y_S > 0$  (remark: it is never convenient to raise sets S having f(S) = 0)
- edge (i,j) feels dual  $y_S$  if  $(i,j) \in \delta(S)$  and  $y_S > 0$
- edge (i,j) is tight (resp overtight) if the sum of duals it feels equals (resp exceeds) its cost





# **Optimality** conditions

Primal (slackness) conditions: for each  $(i,j) \in E$ ,

$$x_{ij} \neq 0 \rightarrow \sum_{S:(i,j) \in \delta(S)} y_S = c_{ij}$$

Dual (slackness) conditions: for each  $S \subseteq V$ ,

$$y_S \neq 0 \rightarrow \sum_{(i,j) \in \delta(S)} x_{ij} = 1$$

Dual (relaxed slackness) conditions: for each  $S \subseteq V$ ,

$$y_S \neq 0 \rightarrow \sum_{(i,j) \in \delta(S)} x_{ij} \leq 2 \cdot f(S)$$
 "on the average"

(every raised cut has degree at most 2).



# Primal-dual algorithm

#### Idea:

- start with a super-optimal (infeasible) primal and a sub-optimal (feasible) dual
- iteratively improve the feasibility of the primal and the optimality of the dual, until a feasible primal is obtained
- x<sub>ii</sub> vars indicate which cuts need to be raised
- y<sub>S</sub> vars indicate which edges need to be picked

Invariant: the set of vars  $x_{ij}$  always identifies a *forest*.





# Primal-dual algorithm

A key step: **unsatisfied** and **active** sets. Given a primal solution, a set S is **unsatisfied** if

- has f(S) = 1
- there is no picked edge crossing the cut  $(S, V \setminus S)$ ;
- a set S is active if
  - it is unsatisfied
  - it does not contain unsatisfied sets (i.e. it is minimal wrt inclusion)

**Lemma:** A set S is active iff it is a connected component in the currently picked forest (and f(S) = 1). (proof on the whiteboard, AA page 200).





# Primal-dual algorithm for SFP

Primal-dual algorithm for SFP:

- (init)  $x_{ii} := 0$ ;  $y_S := 0$
- (augmentation) while there exists an unsatisfied set S do find active sets (by listing connected components) simultaneously raise  $y_S$  for each active set S until some edge (ij) becomes tight set  $x_{ii} := 1$  for each tight edge
- (pruning) for each (i,j) such that  $x_{ij} = 1$ , set  $x_{ij} := 0$  if the primal solution remains feasible

(example on the dashboard, AA pages 202-204).



## **Analysis**

**Theorem:** Primal-dual algorithm for the SFP achieves an approximation guarantee of 2. (proof on the whiteboard, AA pages 204-206)



# Tightness of the analysis

Are the analyses tight?

• try to find a STP (or SFP) instance in which our algorithms reach the worst case guarantee ...

example: page 30 AA.



## Further remarks

Some final observations. Both STP and MST are special cases of SFP:

- when run on a STP instance, the primal-dual algorithm builds a Spanning Tree on set T
- → the MST algorithm for STP is a special case of the primal-dual;
- when run on a MST instance (i.e. T=V), the primal-dual algorithm is essentially Kruskal's algorithm.



