Module 3: Duality through examples (Shortest Path Algorithm)

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min 
$$\sum (x_e : e \in E)$$
  
s.t.  $\sum (x_e : e \in \delta(S)) \ge 1$   
 $(\delta(S) \ s, t\text{-cut})$   
 $x \ge 0$ 



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$$\max \sum (y_S : \delta(S) \ s, t\text{-cut})$$
  
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$$x_e = \begin{cases} 1 & e \text{ bold in figure} \\ 0 & \text{otherwise} \end{cases}$$

for all  $e \in E$  is feasible for shortest path LP.

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$$y_{\{s\}} = y_{\{s,b\}} = 1, \ y_{\{s,a,b,c\}} = 3,$$

and  $y_S = 0$  for all other s, t-cuts  $\delta(S)$  yields a feasible dual solution of value 5!

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If  $\bar{x}$  is feasible for shortest path LP, and  $\bar{y}$  is feasible for its dual then  $b^T \bar{y} \leq c^T \bar{x}$ .

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- 2. How did we find the dual solution?



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Today:

- 1. How did we find the bold path?
- 2. How did we find the dual solution?
- 3. Is there always a shortest s, t-path and a dual solution whose value matches its length?

An Algorithm for the Shortest s, t-Path Problem

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A directed path is then a sequence of arcs:

$$\overrightarrow{v_1v_2}, \overrightarrow{v_2v_3}, \ldots, \overrightarrow{v_{k-1}v_k},$$

where  $\overrightarrow{v_i v_{i+1}}$  is an arc in the given graph, and  $v_i \neq v_j$  for all  $i \neq j$ .



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Example:

$$\overrightarrow{uv}, \overrightarrow{vw}, \overrightarrow{wx}$$

is a directed u, x-path.



Idea: Find an s, t-path P and a feasible dual y s.t.  $c(P) = \mathbb{1}^T y$ . How?



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#### Definition

Let y be a feasible dual solution. The slack of an edge  $e \in E$  is defined as

$$\label{eq:slack} \begin{split} \mathsf{slack}_y(e) &= c_e - \sum(y_U \, : \\ \delta(U) \; s, t\text{-cut, } e \in \delta(U)) \end{split}$$



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 $y \geq \mathbb{0}$ 

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Next: Look at all vertices that are reachable from *s* via directed paths:

$$U=\{s,c\}$$



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**Q**: By how much can you increase  $y_U$ ?





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Q: By how much can you increase  $y_U$ ? The maximum increase possible for  $y_{\{s,c\}}$  is determined by the slack of edges in  $\delta(\{s,c\})!$ 

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Edges *cd* and *sa* minimize slack. Pick one arbitrarily: *sa*.



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Also: change cd into  $\overrightarrow{cd}$ , and let

 $U = \{s, a, c, d\}$ 

be the reachable vertices from s



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Vertices reachable from s by directed paths:

 $U = \{s, a, c, d\}$ 

Let us compute the slack of edges in  $\delta(U)$ :

$$slack_y(ab) = 1$$
  

$$slack_y(cb) = 2 - 1 = 1$$
  

$$slack_y(ct) = 4 - 1 = 3$$
  

$$slack_y(dt) = 2$$

Let  $y_{\{s,a,c,d\}} = 1$ , add equality arc  $\overrightarrow{cb}$ , and update the set

$$U = \{s, a, b, c, d\}$$

of vertices reachable from  $\boldsymbol{s}$ 



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Note: we now have a directed s, t-path in our graph:

$$P = \overrightarrow{sc}, \overrightarrow{cd}, \overrightarrow{dt},$$

and its length is 4!



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and  $y_U = 0$  otherwise. Its value is 4!

 $\longrightarrow$  Path P is a shortest path!



# **Shortest Path Algorithm**

To compute the shortest Path for the instance on the right, we used the following algorithm:

Algorithm 3.2 Shortest path.

**Input:** Graph G = (V, E), costs  $c_e \ge 0$  for all  $e \in E$ ,  $s, t \in V$  where  $s \ne t$ .

Output: A shortest st-path P

- 1:  $y_W := 0$  for all *st*-cuts  $\delta(W)$ . Set  $U := \{s\}$
- 2: while  $t \notin U$  do
- 3: Let *ab* be an edge in  $\delta(U)$  of smallest slack for *y* where  $a \in U$ ,  $b \notin U$
- 4:  $y_U := \operatorname{slack}_y(ab)$
- 5:  $U := U \cup \{b\}$
- 6: change edge ab into an arc  $\overrightarrow{ab}$
- 7: end while
- 8: return A directed st-path P.



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- Have a look at the book. It has another full example run of the shortest path algorithm