

PRIMAL DUAL ALGORITHMSCombinatorial Algorithms

where LP comes only in the proofs

PRIMAL

$$\min \quad 2x_1 + 3x_2$$

$$4x_1 + 8x_2 \geq 12$$

$$2x_1 + x_2 \geq 3$$

$$3x_1 + 4x_2 \geq 4$$

$$x_1, x_2 \geq 0$$

y_1 y_2 y_3

$$y_1(4x_1 + 8x_2) + y_2(2x_1 + x_2) + y_3(3x_1 + 4x_2) \geq 12y_1 + 3y_2 + 4y_3$$

(holds for any feasible (x_1, x_2) for LP & $y_1, y_2, y_3 \geq 0$)

by rearranging variables, this is equal to

$$x_1(4y_1 + 2y_2 + 3y_3) + x_2(8y_1 + y_2 + 4y_3) \geq 12y_1 + 3y_2 + 4y_3$$

Suppose I enforce that

$$2 \geq 4y_1 + 2y_2 + 3y_3 \quad \& \quad y_1, y_2, y_3 \geq 0$$

$$3 \geq 8y_1 + y_2 + 4y_3$$

Then we get

$$2x_1 + 3x_2 \geq 12y_1 + 3y_2 + 4y_3$$

Create the following dual LP

Maximize	$12y_1 + 3y_2 + 4y_3$	G
s.t.	$4y_1 + 2y_2 + 3y_3 \leq 2$	DUAL
	$8y_1 + y_2 + 4y_3 \leq 3$	
	$y_1, y_2, y_3 \geq 0$	

If y_1, y_2, y_3 is any feasible soln to dual LP
of value say V, then

$$\text{OPT (Primal LP)} \geq V$$

Weak Duality Theorem

TheoremIf \bar{x} is any feasible soln to primal & \bar{y} is any feasible soln to dual LP,
 $\text{Opt}(\bar{x}) \geq \text{Opt}(\bar{y})$

If any feasible soln to dual LP,
then $\text{Obj}_P(\bar{x}) \geq \text{Obj}_D(\bar{y})$

Proof:

$$\begin{aligned}
 \text{Obj}_P(\bar{x}) &= 2\bar{x}_1 + 3\bar{x}_2 \\
 &\geq (4\bar{y}_1 + 2\bar{y}_2 + 3\bar{y}_3)\bar{x}_1 + (8\bar{y}_1 + \bar{y}_2 + 2\bar{y}_3)\bar{x}_2 \quad (\bar{y} \text{ is dual feasible}) \\
 &= \bar{y}_1(4\bar{x}_1 + 8\bar{x}_2) + \bar{y}_2(2\bar{x}_1 + \bar{x}_2) \\
 &\quad + \bar{y}_3(3\bar{x}_1 + 2\bar{x}_2) \leftarrow \\
 &\geq 12\bar{y}_1 + 3\bar{y}_2 + 4\bar{y}_3 \quad (\bar{x} \text{ is primal feasible}) \\
 &\geq \text{Obj}_D(\bar{y})
 \end{aligned}$$

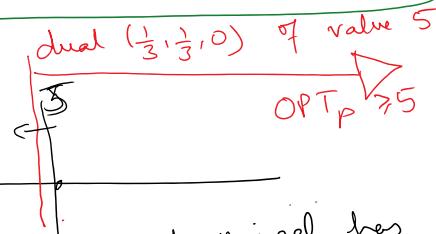
Consider $\bar{y}_1 = \frac{1}{3}$ & $\bar{y}_2 = \frac{1}{3}$, $\bar{y}_3 = 0$

This is feasible for dual LP

Summary

For above problem, to show that $\text{OPT}_P \leq 5$,
→ enough to produce 1 feasible soln to primal
(in the example $(1, 1)$)

→ to show that $\text{OPT}_P \geq 5$,
enough to produce 1 feasible soln to dual
(in this example $(\frac{1}{3}, \frac{1}{3}, 0)$)



$(1, 1)$ to primal has cost 5
 $\Rightarrow \text{OPT}_P \leq 5$

Theorem 2: If \bar{x}^* \bar{y}^* are feasible to primal & dual
respectively &

$\text{OBJ}_P(x^*) = \text{OBJ}_D(y^*)$ then
 x^* is optimal soln for primal
& y^* is optimal soln for dual.

$$P = \min_{\underline{x} \geq 0} c^T x \quad | \quad D = \max_{y \geq 0} b^T y$$

$$Ax \geq b \quad | \quad A^T y \leq c$$

If \bar{x} is primal feasible & \bar{y} is dual feasible then
 $c^T \bar{x} \geq b^T \bar{y}$ $\leftarrow (\bar{x}$ is primal feasible)
 $c^T \bar{x} \geq y^T A \bar{x} \geq y^T b = b^T y$.
 \uparrow (\bar{y} is dual feasible)

More generally

$$P = \min_{\begin{cases} x_i \geq 0 \\ x_i \in \mathbb{R} \end{cases}} c^T x$$

$$\rightarrow \begin{cases} a_i^T x \geq b_i & i \in I_1 \\ a_i^T x = b_i & i \in I_2 \\ x_j \geq 0 & j \in J_1 \\ x_j \in \mathbb{R} & j \in J_2 \end{cases}$$

$$D = \max_{\begin{cases} y_i \geq 0 \\ y_i \in \mathbb{R} \end{cases}} b^T y$$

$$\begin{cases} y_i \geq 0 & \forall i \in I_1 \\ y_i \in \mathbb{R} & \forall i \in I_2 \\ A^T y \geq c_j & \forall j \in J_1 \\ A^T y = c_j & \forall j \in J_2 \end{cases}$$

Nice Fact

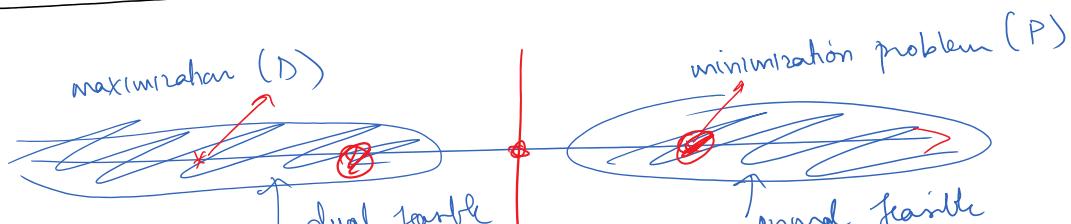
$$\text{Dual (Dual)} = \text{Primal}$$

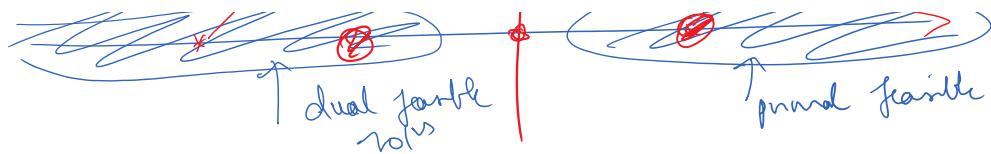
$P = \min ; D = \max$

Every constraint (P) \iff dual variable
 \rightarrow ($> b$ constraint) \iff (> 0 variable)
 \quad ($= b$ constraint) \iff ($\in \mathbb{R}$ variable)

Every variable (P) \iff dual constraint
 \rightarrow (≥ 0 variable) \iff ($\leq c$ constraint)
 \quad ($\in \mathbb{R}$ variable) \iff ($= c$ constraint)

(primal, dual) pair





Strong duality Theorem

For any primal dual pair, the following situations can arise

- A) Primal is infeasible, dual is infeasible
- B) Primal is unbounded, dual is infeasible
- C) Dual is unbounded, primal is infeasible
- D) both are feasible & $\text{OPT}(P) = \text{OPT}(D) = V^*$ }
for some V^*

Example:

$$\begin{array}{c}
 \text{Min} \quad 2x_1 + 3x_2 \\
 x_1 \leq 2 \\
 x_2 \leq 3
 \end{array}
 \quad =
 \quad
 \begin{array}{l}
 \text{Min} \quad 2x_1 + 3x_2 \\
 x_1 \geq -2 \quad (y_1) \\
 -x_2 \geq -3 \quad (y_2)
 \end{array}$$

↓

Dual

Max-2y₁ - 3y₂-y₁ = 2-y₂ = 3y₁, y₂ ≥ 0

$$\begin{array}{c}
 \text{Min} \quad 0 \\
 y_1 \quad x_1 \geq 2 \\
 y_2 \quad -x_1 \geq -1 \\
 x_1 \in \mathbb{R}
 \end{array}
 \quad
 \begin{array}{l}
 \text{Max} \quad 2y_1 - y_2 \\
 y_1 - y_2 \leq 0 \\
 y_1 \geq 0 \\
 y_2 \geq 0
 \end{array}$$

dual unbounded
 primal infeasible

Corollary

- * If P/D is unbounded $\Rightarrow D/P$ is infeasible
- * if P or D has bounded optimum value, then both have equal optimum value

- - - slackness - - -

Complementary Slackness

let (x^*, y^*) be an optimal primal dual pair

$$\begin{array}{ll} \min & C^T x \\ \boxed{P} & Ax \geq b \\ & x \geq 0 \end{array} \quad \left| \quad \begin{array}{ll} \max & b^T y \\ \boxed{D} & A^T y \leq C \\ & y \geq 0 \end{array} \right.$$

STRONG DUALITY THM: $C^T x^* = b^T y^*$

if $x_j^* > 0$ for some variable, then corresponding dual constraint is tight

$$A_j^T y^* = c_j$$

if $y_i^* > 0$ for some dual variable, then corr. primal constraint is tight
(ie) $A_i^T x^* = b_i$

Proof:-

Dual $b^T y^* \leq x^T A^T y^*$ (uses primal feasibility of x^*)

① If $y_i^* > 0$ & $A_i^T x^* > b_i$ then we'll have

$$b^T y^* < x^T A^T y^*$$

Primal $C^T x^* \geq y^T A x^* \geq b^T y^*$ (if ① holds)

↑ contradicts strong duality because $C^T x^* = b^T y^*$.

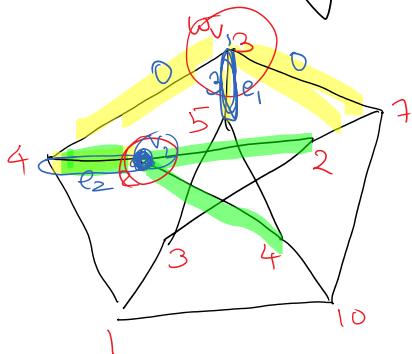
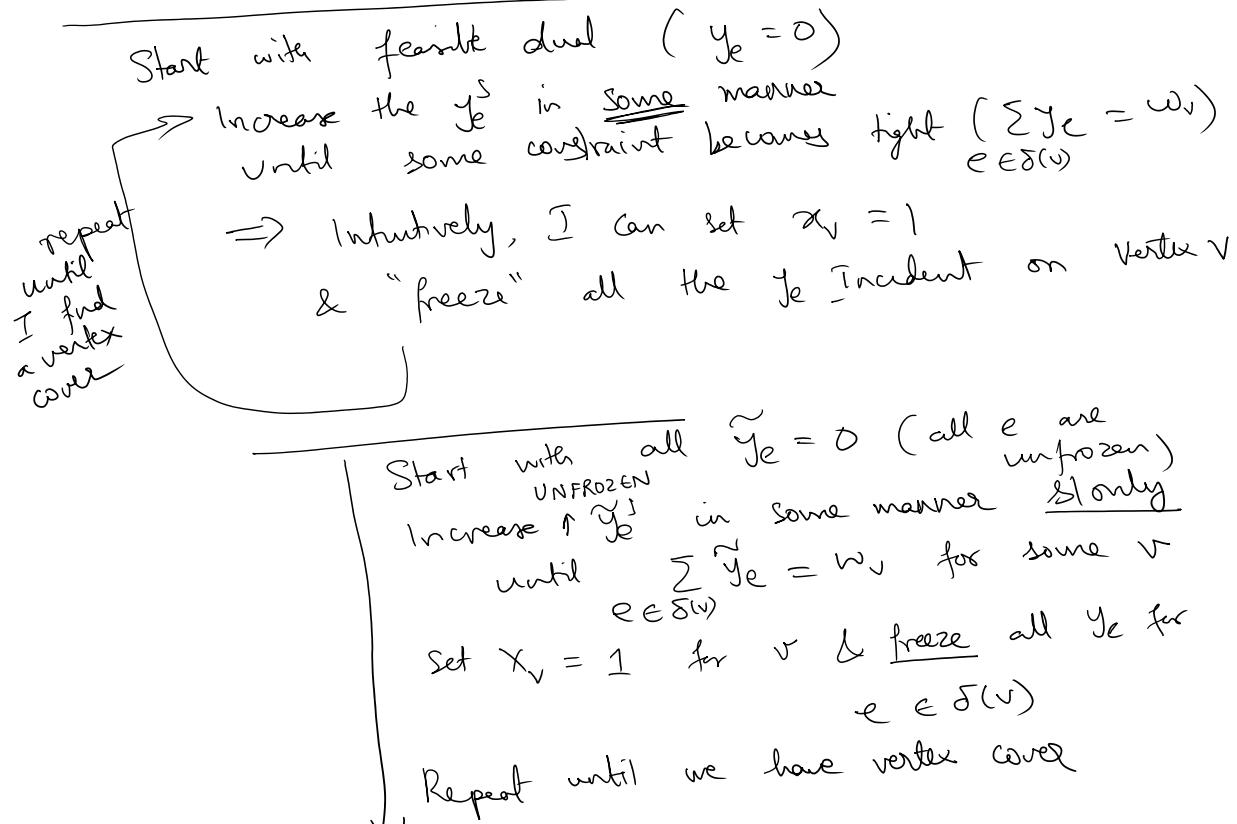
Often times we want to solve the minimization primal

$$\begin{array}{ll} \text{Min} & \sum w_v x_v \\ \boxed{P} & x_u + x_v \geq 1 \quad \forall (u,v) \in E \\ & x_u \geq 0 \quad \forall u \end{array} \quad \left| \quad \begin{array}{ll} \text{Max} & \sum y_e \\ \boxed{D} & \sum_{e \in S(v)} y_e \leq w_v \quad \forall v \\ & y_e \geq 0 \quad \forall e \end{array} \right.$$

WANT (goal)

Algo which computes integer soln to primal & "good" feasible soln to dual

- A) if primal variable $> 0 \Rightarrow$ dual constraint is tight
- B) if dual variable $> 0 \Rightarrow$ primal constraint is "approx" tight



set all $\bar{y}_e = 0 + e$
~~inter~~ increase $\bar{y}_{e_1} = 3$
 Set $x_{v_1} = 1$
 increase $\bar{y}_{e_2} = 2$
 & set $x_{v_2} = 1$

Invariant My \bar{y}_e → always dual feasible

Weak duality

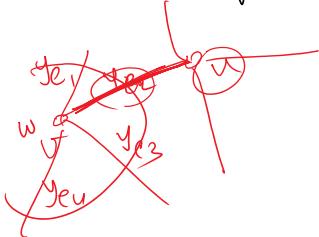
If x^* is any primal soln (feasible)

$$\text{OPT}(vc) \geq \sum_{v \in V} w_v x_v^* \geq \sum_{e \in E} \bar{y}_e$$

Our Cost

if $v \in \text{final soln}$,

$$w_v = \sum_{e \in E} \bar{y}_e$$



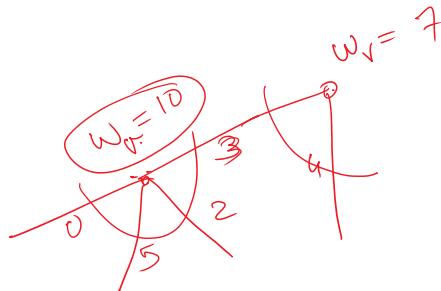
$$= \sum_{v \in S^*} \sum_{e \in E} \bar{y}_e$$

$$w_v = \sum_{e \in E} \bar{y}_e$$

$$2 \sum_{e \in E} \bar{y}_e \leq 2 \text{OPT}(vc)$$

∴ ... find a vertex cover at the end!

Q: Why did we find a vertex cover at the end?
 A: If not, some edge e would have been frozen,
 & algo would not have terminated.



$\text{Min } \sum_{v \in V} w_v x_v$ $x_u + x_v \geq 1 \quad \forall (u, v) \in E$ $x_u \geq 0$	$\text{Max } \sum_{e \in E} y_e$ $\sum_{e \in \delta(v)} y_e \leq w_v$ $y_e \geq 0$
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P D

<u>Weak Duality Theorem</u>	$\overrightarrow{y^*}$ $\overrightarrow{x^*}$ $\sum_e y_e^* \leq \sum_u w_u x_u^*$
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$\text{Obj}_D(y^*) \leq \text{Obj}_P(x^*)$

(i.e.) $\sum_e y_e^* \leq \sum_u w_u x_u^*$

all feasible
primal soln
 x^*

In other words, if we find a good dual feasible solution with large $\sum_e y_e^*$, then that gives us a lower bound on $\text{OPT(VC)} \geq \sum_e y_e^*$

goal in primal dual approx algos
 → a) try to find a good dual soln &
 b) using it construct a good primal soln.

constraint: $\sum_e y_e = 0 \quad \forall e, S = \emptyset$

Start with all $\tilde{y}_e = 0 \forall e, S = \emptyset$

Set all edges as unfrozen

\rightarrow Until S is a valid vertex cover

we pick arbitrary unfrozen edge ' e '
 & increase \tilde{y}_e till some endpoint
 becomes tight (i.e. $\sum_{e \in \delta(u)} \tilde{y}_e = w_u$)

Freeze ' e ' & add u to S

Fact 1 : $\text{OPT}(\text{VC}) \geq \sum_{e \in E} \tilde{y}_e$ (weak Duality)

Fact 2 :

$$\begin{aligned} \text{Alg Cost} &= \sum_{u \in S} w_u = \sum_{u \in S} \sum_{e \in \delta(u)} \tilde{y}_e \\ &\leq 2 \sum_{e \in E} \tilde{y}_e \leq 2 \text{OPT}(\text{VC}) \end{aligned}$$

Fact 3

Why is S feasible?

- Firstly we run the algo till S is feasible

If all edges are frozen, then S is a VC

- if some edge is unfrozen, algo can continue.

PRIMAL DUAL ALGORITHMS FOR GRAPH CONNECTIVITY PROBLEMS

Steiner Tree Problem

Given |

Given a graph $G = (V, E)$ & there are ≥ 0 costs on edges $c_e \geq 0 \forall e \in E$

& there are collection of terminals $t_1, t_2, \dots, t_k \in V$ and one root $r \in V$

Goal | Find a minimum cost subgraph $F \subseteq E$ st for all t_i , there exists a $r - t_i$ path in F .



Steiner Forest Problem

Same input

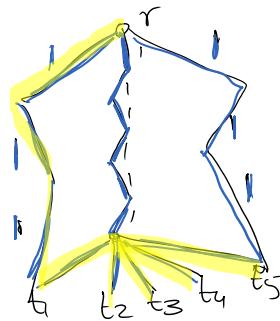
$$G = (V, E)$$

ϵ cost > 0

(S_1, t_1)
 (S_2, t_2)
 \vdots
 (S_k, t_k)

} terminal pairs

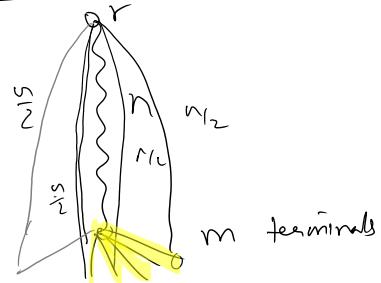
Fund min. cost $P \subseteq E$ st each S_i has a path to t_i



Thm:
 There is a 2-approximation using the PD framework

Union of Shortest Paths: $m \cdot \frac{n}{2}$ cost

$$\text{Opt cost} = \frac{n}{2} + m$$



→ We know how to solve MST well
 (min. spanning tree)

→ For every pair of terminals (t_i, t_j) compute
 the shortest path between t_i & t_j and
~~create~~ add a fake edge of that
 length l_{ij} .

{ Do the same for (t_i, r) & t_i & add
 these edges also

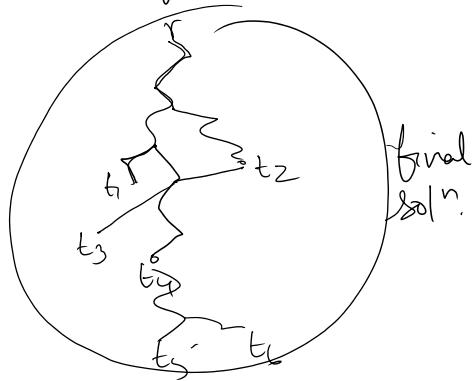
on a new graph on only $(k+1)$ vertices
 r, t_1, t_2, \dots, t_k .

↓ Compute MST on new graph.

↓ Compute MST on new graph.

↓
Map it back to original edges

Suppose M ST

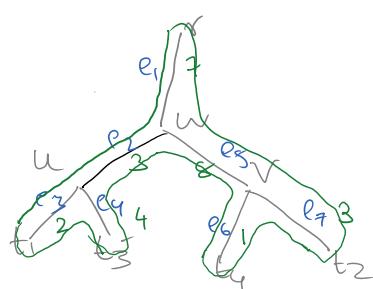


Lemma ① In fake graph G suppose MST has cost C' then we can find a subgraph $\subseteq C'$ in the real graph.

Need to show \exists Spanning tree in G' of cost $\leq 2 \cdot OPT$

Proof: Consider the optimal Steiner Tree in G has cost C^*

Suppose $\partial T =$
 $S_0 \quad C^* = 28$



Goal: Show some spanning tree in fake graph of cost ≤ 56

Proof

Consider Spanning path

$$r \leq t_1 \leq t_3 \leq t_4 \leq t_2 \leq r$$

51

2-approximation for Steiner Forest

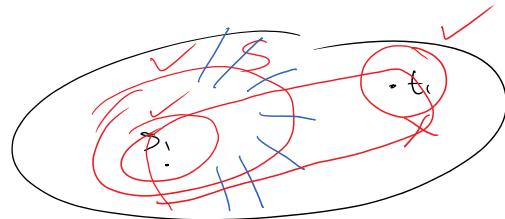
Primal Dual approach.

Special Case

$G = (V, E)$ $c_e \geq 0$ on edges & (s_i, t_i)

What's a nice LP for this problem?
 x_e be a variable for whether e is included or not.

$$\begin{aligned} \text{Min } & \sum c_e x_e \\ (y_s) \quad & \sum_{e \in \delta(S)} x_e \geq 1 \quad \forall S \subseteq V : |S \cap \{s_i, t_i\}| = 1 \\ & x_e \geq 0 \end{aligned}$$



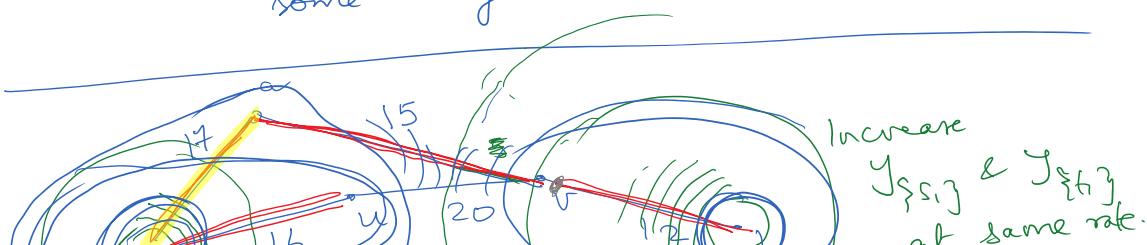
Lemma

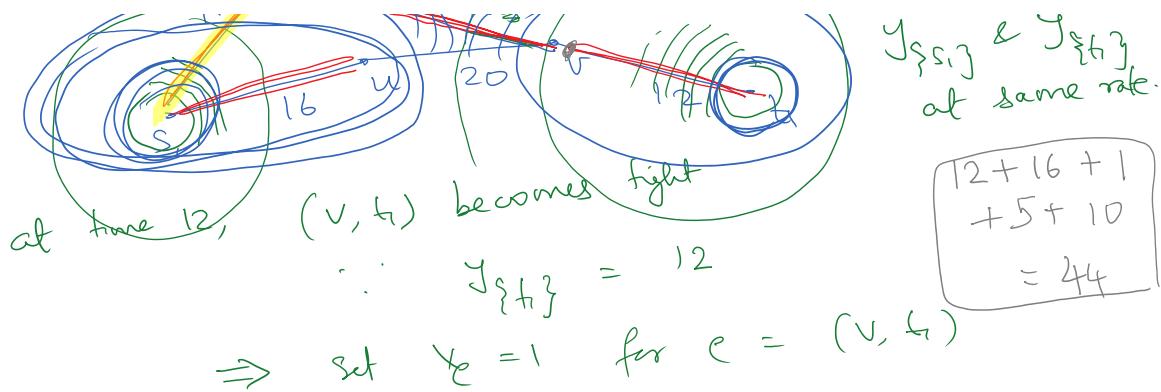
For any integral $\{x_e\}$ feasible for the above LP,
 s_i & t_i are connected.

$$\begin{aligned} \text{Min } & \sum c_e x_e \\ y_s \quad & \sum_{e \in \delta(S)} x_e \geq 1 \quad \forall S \subseteq V : |S \cap \{s_i, t_i\}| = 1 \\ & S = S : |\{S \cap \{s_i, t_i\}\}| = 1 \end{aligned} \quad \begin{aligned} \text{Max } & \sum_S y_s \\ & \sum_{S: e \in \delta(S)} y_s \leq c_e \\ & y_s \geq 0 \end{aligned}$$

Key Ideas

- 1) Start with $y_s = 0 \ \forall S$, $x_e = 0 \ \forall e$
- 2) Increase y_s for all minimally violated sets until some edge becomes tight,





at time $12 + \epsilon$
increase $\gamma_{\{s, t\}}$ & $\gamma_{\{v, t\}}$ at same rate.

at time 16, (s_1, u) edge becomes tight.
 \Rightarrow set $\gamma_e = 1$ for $e = (s_1, u)$

after that increase

$\gamma_{\{s_1, u\}}$ & $\gamma_{\{v, t\}}$ at same rate.

at time 17, (s_1, a) becomes tight
 \Rightarrow set $\gamma_e = 1$ for $e = (s_1, a)$

after that, increase

$\gamma_{\{s_1, u, a\}}$ & $\gamma_{\{v, t\}}$ at same rate.

at time 22, edge (a, v) becomes tight
 \Rightarrow set $\gamma_e = 1$ for $e = (a, v)$.
 S₁ & t are connected now! stop the dual growing process
 & delete redundant edges.

- ① Algo will find a feasible soln. \leftarrow
 Our dual values are feasible for dual problem
- ② $\Rightarrow \text{OPT}(\text{Problem}) \geq \sum \tilde{\gamma}_s$ (^{dual values we have maintained at least})

$$\begin{aligned} \gamma_{\{t\}} &= 12 \\ \gamma_{\{s, t\}} &= 16 \end{aligned}$$

$$\left. \begin{aligned} \text{Any feasible primal} &\geq 44 \\ \rightarrow \text{optimal primal} &\geq 44 \end{aligned} \right\}$$

$$\begin{aligned} y_{\{t_1\}} &= 1 \\ y_{\{s_1\}} &= 16 \\ y_{\{s_1, u\}} &= 1 \\ y_{\{v, t_1\}} &= 10 \\ y_{\{s_1, u, a\}} &= 5 \end{aligned}$$

Any feasible \bar{y} ...
 \Rightarrow Optimal primal ≥ 44
 \Rightarrow Optimal integer soln $\rightarrow 44$

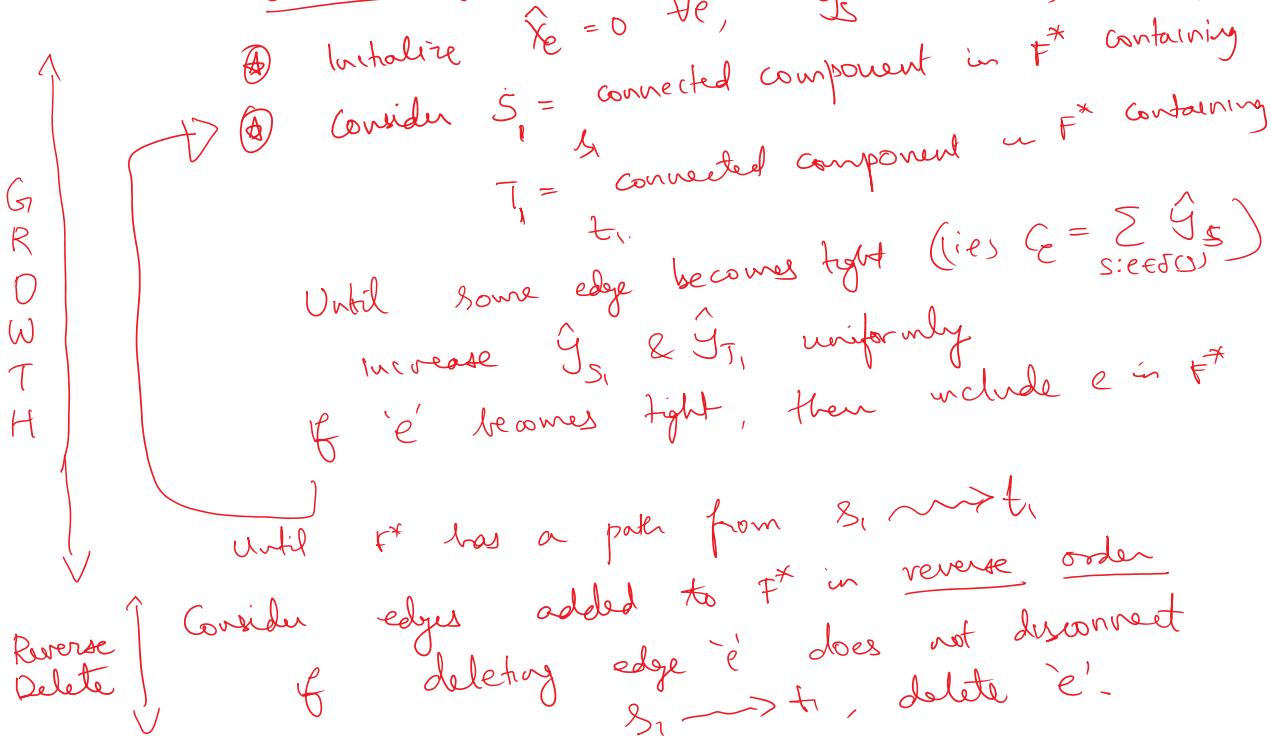
Cost of Alg :-
 If edge $e \in$ final soln, $c_e = \sum_{S: e \in \delta(S)} y_S$

let F^* be final soln after deletion

Final cost of algorithm

$$\begin{aligned} &= \sum_{e \in F^*} c_e = \sum_{e \in F^*} \sum_{S: e \in \delta(S)} y_S = \sum_S \sum_{e \in \delta(S) \cap F^*} y_S \\ &= \sum_S y_S \quad \text{If } F^* \cap \delta(S) \neq \emptyset \\ &= \sum_S y_S \end{aligned}$$

Overall algorithm



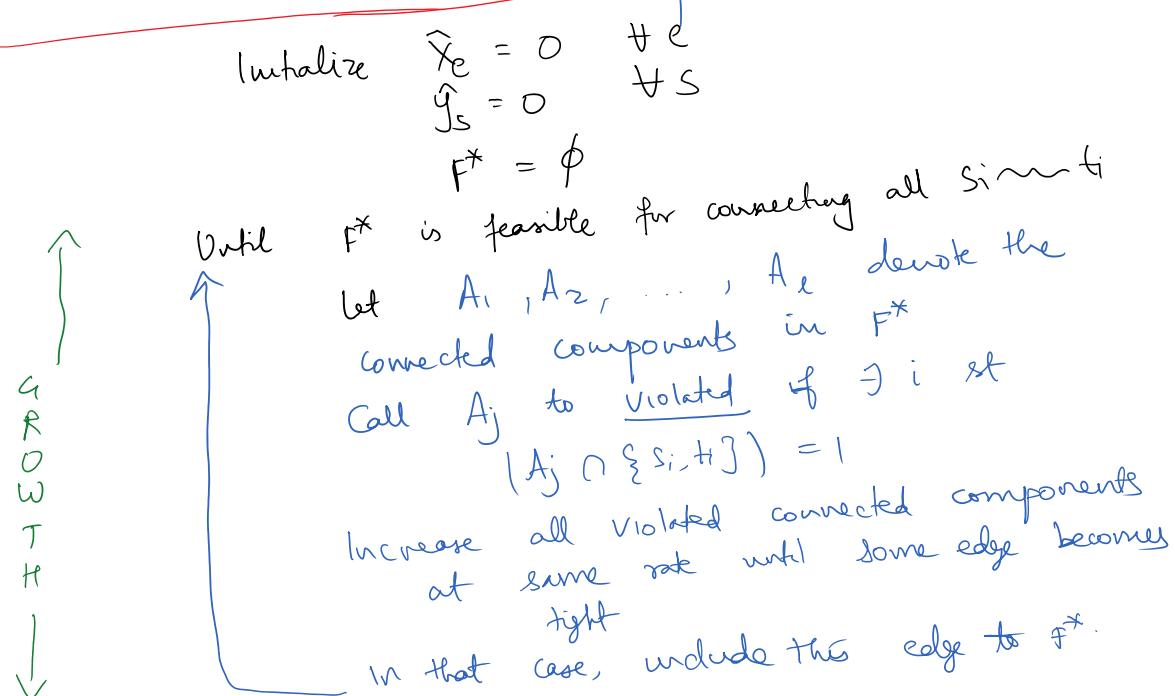
Lemma

after the reverse debt process, if $y_s > 0$
for any set $S \subseteq V$, then

$$|F^* \cap \bar{\delta}(S)| = 1$$

How to generalize to Steiner Forest?

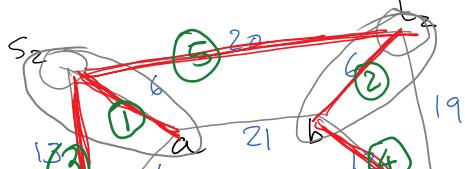
$$\begin{array}{l} \text{Min } \sum c_e x_e \\ \text{Max } \sum y_s \\ \text{S.t. } \sum_{e \in \delta(S)} x_e \geq 1 \quad \# S: J_i : |S \cap \{s_i, t_i\}| = 1 \\ \quad x_e \geq 0 \quad \sum_{e \in \delta(S)} y_s \leq \epsilon \\ \quad y_s \geq 0 \end{array}$$



Reverse
Delete

Consider edges added to F^* in reverse order.
if any edge e can be deleted w/o disconnecting some $s_i - t_i$ pair,
delete it.

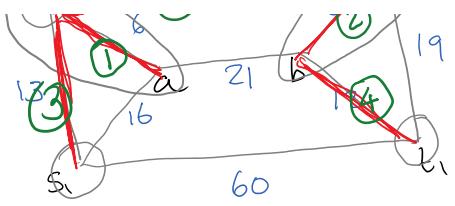
Output remaining F^* .



At time 6,

$$y_{\{s_2\}} = 6 = c_{(s_2, a)}$$

Include (s_2, a) in F^*



$\cup \{s_2\}$
Include (s_2, a) in F^*

& $y_{\{t_2\}} = 6 = (b, t_2)$
Include (b, t_2) in F^*

Violated Components
 $\{s_1\}, \{t_1\}, \{s_2, a\}, \{t_2, b\}$

At time 6.5,
 $y_{\{s_2\}} + y_{\{s_2, a\}} + y_{\{s_1\}} = 6 + 0.5 + 6.5 = 13$
 $= (s_2, s_1)$
 Include $(s_1, s_2) \in F^*$

Violated Components:
 $\{s_1, s_2, a\}, \{t_2, b\}, \{t_1\}$

At time 9
 (b, t_1) becomes tight

Violated Components
 $\{s_1, s_2, a\} \& \{b, t_1, t_2\}$

At time 10
 (s_2, t_2) becomes tight
 \Rightarrow include in F^*

Reverse Delete only deletes (s_2, a) .
 All other edges remain

② $\text{OPT}(\text{Steiner Forest}) \geq \sum_s \hat{y}_s$ (total dual we've raised)

① $\{\hat{y}_s\}$ is dual feasible

③ If I show $\sum_{e \in F^*} c_e \leq 2 \sum_s \hat{y}_s$ we're done.

④ $\forall e \in F^*, c_e = \sum_{s: e \in \delta(s)} \hat{y}_s$

\Rightarrow Remains to show

$$\sum_{S \in F^*} \sum_{s: e \in \delta(s)} \hat{y}_s \leq 2 \sum_s \hat{y}_s$$

equivalently

w.t.s

$$\boxed{\sum_s \hat{y}_s |F^* \cap \delta(s)| \leq 2 \sum_s \hat{y}_s}$$

what we'll show is that while all $|F^* \cap \delta(s)|$ may not be bounded by 2, it will be true on average.

lets do this over time (from $t=0$)

