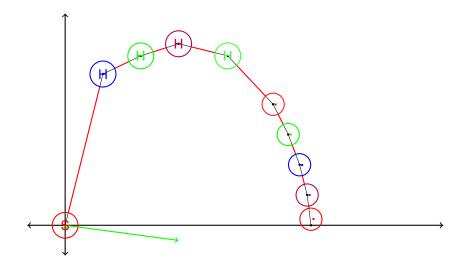
CS 170: Algorithms



Lecture in a Minute

Simplex Implementation: Start at a (feasible) vertex.

Lecture in a Minute

Simplex Implementation:

Start at a (feasible) vertex.

(defined by linear system $A'x = [b', 0, \dots, 0]$).

Begin at origin. Move to better neighboring vertex.

Coordinate system: distance from tight constraints.

Vertex at origin in coordinate system.

O(mn) time to update linear system.

Until no better neighboring vertex.

Objective function in coordinate system is non-positive.

Dual Variables: new objective function!

Lecture in a Minute

```
Simplex Implementation:
 Start at a (feasible) vertex.
   (defined by linear system A'x = [b', 0, \dots, 0]).
   Begin at origin. Move to better neighboring vertex.
    Coordinate system: distance from tight constraints.
    Vertex at origin in coordinate system.
    O(mn) time to update linear system.
 Until no better neighboring vertex.
 Objective function in coordinate system is non-positive.
  Dual Variables: new objective function!
Maximum flow.
  "Greedy" augment path...
     Except reverse old decisions ..
     Reverse residual capacities.
  (Friday): Optimality?
     No augmenting path \Longrightarrow
      s-t cut size = flow value.
    Find flow and s-t cut with equal value!
```

Two tasks:

Two tasks:

- 1. Check optimality of vertex?
- 2. Where to go next?

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Canonical LP.

$$\max c^T x$$
$$Ax \le b$$
$$x \ge 0$$

Start at origin, supposing it is feasible.

Two tasks:

- 1. Check optimality of vertex?
- 2. Where to go next?

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$$\max c^T x$$
$$Ax \le b$$
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Start at origin, supposing it is feasible.

Vertex since intersection of *n* constraints of form $x_i = 0$.

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Optimal?

If all $c_i \leq 0$

Two tasks:

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Optimal?

If all $c_i \le 0 \implies$ increasing any x_i decreases value

Two tasks:

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$$Ax \le b$$
$$x \ge 0$$

Start at origin, supposing it is feasible.

Vertex since intersection of *n* constraints of form $x_i = 0$.

Optimal?

If all $c_i \le 0 \implies$ increasing any x_i decreases value \implies optimal!

Two tasks:

- Check optimality of vertex?
- 2. Where to go next?

Canonical LP.

$$\max c^T x$$
$$Ax \le b$$
$$x \ge 0$$

Start at origin, supposing it is feasible.

Vertex since intersection of *n* constraints of form $x_i = 0$.

Optimal?

If all $c_i \le 0 \implies$ increasing any x_i decreases value \implies optimal! if there is $c_i > 0$

Two tasks:

- Check optimality of vertex?
- 2. Where to go next?

Canonical LP.

$$\max c^T x$$
$$Ax \le b$$
$$x \ge 0$$

Start at origin, supposing it is feasible.

Vertex since intersection of *n* constraints of form $x_i = 0$.

Optimal?

If all $c_i \le 0 \implies$ increasing any x_i decreases value \implies optimal! if there is $c_i > 0$ increasing x_i increases value

Two tasks:

- 1. Check optimality of vertex?
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Canonical LP.

$$\max c^T x$$
$$Ax \le b$$
$$x \ge 0$$

Start at origin, supposing it is feasible.

Vertex since intersection of *n* constraints of form $x_i = 0$.

Optimal?

If all $c_i \le 0 \implies$ increasing any x_i decreases value \implies optimal! if there is $c_i > 0$ increasing x_i increases value \implies not optimal.

Two tasks:

- Check optimality of vertex?
- 2. Where to go next?

Canonical LP.

$$\max c^T x$$
$$Ax \le b$$
$$x \ge 0$$

Start at origin, supposing it is feasible.

Vertex since intersection of *n* constraints of form $x_i = 0$.

Optimal?

If all $c_i \le 0 \implies$ increasing any x_i decreases value \implies optimal! if there is $c_i > 0$ increasing x_i increases value \implies not optimal. Done with task 1.

Two tasks:

Two tasks:

- 1. Check optimality of vertex?
- 2. Where to go next?

Two tasks:

- 1. Check optimality of vertex?
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At origin, there is positive c_i , so increase x_i .

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...until you hit another constraint.

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 $x_i \ge 0$ is no longer tight, but new constraint is.

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 \implies *n* constraints!

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At origin, there is positive c_i , so increase x_i .

...until you hit another constraint.

 $x_i \ge 0$ is no longer tight, but new constraint is.

 \implies *n* constraints!

At vertex!

$$\max 2x_1 + 5x_2
2x_1 - x_2 \le 4
 x_1 + 2x_2 \le 9
 -x_1 + x_2 \le 3
 x_1 \ge 0
 x_2 \ge 0$$

$$3
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$$\max 2x_1 + 5x_2
2x_1 - x_2 \le 4
 x_1 + 2x_2 \le 9
 -x_1 + x_2 \le 3
 x_1 \ge 0
 x_2 \ge 0$$
(1)
(2)
(3)
(4)
(5)

Origin: feasible, value 0.

$$\max 2x_1 + 5x_2$$
 $2x_1 - x_2 \le 4$
 $x_1 + 2x_2 \le 9$
 $-x_1 + x_2 \le 3$
 $x_1 \ge 0$
 $x_2 \ge 0$
 (1)
 (2)
 (3)
 (3)
 (4)
 (5)

Origin: feasible, value 0. Inequalities 4 and 5 are tight.

$$\max 2x_1 + 5x_2$$

$$2x_1 - x_2 \le 4$$
$$x_1 + 2x_2 \le 9$$

$$-x_1+x_2\leq 3$$

$$x_1 \geq 0$$

$$x_2 \geq 0$$

1)

(3)

4

(5)

Origin: feasible, value 0. Inequalities (4) and (5) are tight.

Relax constraint $x_2 = 0$.

$$\max 2x_1 + 5x_2$$

$$2x_1 - x_2 \le 4$$

$$x_1+2x_2\leq 9$$

$$-x_1+x_2\leq 3$$

$$x_1 \ge 0$$
$$x_2 \ge 0$$

$$x_2 \geq 0$$

(5)

Origin: feasible, value 0. Inequalities (4) and (5) are tight.

Relax constraint $x_2 = 0$. Increase x2 until

$$\max 2x_1 + 5x_2$$

$$2x_1 - x_2 \le 4$$

$$x_1+2x_2\leq 9$$

$$-x_1 + x_2 \le 3$$

$$x_1 \geq 0$$

$$x_2 \geq 0$$



Origin: feasible, value 0. Inequalities (4) and (5) are tight.

Relax constraint $x_2 = 0$.

Increase x₂ until

...Inequality (3) becomes tight constraint.

$$\max 2x_1 + 5x_2$$

$$2x_1 - x_2 \le 4$$

$$x_1+2x_2\leq 9$$

$$-x_1+x_2\leq 3$$

$$x_1 \geq 0$$

$$x_2 \geq 0$$



Origin: feasible, value 0. Inequalities (4) and (5) are tight.

Relax constraint $x_2 = 0$.

Increase x2 until

...Inequality 3 becomes tight constraint.

...Tight constraints: (3) and (4).

$$\max 2x_1 + 5x_2$$

$$2x_1 - x_2 \le 4$$

$$x_1+2x_2\leq 9$$

$$-x_1+x_2\leq 3$$

$$x_1 \geq 0$$

$$x_2 \geq 0$$



Origin: feasible, value 0. Inequalities (4) and (5) are tight.

Relax constraint $x_2 = 0$.

Increase x_2 until

...Inequality (3) becomes tight constraint.

...Tight constraints: (3) and (4).

...new vertex: (0,3) with value 15.

$$\max 2x_1 + 5x_2$$

$$2x_1 - x_2 \le 4$$

$$x_1+2x_2\leq 9$$

$$-x_1 + x_2 \le 3$$

$$x_1 \geq 0$$

$$x_2 \geq 0$$



Origin: feasible, value 0. Inequalities (4) and (5) are tight.

Relax constraint $x_2 = 0$.

Increase x₂ until

...Inequality (3) becomes tight constraint.

...Tight constraints: (3) and (4).

...new vertex: (0,3) with value 15.

Easy process from origin:

$$\max 2x_1 + 5x_2$$

$$2x_1 - x_2 \le 4$$

$$x_1 + 2x_2 \le 9$$

$$-x_1+x_2\leq 3$$

$$x_1 \geq 0$$

$$x_2 \geq 0$$

Origin: feasible, value 0. Inequalities (4) and (5) are tight.

Relax constraint $x_2 = 0$.

Increase x_2 until

...Inequality (3) becomes tight constraint.

...Tight constraints: (3) and (4).

...new vertex: (0,3) with value 15.

Easy process from origin: just increase one variable with positive c.

$$\max 2x_1 + 5x_2$$

$$2x_1 - x_2 \le 4$$

$$x_1 + 2x_2 \le 9$$

$$-x_1+x_2\leq 3$$

$$x_1 \geq 0$$

$$x_2 \geq 0$$



Origin: feasible, value 0. Inequalities (4) and (5) are tight.

Relax constraint $x_2 = 0$.

Increase x2 until

...Inequality (3) becomes tight constraint.

...Tight constraints: (3) and (4).

...new vertex: (0,3) with value 15.

Easy process from origin: just increase one variable with positive *c*.

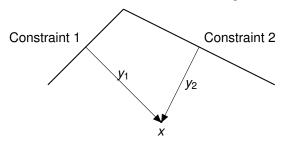
Now what?

A new coordinate system.

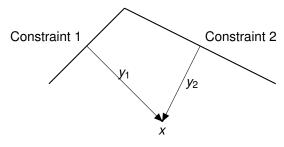
New coordinates: Distance from new tight constraints.

A new coordinate system.

New coordinates: Distance from new tight constraints.

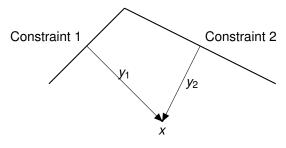


New coordinates: Distance from new tight constraints.



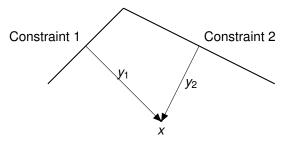
 y_i is distance from constraint i

New coordinates: Distance from new tight constraints.



 y_i is distance from constraint i x is at (y_1, y_2) in new coordinate system.

New coordinates: Distance from new tight constraints.

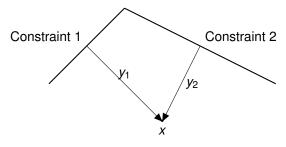


 y_i is distance from constraint i

x is at (y_1, y_2) in new coordinate system.

For constraint *i*: $y_i = b_i - a_i x$

New coordinates: Distance from new tight constraints.



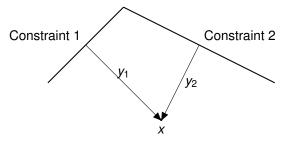
 y_i is distance from constraint i

x is at (y_1, y_2) in new coordinate system.

For constraint i: $y_i = b_i - a_i x$

Recall that for origin: x_i was distance from constraint $x_i \ge 0$.

New coordinates: Distance from new tight constraints.



 y_i is distance from constraint i

x is at (y_1, y_2) in new coordinate system.

For constraint i: $y_i = b_i - a_i x$

Recall that for origin: x_i was distance from constraint $x_i \ge 0$.

At origin in new coordinate system!

Rewrite linear program with new coordinates.

$$\max 2x_1 + 5x_2 \\
 2x_1 - x_2 \le 4 \\
 x_1 + 2x_2 \le 9 \\
 -x_1 + x_2 \le 3 \\
 x_1 \ge 0 \\
 x_2 \ge 0$$
(1)
(2)
(3)
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(5)

New variables: $y_1 = x_1$, $y_2 = 3 + x_1 - x_2$.

Rewrite linear program with new coordinates.

$$\max 2x_1 + 5x_2
2x_1 - x_2 \le 4
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New variables:
$$y_1 = x_1$$
, $y_2 = 3 + x_1 - x_2$.

Solve for x_i 's: $x_1 = y_1$ and $x_2 = 3 - y_2 + y_1$.

Rewrite linear program with new coordinates.

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New variables: $y_1 = x_1$, $y_2 = 3 + x_1 - x_2$.

Solve for x_i 's: $x_1 = y_1$ and $x_2 = 3 - y_2 + y_1$.

Plug in for x_1 and x_2 :

Rewrite linear program with new coordinates.

$$\max 2x_{1} + 5x_{2}$$

$$2x_{1} - x_{2} \le 4$$

$$x_{1} + 2x_{2} \le 9$$

$$-x_{1} + x_{2} \le 3$$

$$x_{1} \ge 0$$

$$x_{2} \ge 0$$
(1)
(2)
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Solve for x_i 's: $x_1 = y_1$ and $x_2 = 3 - y_2 + y_1$.

Plug in for x_1 and x_2 : objective funcition

Rewrite linear program with new coordinates.

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New variables:
$$y_1 = x_1$$
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Solve for
$$x_i$$
's: $x_1 = y_1$ and $x_2 = 3 - y_2 + y_1$.

Plug in for x_1 and x_2 : objective funcition max $2x_1 + 5x_2$

$$\max 2x_1 + 5x_2
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, $y_2 = 3 + x_1 - x_2$.
Solve for x_i 's: $x_1 = y_1$ and $x_2 = 3 - y_2 + y_1$.
Plug in for x_1 and x_2 : objective funcition
$$\max_{x \in \mathcal{X}_1 + 5x_2} 2(y_1) + 5(3 - y_2 + y_1)$$

$$\max 2x_1 + 5x_2 \\
 2x_1 - x_2 \le 4 \\
 x_1 + 2x_2 \le 9 \\
 -x_1 + x_2 \le 3 \\
 x_1 \ge 0 \\
 x_2 \ge 0$$

$$5$$

New variables:
$$y_1 = x_1$$
, $y_2 = 3 + x_1 - x_2$.
Solve for x_i 's: $x_1 = y_1$ and $x_2 = 3 - y_2 + y_1$.
Plug in for x_1 and x_2 : objective funcition $\max 2x_1 + 5x_2 \max 2(y_1) + 5(3 - y_2 + y_1) \max 15 + 7y_1 - 5y_2$

$$\max 2x_1 + 5x_2 \\
 2x_1 - x_2 \le 4 \\
 x_1 + 2x_2 \le 9 \\
 -x_1 + x_2 \le 3 \\
 x_1 \ge 0 \\
 x_2 \ge 0$$
(1)
(2)
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(5)

```
New variables: y_1 = x_1, y_2 = 3 + x_1 - x_2.

Solve for x_i's: x_1 = y_1 and x_2 = 3 - y_2 + y_1.

Plug in for x_1 and x_2: objective funcition \max 2x_1 + 5x_2 \max 2(y_1) + 5(3 - y_2 + y_1) \max 15 + 7y_1 - 5y_2 Are we optimal?
```

$$\max 2x_1 + 5x_2
2x_1 - x_2 \le 4
 x_1 + 2x_2 \le 9
 -x_1 + x_2 \le 3
 x_1 \ge 0
 x_2 \ge 0$$

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New variables: y_1 = x_1, y_2 = 3 + x_1 - x_2.

Solve for x_i's: x_1 = y_1 and x_2 = 3 - y_2 + y_1.

Plug in for x_1 and x_2: objective funcition \max 2x_1 + 5x_2 \max 2(y_1) + 5(3 - y_2 + y_1) \max 15 + 7y_1 - 5y_2 Are we optimal? Yes!
```

$$\max 2x_1 + 5x_2
2x_1 - x_2 \le 4
 x_1 + 2x_2 \le 9
-x_1 + x_2 \le 3
 x_1 \ge 0
 x_2 \ge 0$$

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New variables: y_1 = x_1, y_2 = 3 + x_1 - x_2.

Solve for x_i's: x_1 = y_1 and x_2 = 3 - y_2 + y_1.

Plug in for x_1 and x_2: objective funcition
\max_{max} 2x_1 + 5x_2
\max_{max} 2(y_1) + 5(3 - y_2 + y_1)
\max_{max} 15 + 7y_1 - 5y_2 \text{ Are we optimal? Yes! Maybe not!}
```

$$\max 2x_1 + 5x_2$$
 $2x_1 - x_2 \le 4$
 $x_1 + 2x_2 \le 9$
 $-x_1 + x_2 \le 3$
 $x_1 \ge 0$
 $x_2 \ge 0$
(1)
(2)
(4)
(5)

```
New variables: y_1 = x_1, y_2 = 3 + x_1 - x_2.

Solve for x_i's: x_1 = y_1 and x_2 = 3 - y_2 + y_1.

Plug in for x_1 and x_2: objective funcition \max 2x_1 + 5x_2 \\ \max 2(y_1) + 5(3 - y_2 + y_1) \\ \max 15 + 7y_1 - 5y_2 Are we optimal? Yes! Maybe not! No.
```

$$\max 2x_1 + 5x_2
2x_1 - x_2 \le 4
 x_1 + 2x_2 \le 9
 -x_1 + x_2 \le 3
 x_1 \ge 0
 x_2 \ge 0$$

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New variables: y_1 = x_1, y_2 = 3 + x_1 - x_2.

Solve for x_i's: x_1 = y_1 and x_2 = 3 - y_2 + y_1.

Plug in for x_1 and x_2: objective funcition \max_{max} 2x_1 + 5x_2 \\ \max_{max} 2(y_1) + 5(3 - y_2 + y_1) \\ \max_{max} 15 + 7y_1 - 5y_2 \text{ Are we optimal? Yes! Maybe not! No.}
Positive coefficient for increasing y_1.
```

max
$$15+7y_1-5y_2$$

 $y_1+y_2 \le 7$ ①
$$3y_1-2y_2 \le 3$$
 ②
$$y_2 \ge 0$$
 ③
$$y_1 \ge 0$$
 ④
$$-y_1+y_2 \le 3$$
 ⑤

max
$$15+7y_1-5y_2$$

 $y_1+y_2 \le 7$ ①
$$3y_1-2y_2 \le 3$$
 ②
$$y_2 \ge 0$$
 ③
$$y_1 \ge 0$$
 ④
$$-y_1+y_2 \le 3$$
 ⑤

 y_1, y_2 are non-negative just like x_i 's. (Constraints are satisfied!)

max
$$15+7y_1-5y_2$$

 $y_1+y_2 \le 7$ ①
$$3y_1-2y_2 \le 3$$
 ②
$$y_2 \ge 0$$
 ③
$$y_1 \ge 0$$
 ④
$$-y_1+y_2 \le 3$$
 ⑤

 y_1, y_2 are non-negative just like x_i 's. (Constraints are satisfied!) Improve by increasing y_1 .

max
$$15+7y_1-5y_2$$

 $y_1+y_2 \le 7$ ①
$$3y_1-2y_2 \le 3$$
 ②
$$y_2 \ge 0$$
 ③
$$y_1 \ge 0$$
 ④
$$-y_1+y_2 \le 3$$
 ⑤

 y_1, y_2 are non-negative just like x_i 's. (Constraints are satisfied!) Improve by increasing y_1 .

Which is tight?

max
$$15+7y_1-5y_2$$

 $y_1+y_2 \le 7$ ①
$$3y_1-2y_2 \le 3$$
 ②
$$y_2 \ge 0$$
 ③
$$y_1 \ge 0$$
 ④
$$-y_1+y_2 \le 3$$
 ⑤

 y_1, y_2 are non-negative just like x_i 's. (Constraints are satisfied!) Improve by increasing y_1 .

Which is tight? 1?

max
$$15+7y_1-5y_2$$

 $y_1+y_2 \le 7$ ①
$$3y_1-2y_2 \le 3$$
 ②
$$y_2 \ge 0$$
 ③
$$y_1 \ge 0$$
 ④
$$-y_1+y_2 \le 3$$
 ⑤

 y_1, y_2 are non-negative just like x_i 's. (Constraints are satisfied!) Improve by increasing y_1 .

Which is tight? 1? 2?

max
$$15+7y_1-5y_2$$

 $y_1+y_2 \le 7$ ①
$$3y_1-2y_2 \le 3$$
 ②
$$y_2 \ge 0$$
 ③
$$y_1 \ge 0$$
 ④
$$-y_1+y_2 \le 3$$
 ⑤

 y_1, y_2 are non-negative just like x_i 's. (Constraints are satisfied!) Improve by increasing y_1 .

Which is tight? 1 ? 2 ? 3 ?

max
$$15+7y_1-5y_2$$

 $y_1+y_2 \le 7$ ①
$$3y_1-2y_2 \le 3$$
 ②
$$y_2 \ge 0$$
 ③
$$y_1 \ge 0$$
 ④
$$-y_1+y_2 \le 3$$
 ⑤

 y_1, y_2 are non-negative just like x_i 's. (Constraints are satisfied!) Improve by increasing y_1 .

Which is tight? 1? 2? 3? 4?

max
$$15+7y_1-5y_2$$

 $y_1+y_2 \le 7$ ①
$$3y_1-2y_2 \le 3$$
 ②
$$y_2 \ge 0$$
 ③
$$y_1 \ge 0$$
 ④
$$-y_1+y_2 \le 3$$
 ⑤

 y_1, y_2 are non-negative just like x_i 's. (Constraints are satisfied!) Improve by increasing y_1 .

Which is tight? 1? 2? 3? 4? 5?

max
$$15+7y_1-5y_2$$

 $y_1+y_2 \le 7$ ①
$$3y_1-2y_2 \le 3$$
 ②
$$y_2 \ge 0$$
 ③
$$y_1 \ge 0$$
 ④
$$-y_1+y_2 \le 3$$
 ⑤

 y_1, y_2 are non-negative just like x_i 's. (Constraints are satisfied!) Improve by increasing y_1 .

Which is tight? 1? 2? 3? 4? 5? Note: $y_2 = 0$.

max
$$15+7y_1-5y_2$$

 $y_1+y_2 \le 7$ ①
$$3y_1-2y_2 \le 3$$
 ②
$$y_2 \ge 0$$
 ③
$$y_1 \ge 0$$
 ④
$$-y_1+y_2 \le 3$$
 ⑤

 y_1, y_2 are non-negative just like x_i 's. (Constraints are satisfied!) Improve by increasing y_1 .

Note: $y_2 = 0$.

Smallest right hand side divided by (positive) coefficient of y_2 !

max
$$15+7y_1-5y_2$$

 $y_1+y_2 \le 7$ ①
$$3y_1-2y_2 \le 3$$
 ②
$$y_2 \ge 0$$
 ③
$$y_1 \ge 0$$
 ④
$$-y_1+y_2 \le 3$$
 ⑤

 y_1, y_2 are non-negative just like x_i 's. (Constraints are satisfied!) Improve by increasing y_1 .

Note: $y_2 = 0$.

Smallest right hand side divided by (positive) coefficient of y_2 !

max
$$15+7y_1-5y_2$$

 $y_1+y_2 \le 7$ ①
$$3y_1-2y_2 \le 3$$
 ②
$$y_2 \ge 0$$
 ③
$$y_1 \ge 0$$
 ④
$$-y_1+y_2 \le 3$$
 ⑤

 y_1, y_2 are non-negative just like x_i 's. (Constraints are satisfied!) Improve by increasing y_1 .

Note: $y_2 = 0$.

Smallest right hand side divided by (positive) coefficient of y_2 ! Inequality 2!

max
$$15+7y_1-5y_2$$

 $y_1+y_2 \le 7$ ①
$$3y_1-2y_2 \le 3$$
 ②
$$y_2 \ge 0$$
 ③
$$y_1 \ge 0$$
 ④
$$-y_1+y_2 \le 3$$
 ⑤

 y_1, y_2 are non-negative just like x_i 's. (Constraints are satisfied!) Improve by increasing y_1 .

Note: $y_2 = 0$.

Smallest right hand side divided by (positive) coefficient of y_2 ! Inequality (2)!

New vertex: tight constraints 3 and 2.

max
$$15 + 7y_1 - 5y_2$$

 $y_1 + y_2 \le 7$ ①
$$3y_1 - 2y_2 \le 3$$
 ②
$$y_2 \ge 0$$
 ③
$$y_1 \ge 0$$
 ④
$$-y_1 + y_2 \le 3$$
 ⑤

 y_1, y_2 are non-negative just like x_i 's. (Constraints are satisfied!) Improve by increasing y_1 .

Note: $y_2 = 0$.

Smallest right hand side divided by (positive) coefficient of y_2 ! Inequality (2)!

New vertex: tight constraints (3) and (2).

New solution: $y_1 = 1$, $y_2 = 0$.

max
$$15+7y_1-5y_2$$

 $y_1+y_2 \le 7$ ①
$$3y_1-2y_2 \le 3$$
 ②
$$y_2 \ge 0$$
 ③
$$y_1 \ge 0$$
 ④
$$-y_1+y_2 \le 3$$
 ⑤

 y_1, y_2 are non-negative just like x_i 's. (Constraints are satisfied!) Improve by increasing y_1 .

Note: $y_2 = 0$.

Smallest right hand side divided by (positive) coefficient of y_2 ! Inequality (2)!

New vertex: tight constraints (3) and (2).

New solution: $y_1 = 1$, $y_2 = 0$. New Objective Value:

max
$$15+7y_1-5y_2$$

 $y_1+y_2 \le 7$ ①
$$3y_1-2y_2 \le 3$$
 ②
$$y_2 \ge 0$$
 ③
$$y_1 \ge 0$$
 ④
$$-y_1+y_2 \le 3$$
 ⑤

 y_1, y_2 are non-negative just like x_i 's. (Constraints are satisfied!) Improve by increasing y_1 .

Which is tight? 1? 2? 3? 4? 5?

Note: $y_2 = 0$.

Smallest right hand side divided by (positive) coefficient of y_2 !

Inequality 2!

New vertex: tight constraints (3) and (2).

New solution: $y_1 = 1$, $y_2 = 0$. New Objective Value: 12 + 7(1) - 5(0)

max
$$15+7y_1-5y_2$$

 $y_1+y_2 \le 7$ ①
$$3y_1-2y_2 \le 3$$
 ②
$$y_2 \ge 0$$
 ③
$$y_1 \ge 0$$
 ④
$$-y_1+y_2 \le 3$$
 ⑤

 y_1, y_2 are non-negative just like x_i 's. (Constraints are satisfied!) Improve by increasing y_1 .

Which is tight? 1)? 2)? 3)? 4)? 5)?

Note: $y_2 = 0$.

Smallest right hand side divided by (positive) coefficient of y_2 !

Inequality 2!

New vertex: tight constraints (3) and (2).

New solution: $y_1 = 1$, $y_2 = 0$. New Objective Value: 12 + 7(1) - 5(0) = 22.

Rewrite:
$$z_2 = y_2$$

 $z_1 = 3 - 3y_1 + 2y_2$

Rewrite:
$$z_2 = y_2$$

 $z_1 = 3 - 3y_1 + 2y_2 \rightarrow$

Rewrite: $z_2 = y_2$ $z_1 = 3 - 3y_1 + 2y_2 \rightarrow y_1 = -\frac{1}{3}z_1 + \frac{2}{3}z_2 + 1$ Objective function.

Rewrite:
$$z_2 = y_2$$

 $z_1 = 3 - 3y_1 + 2y_2 \rightarrow y_1 = -\frac{1}{3}z_1 + \frac{2}{3}z_2 + 1$
Objective function.
max $15 + 7(-\frac{1}{3}z_1 + \frac{2}{3}z_2 + 1) - 5z_2$

Rewrite:
$$z_2 = y_2$$

 $z_1 = 3 - 3y_1 + 2y_2 \rightarrow y_1 = -\frac{1}{3}z_1 + \frac{2}{3}z_2 + 1$
Objective function.
max $15 + 7(-\frac{1}{3}z_1 + \frac{2}{3}z_2 + 1) - 5z_2$
 $\max 22 - \frac{7}{3}z_1 - \frac{1}{3}z_2$

Rewrite:
$$z_2 = y_2$$

 $z_1 = 3 - 3y_1 + 2y_2 \rightarrow y_1 = -\frac{1}{3}z_1 + \frac{2}{3}z_2 + 1$
Objective function.
max $15 + 7(-\frac{1}{3}z_1 + \frac{2}{3}z_2 + 1) - 5z_2$
max $22 - \frac{7}{3}z_1 - \frac{1}{3}z_2$ Optimal?

Rewrite:
$$z_2 = y_2$$

 $z_1 = 3 - 3y_1 + 2y_2 \rightarrow y_1 = -\frac{1}{3}z_1 + \frac{2}{3}z_2 + 1$
Objective function.
max $15 + 7(-\frac{1}{3}z_1 + \frac{2}{3}z_2 + 1) - 5z_2$
max $22 - \frac{7}{3}z_1 - \frac{1}{3}z_2$ Optimal? Yes!

Rewrite:
$$z_2 = y_2$$

 $z_1 = 3 - 3y_1 + 2y_2 \rightarrow y_1 = -\frac{1}{3}z_1 + \frac{2}{3}z_2 + 1$
Objective function.
max $15 + 7(-\frac{1}{3}z_1 + \frac{2}{3}z_2 + 1) - 5z_2$
max $22 - \frac{7}{3}z_1 - \frac{1}{3}z_2$ Optimal? Yes! Maybe not!

Rewrite:
$$z_2 = y_2$$

 $z_1 = 3 - 3y_1 + 2y_2 \rightarrow y_1 = -\frac{1}{3}z_1 + \frac{2}{3}z_2 + 1$
Objective function.
max $15 + 7(-\frac{1}{3}z_1 + \frac{2}{3}z_2 + 1) - 5z_2$
max $22 - \frac{7}{3}z_1 - \frac{1}{3}z_2$ Optimal? Yes! Maybe not! Optimal point!

Rewrite:
$$z_2 = y_2$$

 $z_1 = 3 - 3y_1 + 2y_2 \rightarrow y_1 = -\frac{1}{3}z_1 + \frac{2}{3}z_2 + 1$
Objective function.
max $15 + 7(-\frac{1}{3}z_1 + \frac{2}{3}z_2 + 1) - 5z_2$
max $22 - \frac{7}{3}z_1 - \frac{1}{3}z_2$ Optimal? Yes! Maybe not! Optimal point! Increasing z_1, z_2 makes things worse.

In each step:

In each step:

LP in coordinate system from tight constraints.

In each step:

LP in coordinate system from tight constraints.

Optimal?

In each step:

LP in coordinate system from tight constraints.

Optimal?

Does objective function have nonnegative multiplier?

In each step:

LP in coordinate system from tight constraints.

Optimal?

Does objective function have nonnegative multiplier?

$$\max 15 + 7y_1 - 5y_2$$
.

In each step:

LP in coordinate system from tight constraints.

Optimal?

Does objective function have nonnegative multiplier? $\max 15 + 7y_1 - 5y_2$.

Go to tight constraint along improving coordinate.

In each step:

LP in coordinate system from tight constraints.

Optimal?

Does objective function have nonnegative multiplier? $\max 15 + 7y_1 - 5y_2$.

Go to tight constraint along improving coordinate. $3y_1 - 2y_2 < 3$.

In each step:

LP in coordinate system from tight constraints.

Optimal?

Does objective function have nonnegative multiplier?

$$\max 15 + 7y_1 - 5y_2$$
.

Go to tight constraint along improving coordinate.

$$3y_1 - 2y_2 \le 3$$
.

Express LP in coordinate system for new tight constraints.

In each step:

LP in coordinate system from tight constraints.

Optimal?

Does objective function have nonnegative multiplier?

$$\max 15 + 7y_1 - 5y_2$$
.

Go to tight constraint along improving coordinate.

$$3y_1 - 2y_2 \le 3$$
.

Express LP in coordinate system for new tight constraints. See previous slides!

In each step:

LP in coordinate system from tight constraints.

Optimal?

Does objective function have nonnegative multiplier? $\max 15 + 7y_1 - 5y_2$.

Go to tight constraint along improving coordinate. $3y_1 - 2y_2 < 3$.

Express LP in coordinate system for new tight constraints. See previous slides!

Repeat.

What if origin is not feasible?

What if origin is not feasible? How do you find a feasible vertex?

What if origin is not feasible? How do you find a feasible vertex? An x where $Ax \le b$ and at vertex.

What if origin is not feasible? How do you find a feasible vertex? An x where $Ax \le b$ and at vertex. Make a new linear program.

What if origin is not feasible?

How do you find a feasible vertex?

An x where $Ax \le b$ and at vertex.

Make a new linear program.

Introduce positive variables z_i for inequality i.

What if origin is not feasible?

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Make a new linear program.

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Constraints: $a_i x - z_i \le b_i$.

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Vertex solution (x,z) of value zero

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Vertex solution (x,z) of value zero \implies all z's are zero

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- \implies all z's are zero
- \implies all inequalities are satisfied

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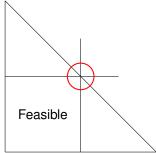
$$\max \sum -z_i$$
.

Vertex solution (x,z) of value zero

- \implies all z's are zero
- \implies all inequalities are satisfied
- \implies *x* is a feasible vertex of $Ax \le b$.

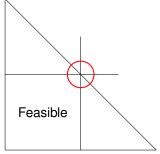
Degenerate vertices.

Degenerate vertices. Intersection of more than *n* constraints.



Degenerate vertices.

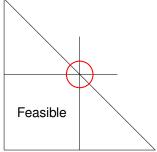
Intersection of more than *n* constraints.



Problem: all neighboring vertices are no better.

Degenerate vertices.

Intersection of more than *n* constraints.

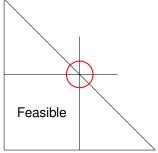


Problem: all neighboring vertices are no better.

Infinite looping: Bland's anticycling rule.

Degenerate vertices.

Intersection of more than *n* constraints.



Problem: all neighboring vertices are no better.

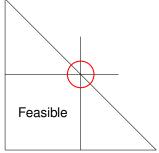
Infinite looping: Bland's anticycling rule.

Or Perturb problem a bit.

Degeneracy.

Degenerate vertices.

Intersection of more than *n* constraints.

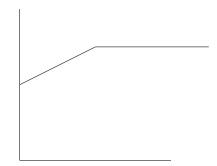


Problem: all neighboring vertices are no better.

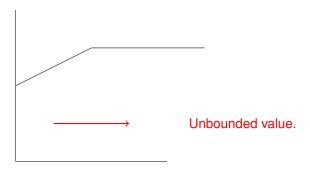
Infinite looping: Bland's anticycling rule.

Or Perturb problem a bit. Unlikely to intersect!

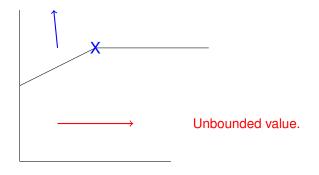
Unboundedness.



Unboundedness.



Unboundedness.



Simplex can tell difference.

From X: either unbounded improvement or optimal.

Check optimality? O(n).

Check optimality? O(n).

Find tight constraint:

Check optimality? O(n).

Find tight constraint: O(m) constraints.

Check optimality? O(n).

Find tight constraint:

O(m) constraints. O(1) time per constraint.

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Find tight constraint:

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O(m) total.

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Find new coordinate system, rewrite LP.

Check optimality? O(n).

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Recall $y_i = b_i - a_i x$

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Find tight constraint:

O(m) constraints. O(1) time per constraint.

O(m) total.

Find new coordinate system, rewrite LP.

Recall $y_i = b_i - a_i x$

Rewrite in terms of y_i .

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Naively: $O(n^3)$ time.

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O(m) total.

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Naively: $O(n^3)$ time.

Only one new constraint.

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O(m) total.

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Plug in.

Naively: $O(n^3)$ time.

Only one new constraint. Have x in terms of y_1, \ldots, y_n .

Check optimality? O(n).

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Only one y_i goes to y'_i .

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O(nm) time to update LP. (Like backsolving.)

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How many simplex steps?

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How many simplex steps?

Could be large. Exponential in worst case!

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Fast, in practice!

The negations of coefficients of new function!

The negations of coefficients of new function! Let A' be matrix of "tight constraints."

The negations of coefficients of new function!

Let A' be matrix of "tight constraints."

Coordinate System: y = b' - A'x.

The negations of coefficients of new function!

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Coordinate System: y = b' - A'x. $x = (A')^{-1}(b' - y)$

The negations of coefficients of new function!

Let A' be matrix of "tight constraints."

Coordinate System:
$$y = b' - A'x$$
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$$\max cx = \max c((A')^{-1})(b'-y) = \max c((A')^{-1})b') - (c(A')^{-1})y.$$

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 gives coefficients of new objective function.

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All positive at optimal!

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 for subset of tight equations.

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All positive at optimal! $\rightarrow z \ge 0$

$$A'^Tz = A^T((A')^{-1})^Tc = c$$
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$$\rightarrow A'^T z > c$$
.

Set all other dual variables to 0. $\implies A^T z \ge c$.

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Feasible!

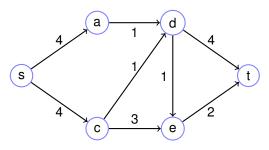
Next Up: Maximum Flow.

Maximum Flow.

Flow network G = (V, E), source s, sink $t \in V$, capacities $c_e > 0$.

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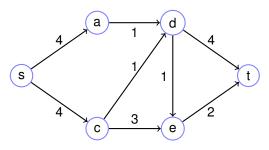


Find Flow: fe

- 1. $0 \le f_e \le c_e$. "Capacity constraints."
- 2. If u is not s or t $\sum_{(w,u)\in E} f_{wu} = \sum_{(u,w)\in E} f_{uw}.$

Maximize: size $(f) = \sum_{(s,u) \in E} f_{su}$.

Flow network G = (V, E), source s, sink $t \in V$, capacities $c_e > 0$.

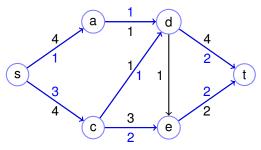


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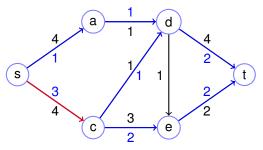


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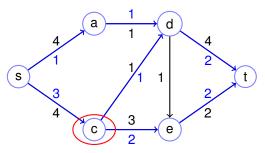


Find Flow: fe

- 1. $0 \le f_e \le c_e$. "Capacity constraints." $3 = f_{s,c} \le c_{s,c} = 4$.
- 2. If u is not s or t $\sum_{(w,u)\in E} f_{wu} = \sum_{(u,w)\in E} f_{uw}.$

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Flow network G = (V, E), source s, sink $t \in V$, capacities $c_e > 0$.

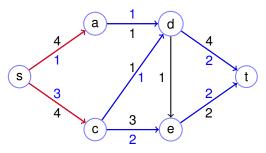


Find Flow: fe

- 1. $0 \le f_e \le c_e$. "Capacity constraints." $3 = f_{s,c} \le c_{s,c} = 4$.
- 2. If *u* is not *s* or *t* $\sum_{(w,u)\in E} f_{wu} = \sum_{(u,w)\in E} f_{uw}. \ 3 = f_{s,c} = f_{c,d} + f_{c,e} = 2 + 1.$

Maximize: size(f) = $\sum_{(s,u)\in E} f_{su}$.

Flow network G = (V, E), source s, sink $t \in V$, capacities $c_e > 0$.



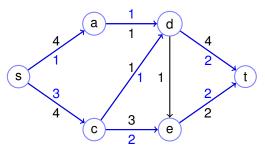
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Maximize: size(f) = $\sum_{(s,u)\in E} f_{su}$. $f_{sa} + f_{sc} = 1 + 3 = 4$

Flow network G = (V, E), source s, sink $t \in V$, capacities $c_e > 0$.



Find Flow: fe

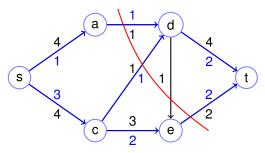
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Optimal?

Flow network G = (V, E), source s, sink $t \in V$, capacities $c_e > 0$.



Find Flow: fe

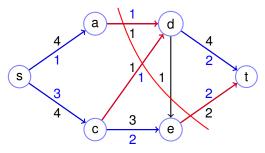
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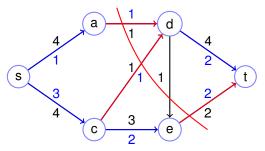
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Optimal? $c_{ad} + c_{cd} + c_{et} = 1 + 1 + 2 = 4$.

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Find Flow: fe

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$$\sum_{(w,u)\in E} f_{wu} = \sum_{(u,w)\in E} f_{uw}$$
. $3 = f_{s,c} = f_{c,d} + f_{c,e} = 2 + 1$.

Maximize: size(f) = $\sum_{(s,u)\in E} f_{su}$. $f_{sa} + f_{sc} = 1 + 3 = 4$

Optimal? $c_{ad} + c_{cd} + c_{et} = 1 + 1 + 2 = 4$.

Any s-t cut gives an upper bound.

S-T cut.

An s-t cut is a partition of V into S and T where $s \in S$ and $t \in T$. Its capacity is the total capacity of edges from S to T.

Find Flow: fe

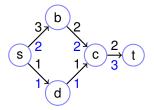
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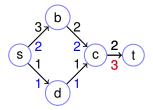
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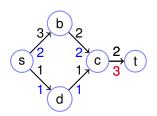
Find Flow: fe

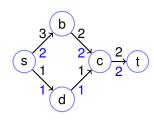
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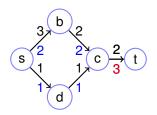
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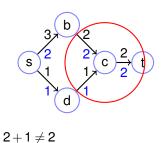




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Vertex solution to linear program must be integral!

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Add to flow variables along path.

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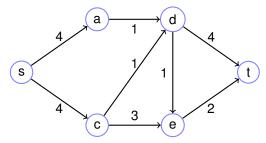
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Repeat.



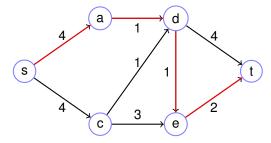
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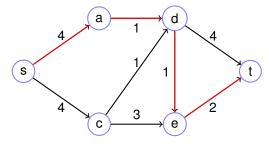
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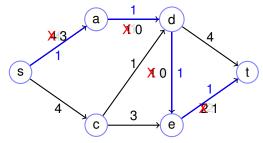
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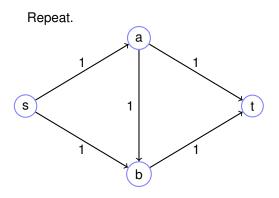


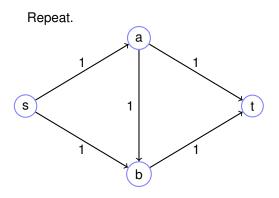
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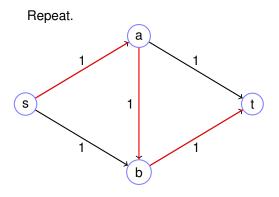
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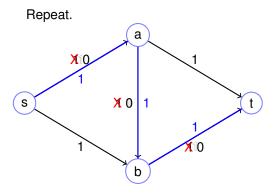
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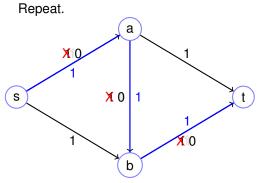






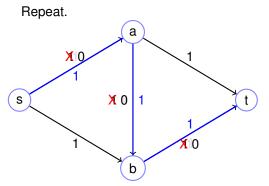


Find *s* to *t* path with remaining capacity. Add to flow along path. Update remaining capacity.



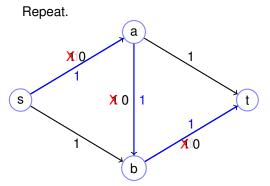
No remaining path.

Find *s* to *t* path with remaining capacity. Add to flow along path. Update remaining capacity.

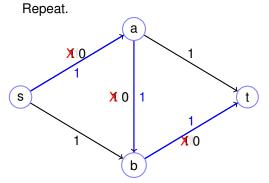


Find *s* to *t* path with remaining capacity. Add to flow along path.

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Find *s* to *t* path with remaining capacity. Add to flow along path. Or reduce flow on reverse edge. Update remaining capacity.



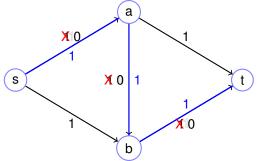
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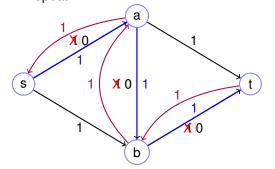
Update remaining capacity.

Reduce $r_e = c_e - f_e$

Repeat.

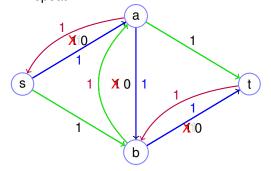


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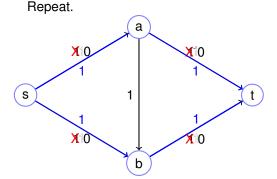
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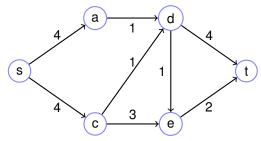
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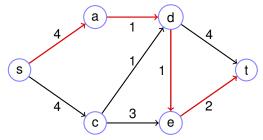
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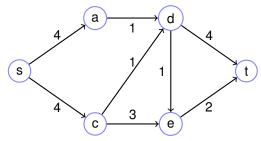
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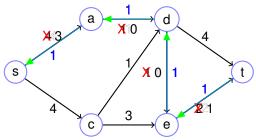
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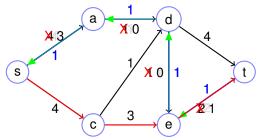
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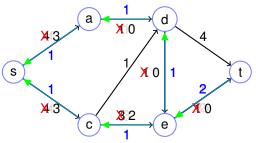
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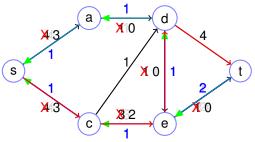
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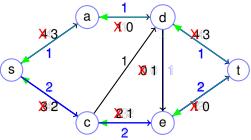
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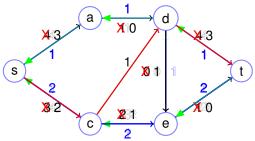
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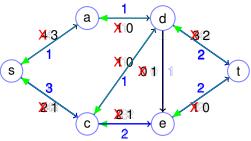
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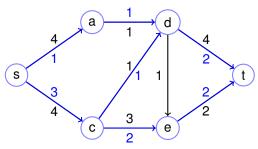


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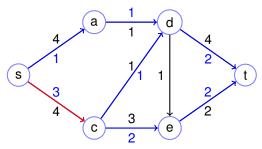




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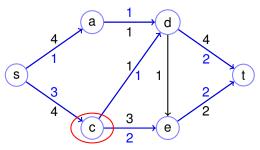
Maximize: size(f) = $\sum_{(s,u)\in E} f_{su}$.



Find Flow: fe

- 1. $0 \le f_e \le c_e$. "Capacity constraints." $3 = f_{s,c} \le c_{s,c} = 4$.
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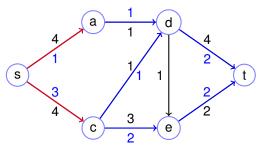


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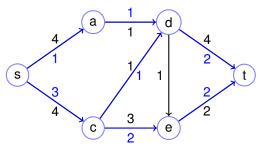


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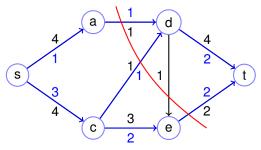
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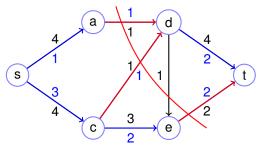
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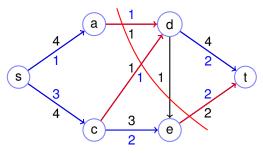
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Any s-t cut gives an upper bound.

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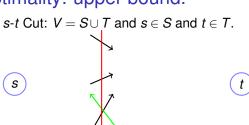
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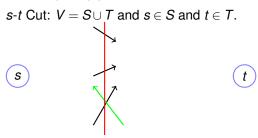
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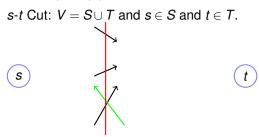
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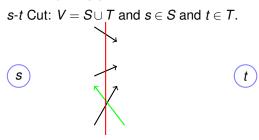


Lemma: Capacity of any s-t cut is an upper bound on the flow.



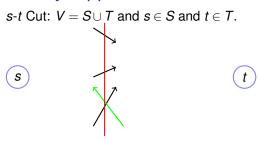
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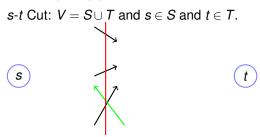
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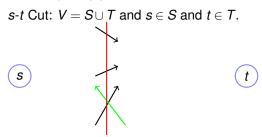
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Flow out of (S) = Flow out of s.



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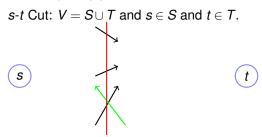
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For any valid flow, $f: E \to Z+$, the flow out of S (into T) $\sum_{e \in S \times T} f_e$

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- t Cut: $V = S \cup T$ and $s \in S$ and $t \in T$.

Lemma: Capacity of any s-t cut is an upper bound on the flow.

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(s)

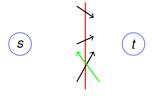


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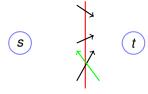
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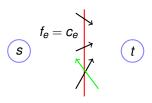
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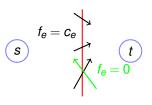
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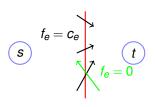
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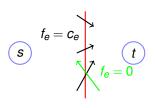
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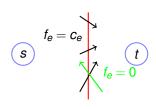
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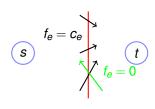
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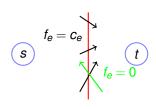
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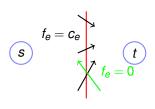
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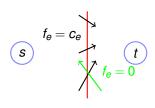
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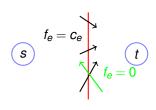
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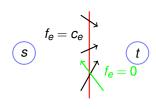
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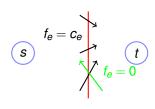
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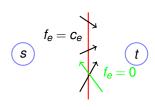
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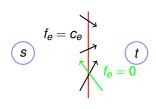
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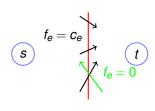
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→ Flow is maximum!!

Cut is minimum s-t cut too!

"any flow" \leq "any cut" and this flow = this cut.

 \rightarrow Maximum flow and minimum s-t cut!

Celebrated max flow -minimum cut theorem.

Theorem: In any flow network, the maximum *s-t* flow is equal to the minimum cut.

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Celebrate!

Lecture in a Minute

Simplex Implementation: Start at a (feasible) vertex.

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(defined by linear system $A'x = [b', 0, \dots, 0]$).

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Coordinate system: distance from tight constraints.

Vertex at origin in coordinate system.

O(mn) time to update linear system.

Until no better neighboring vertex.

Objective function in coordinate system is non-positive.

Dual Variables: new objective function!

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 Until no better neighboring vertex.
 Objective function in coordinate system is non-positive.
  Dual Variables: new objective function!
Maximum flow.
  "Greedy" augment path...
     Except reverse old decisions ..
     Reverse residual capacities.
  (Friday): Optimality?
     No augmenting path \Longrightarrow
      s-t cut size = flow value.
    Find flow and s-t cut with equal value!
```