

Continue Review: linear programming.

Continue Review: linear programming. Simplex and Matching.

Continue Review: linear programming. Simplex and Matching. Taking Duals.

Plant Carrots or Peas?

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2\$ bushel of carrots.

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2\$ bushel of carrots. 4\$ for peas.

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2\$ bushel of carrots. 4\$ for peas.

Carrots take 3 unit of water/bushel.

Plant Carrots or Peas?

2\$ bushel of carrots. 4\$ for peas.

Carrots take 3 unit of water/bushel. Peas take 2 units of water/bushel.

Plant Carrots or Peas?

2\$ bushel of carrots. 4\$ for peas.

Carrots take 3 unit of water/bushel. Peas take 2 units of water/bushel.

100 units of water.

Plant Carrots or Peas?

2\$ bushel of carrots. 4\$ for peas.

Carrots take 3 unit of water/bushel. Peas take 2 units of water/bushel.

100 units of water.

Peas require 2 yards/bushel of sunny land.

Plant Carrots or Peas?

2\$ bushel of carrots. 4\$ for peas.

Carrots take 3 unit of water/bushel. Peas take 2 units of water/bushel.

100 units of water.

Peas require 2 yards/bushel of sunny land. Carrots require 1 yard/bushel of shadyland.

Plant Carrots or Peas?

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Carrots take 3 unit of water/bushel. Peas take 2 units of water/bushel.

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Peas require 2 yards/bushel of sunny land. Carrots require 1 yard/bushel of shadyland.

Garden has 60 yards of sunny land and 75 yards of shady land.

Plant Carrots or Peas?

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100 units of water.

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Garden has 60 yards of sunny land and 75 yards of shady land.

Plant Carrots or Peas?

2\$ bushel of carrots. 4\$ for peas.

Carrots take 3 unit of water/bushel. Peas take 2 units of water/bushel.

100 units of water.

Peas require 2 yards/bushel of sunny land. Carrots require 1 yard/bushel of shadyland.

Garden has 60 yards of sunny land and 75 yards of shady land.

To pea or not to pea, that is the question!

4\$ for peas.

4\$ for peas. 2\$ bushel of carrots.

4\$ for peas. 2\$ bushel of carrots. x_1 - to pea!

4\$ for peas. 2\$ bushel of carrots. x_1 - to pea! x_2 to carrot

4\$ for peas. 2\$ bushel of carrots. x_1 - to pea! x_2 to carrot ?

4\$ for peas. 2\$ bushel of carrots. x_1 - to pea! x_2 to carrot ? Money $4x_1 + 2x_2$

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4\$ for peas. 2\$ bushel of carrots. x_1 - to pea! x_2 to carrot ? Money $4x_1 + 2x_2$ maximize max $4x_1 + 2x_2$. Carrots take 2 unit of water/bushel.

4\$ for peas. 2\$ bushel of carrots. x_1 - to pea! x_2 to carrot ? Money $4x_1 + 2x_2$ maximize max $4x_1 + 2x_2$. Carrots take 2 unit of water/bushel. Peas take 3 units of water/bushel.

4\$ for peas. 2\$ bushel of carrots. x_1 - to pea! x_2 to carrot ? Money $4x_1 + 2x_2$ maximize max $4x_1 + 2x_2$. Carrots take 2 unit of water/bushel. Peas take 3 units of water/bushel. 100 units of water.

4\$ for peas. 2\$ bushel of carrots. x_1 - to peal x_2 to carrot ? Money $4x_1 + 2x_2$ maximize max $4x_1 + 2x_2$. Carrots take 2 unit of water/bushel. Peas take 3 units of water/bushel. 100 units of water.

 $3x_1 + 2x_2 \le 100$

4\$ for peas. 2\$ bushel of carrots. x_1 - to pea! x_2 to carrot ? Money $4x_1 + 2x_2$ maximize $\max 4x_1 + 2x_2$. Carrots take 2 unit of water/bushel. Peas take 3 units of water/bushel. 100 units of water.

 $3x_1 + 2x_2 \le 100$

Peas 2 yards/bushel of sunny land. Have 40 sq yards.

4\$ for peas. 2\$ bushel of carrots. x_1 - to pea! x_2 to carrot ? Money $4x_1 + 2x_2$ maximize $\max 4x_1 + 2x_2$. Carrots take 2 unit of water/bushel. Peas take 3 units of water/bushel. 100 units of water.

 $3x_1 + 2x_2 \le 100$

Peas 2 yards/bushel of sunny land. Have 40 sq yards. $2x_1 \leq 40$

4\$ for peas. 2\$ bushel of carrots. x_1 - to pea! x_2 to carrot ? Money $4x_1 + 2x_2$ maximize $\max 4x_1 + 2x_2$. Carrots take 2 unit of water/bushel. Peas take 3 units of water/bushel. 100 units of water.

 $3x_1 + 2x_2 \le 100$

Peas 2 yards/bushel of sunny land. Have 40 sq yards. $2x_1 \le 40$

Carrots get 3 yards/bushel of shady land. Have 75 sq. yards.

4\$ for peas. 2\$ bushel of carrots. x_1 - to pea! x_2 to carrot ? Money $4x_1 + 2x_2$ maximize $\max 4x_1 + 2x_2$. Carrots take 2 unit of water/bushel. Peas take 3 units of water/bushel. 100 units of water.

 $3x_1 + 2x_2 \le 100$

Peas 2 yards/bushel of sunny land. Have 40 sq yards. $2x_1 \le 40$ Carrots get 3 yards/bushel of shady land. Have 75 sq. yards. $3x_2 \le 75$

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Peas 2 yards/bushel of sunny land. Have 40 sq yards. $2x_1 \le 40$ Carrots get 3 yards/bushel of shady land. Have 75 sq. yards. $3x_2 \le 75$ Can't make negative!

4\$ for peas. 2\$ bushel of carrots. x_1 - to pea! x_2 to carrot ? Money $4x_1 + 2x_2$ maximize $\max 4x_1 + 2x_2$. Carrots take 2 unit of water/bushel. Peas take 3 units of water/bushel. 100 units of water.

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Peas 2 yards/bushel of sunny land. Have 40 sq yards. $2x_1 \le 40$ Carrots get 3 yards/bushel of shady land. Have 75 sq. yards. $3x_2 \le 75$ Can't make negative! $x_1, x_2 \ge 0$.

4\$ for peas. 2\$ bushel of carrots. x_1 - to pea! x_2 to carrot ? Money $4x_1 + 2x_2$ maximize $\max 4x_1 + 2x_2$. Carrots take 2 unit of water/bushel. Peas take 3 units of water/bushel. 100 units of water.

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Peas 2 yards/bushel of sunny land. Have 40 sq yards. $2x_1 \le 40$ Carrots get 3 yards/bushel of shady land. Have 75 sq. yards. $3x_2 \le 75$ Can't make negative! $x_1, x_2 \ge 0$. A linear program.

4\$ for peas. 2\$ bushel of carrots. x_1 - to pea! x_2 to carrot ? Money $4x_1 + 2x_2$ maximize $\max 4x_1 + 2x_2$. Carrots take 2 unit of water/bushel. Peas take 3 units of water/bushel. 100 units of water.

 $3x_1 + 2x_2 \le 100$

Peas 2 yards/bushel of sunny land. Have 40 sq yards. $2x_1 \le 40$ Carrots get 3 yards/bushel of shady land. Have 75 sq. yards. $3x_2 \le 75$ Can't make negative! $x_1, x_2 \ge 0$.

A linear program.

$$max 4x_1 + 2x_2 2x_1 \le 40 3x_2 \le 75 3x_1 + 2x_2 \le 100 x_1, x_2 \ge 0$$

$$max 4x_1 + 2x_2 2x_1 \le 40 3x_2 \le 75 3x_1 + 2x_2 \le 100 x_1, x_2 \ge 0$$

Optimal point?

$$max 4x_1 + 2x_2 2x_1 \le 40 3x_2 \le 75 3x_1 + 2x_2 \le 100 x_1, x_2 \ge 0$$

Optimal point? Try every point
$$max 4x_1 + 2x_2 2x_1 \le 40 3x_2 \le 75 3x_1 + 2x_2 \le 100 x_1, x_2 \ge 0$$

Try every point if we only had time!

$$\begin{array}{c} \max 4x_{1}+2x_{2}\\ 2x_{1}\leq 40\\ 3x_{2}\leq 75\\ 3x_{1}+2x_{2}\leq 100\\ x_{1},x_{2}\geq 0 \end{array}$$

Try every point if we only had time! How many points?

$$max 4x_1 + 2x_2 2x_1 \le 40 3x_2 \le 75 3x_1 + 2x_2 \le 100 x_1, x_2 \ge 0$$

Try every point if we only had time!

How many points?

Real numbers?

$$max 4x_1 + 2x_2 2x_1 \le 40 3x_2 \le 75 3x_1 + 2x_2 \le 100 x_1, x_2 \ge 0$$

Try every point if we only had time!

How many points?

Real numbers?

Infinite.

$$max 4x_1 + 2x_2 2x_1 \le 40 3x_2 \le 75 3x_1 + 2x_2 \le 100 x_1, x_2 \ge 0$$

Try every point if we only had time!

How many points?

Real numbers?

Infinite. Uncountably infinite!

$$max 4x_1 + 2x_2 2x_1 \le 40 3x_2 \le 75 3x_1 + 2x_2 \le 100 x_1, x_2 \ge 0$$



Optimal point?



Optimal point?



Optimal point?



Optimal point?



Optimal point?









Convex.

Any two points in region connected by a line in region.



Convex.

Any two points in region connected by a line in region. Algebraically:



Convex.

Any two points in region connected by a line in region. Algebraically:

If x and x' satisfy onstraint,



Convex.

Any two points in region connected by a line in region. Algebraically:

If x and x' satisfy onstraint,

$$\rightarrow x'' = \alpha x + (1 - \alpha)x$$

















Optimal at pointy part of feasible region!





Intersection of two of the constraints (lines in two dimensions)!



Vertex of region.

Intersection of two of the constraints (lines in two dimensions)! Try every vertex!



Intersection of two of the constraints (lines in two dimensions)! Try every vertex! Choose best.



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 $O(m^2)$ if *m* constraints and 2 variables.



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 $O(m^2)$ if *m* constraints and 2 variables.

nm?
$$\binom{m}{n}$$
? *n*+*m*?



Vertex of region.

Intersection of two of the constraints (lines in two dimensions)! Try every vertex! Choose best.

 $O(m^2)$ if *m* constraints and 2 variables.

$$nm? \binom{m}{n}? n+m? \binom{m}{n}$$



Vertex of region.

Intersection of two of the constraints (lines in two dimensions)! Try every vertex! Choose best.

 $O(m^2)$ if *m* constraints and 2 variables.

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Vertex of region.

Intersection of two of the constraints (lines in two dimensions)! Try every vertex! Choose best.

 $O(m^2)$ if *m* constraints and 2 variables.

```
nm? \binom{m}{n}? n+m? \binom{m}{n}
Finite!!!!!!
```


Optimal at pointy part of feasible region!

Vertex of region.

Intersection of two of the constraints (lines in two dimensions)! Try every vertex! Choose best.

 $O(m^2)$ if *m* constraints and 2 variables.

For *n* variables, *m* constraints, how many?

$$nm? \binom{m}{n}? n+m?$$

 $\binom{m}{n}$

Finite!!!!!!

Exponential in the number of variables.









Until no better neighbor.



Until no better neighbor. This example.





(0,0) objective 0. \rightarrow (0,25) objective 50.



Until no better neighbor. This example.

(0,0) objective 0. \rightarrow (0,25) objective 50.

 \rightarrow (16 $\frac{2}{3}$,25) objective 115 $\frac{2}{3}$



Simplex: Start at vertex. Move to better neighboring vertex. Until no better neighbor. This example. (0,0) objective $0. \rightarrow (0,25)$ objective 50.

$$\rightarrow$$
 (16 $\frac{2}{3}$,25) objective 115 $\frac{2}{3}$

 \rightarrow (20,20) objective 120.



Simplex: Start at vertex. Move to better neighboring vertex. Until no better neighbor. This example. (0,0) objective 0. \rightarrow (0,25) objective 50. \rightarrow (16²/₂,25) objective 115²/₂

 \rightarrow (20,20) objective 120.

Duality:



(0,0) objective 0.
$$\rightarrow$$
 (0,25) objective 50.

$$ightarrow$$
 (16 $\frac{2}{3}$,25) objective 115 $\frac{2}{3}$

$$\rightarrow$$
 (20,20) objective 120.

Duality:

Add blue equations to get objective function?



(0,0) objective 0.
$$\rightarrow$$
 (0,25) objective 50.

$$ightarrow$$
 (16 $rac{2}{3}$,25) objective 115 $rac{2}{3}$

$$\rightarrow$$
 (20,20) objective 120.

Duality:

Add blue equations to get objective function?

1/3 times first plus second.



(0,0) objective 0.
$$\rightarrow$$
 (0,25) objective 50.

$$ightarrow$$
 (16 $rac{2}{3}$,25) objective 115 $rac{2}{3}$

$$\rightarrow$$
 (20,20) objective 120.

Duality:

Add blue equations to get objective function?

1/3 times first plus second.

Get $4x_1 + 2x_2 \le 120$.



(0,0) objective 0.
$$\rightarrow$$
 (0,25) objective 50.

$$ightarrow$$
 (16 $rac{2}{3}$,25) objective 115 $rac{2}{3}$

$$\rightarrow$$
 (20,20) objective 120.

Duality:

Add blue equations to get objective function?

1/3 times first plus second.

Get $4x_1 + 2x_2 \le 120$. Every solution must satisfy this inequality!



(0,0) objective 0.
$$\rightarrow$$
 (0,25) objective 50.

$$ightarrow$$
 (16 $rac{2}{3}$,25) objective 115 $rac{2}{3}$

$$\rightarrow$$
 (20,20) objective 120.

Duality:

Add blue equations to get objective function?

1/3 times first plus second.

Get $4x_1 + 2x_2 \le 120$. Every solution must satisfy this inequality! Objective value: 120.

Can we do better?



(0,0) objective 0.
$$\rightarrow$$
 (0,25) objective 50.

$$ightarrow$$
 (16 $rac{2}{3}$,25) objective 115 $rac{2}{3}$

$$\rightarrow$$
 (20,20) objective 120.

Duality:

Add blue equations to get objective function?

1/3 times first plus second.

Get $4x_1 + 2x_2 \le 120$. Every solution must satisfy this inequality! Objective value: 120.

Can we do better? No!

Dual problem: add equations to get best upper bound.

$\max x_{1} + 8x_{2}$ $x_{1} \le 4$ $x_{2} \le 3$ $x_{1} + 2x_{2} \le 7$ $x_{1}, x_{2} \ge 0$

 $\max x_{1} + 8x_{2}$ $x_{1} \leq 4$ $x_{2} \leq 3$ $x_{1} + 2x_{2} \leq 7$ $x_{1}, x_{2} \geq 0$

One Solution: $x_1 = 1, x_2 = 3$.

$\max x_{1} + 8x_{2}$ $x_{1} \leq 4$ $x_{2} \leq 3$ $x_{1} + 2x_{2} \leq 7$ $x_{1}, x_{2} \geq 0$

One Solution: $x_1 = 1, x_2 = 3$. Value is 25.

 $\max x_{1} + 8x_{2}$ $x_{1} \leq 4$ $x_{2} \leq 3$ $x_{1} + 2x_{2} \leq 7$ $x_{1}, x_{2} \geq 0$

One Solution: $x_1 = 1, x_2 = 3$. Value is 25. Best possible?

 $\max x_{1} + 8x_{2}$ $x_{1} \leq 4$ $x_{2} \leq 3$ $x_{1} + 2x_{2} \leq 7$ $x_{1}, x_{2} \geq 0$

One Solution: $x_1 = 1, x_2 = 3$. Value is 25.

Best possible?

For any solution. $x_1 \le 4$ and $x_2 \le 3$...

 $\max x_1 + 8x_2$ $x_1 \le 4$ $x_2 \le 3$ $x_1 + 2x_2 \le 7$ $x_1, x_2 \ge 0$

One Solution: $x_1 = 1, x_2 = 3$. Value is 25.

Best possible?

For any solution. $x_1 \le 4$ and $x_2 \le 3$so $x_1 + 8x_2 \le 4 + 8(3) = 28$. Added equation 1 and 8 times equation 2 yields bound on objective..

 $\max x_{1} + 8x_{2}$ $x_{1} \leq 4$ $x_{2} \leq 3$ $x_{1} + 2x_{2} \leq 7$ $x_{1}, x_{2} > 0$

One Solution: $x_1 = 1, x_2 = 3$. Value is 25.

Best possible?

For any solution. $x_1 \le 4$ and $x_2 \le 3$ so $x_1 + 8x_2 \le 4 + 8(3) = 28$. Added equation 1 and 8 times equation 2 yields bound on objective.. Better solution?

 $\max x_{1} + 8x_{2}$ $x_{1} \leq 4$ $x_{2} \leq 3$ $x_{1} + 2x_{2} \leq 7$ $x_{1}, x_{2} > 0$

One Solution: $x_1 = 1, x_2 = 3$. Value is 25.

Best possible?

For any solution. $x_1 \le 4$ and $x_2 \le 3$ so $x_1 + 8x_2 \le 4 + 8(3) = 28$. Added equation 1 and 8 times equation 2 yields bound on objective.. Better solution? Better upper bound?

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For any solution. $x_1 \le 4$ and $x_2 \le 3$ so $x_1 + 8x_2 \le 4 + 8(3) = 28$. Added equation 1 and 8 times equation 2 yields bound on objective.. Better solution? Better upper bound?

 $\max x_{1} + 8x_{2}$ $x_{1} \leq 4$ $x_{2} \leq 3$ $x_{1} + 2x_{2} \leq 7$ $x_{1}, x_{2} \geq 0$

Solution value: 25.

 $\max x_{1} + 8x_{2}$ $x_{1} \leq 4$ $x_{2} \leq 3$ $x_{1} + 2x_{2} \leq 7$ $x_{1}, x_{2} \geq 0$

Solution value: 25. Add equation 1 and 8 times equation 2 gives.

 $\max x_{1} + 8x_{2}$ $x_{1} \leq 4$ $x_{2} \leq 3$ $x_{1} + 2x_{2} \leq 7$ $x_{1}, x_{2} \geq 0$

Solution value: 25. Add equation 1 and 8 times equation 2 gives.. $x_1 + 8x_2 \le 4 + 24 = 28.$

 $\max x_{1} + 8x_{2}$ $x_{1} \leq 4$ $x_{2} \leq 3$ $x_{1} + 2x_{2} \leq 7$ $x_{1}, x_{2} \geq 0$

Solution value: 25. Add equation 1 and 8 times equation 2 gives.. $x_1 + 8x_2 \le 4 + 24 = 28.$

Better way to add equations to get bound on function?

 $\max x_{1} + 8x_{2}$ $x_{1} \leq 4$ $x_{2} \leq 3$ $x_{1} + 2x_{2} \leq 7$ $x_{1}, x_{2} \geq 0$

Solution value: 25. Add equation 1 and 8 times equation 2 gives.. $x_1 + 8x_2 \le 4 + 24 = 28.$

Better way to add equations to get bound on function? Sure:

 $\max x_{1} + 8x_{2}$ $x_{1} \leq 4$ $x_{2} \leq 3$ $x_{1} + 2x_{2} \leq 7$ $x_{1}, x_{2} \geq 0$

Solution value: 25. Add equation 1 and 8 times equation 2 gives.. $x_1 + 8x_2 \le 4 + 24 = 28.$

Better way to add equations to get bound on function? Sure: 6 times equation 2 and 1 times equation 3.

 $\max x_{1} + 8x_{2}$ $x_{1} \leq 4$ $x_{2} \leq 3$ $x_{1} + 2x_{2} \leq 7$ $x_{1}, x_{2} \geq 0$

Solution value: 25. Add equation 1 and 8 times equation 2 gives.. $x_1 + 8x_2 \le 4 + 24 = 28.$

Better way to add equations to get bound on function? Sure: 6 times equation 2 and 1 times equation 3. $x_1 + 8x_2 \le 6(3) + 7 = 25.$

 $max x_1 + 8x_2$ $x_1 \le 4$ $x_2 \le 3$ $x_1 + 2x_2 \le 7$ $x_1, x_2 \ge 0$

Solution value: 25. Add equation 1 and 8 times equation 2 gives.. $x_1 + 8x_2 \le 4 + 24 = 28.$

Better way to add equations to get bound on function? Sure: 6 times equation 2 and 1 times equation 3. $x_1 + 8x_2 \le 6(3) + 7 = 25.$

Thus, the value is at most 25.

 $max x_1 + 8x_2$ $x_1 \le 4$ $x_2 \le 3$ $x_1 + 2x_2 \le 7$ $x_1, x_2 \ge 0$

Solution value: 25. Add equation 1 and 8 times equation 2 gives.. $x_1 + 8x_2 \le 4 + 24 = 28.$

Better way to add equations to get bound on function? Sure: 6 times equation 2 and 1 times equation 3.

 $x_1 + 8x_2 \le 6(3) + 7 = 25.$

Thus, the value is at most 25.

The upper bound is same as solution!

 $max x_1 + 8x_2$ $x_1 \le 4$ $x_2 \le 3$ $x_1 + 2x_2 \le 7$ $x_1, x_2 \ge 0$

Solution value: 25. Add equation 1 and 8 times equation 2 gives.. $x_1 + 8x_2 \le 4 + 24 = 28.$

Better way to add equations to get bound on function? Sure: 6 times equation 2 and 1 times equation 3.

 $x_1 + 8x_2 \le 6(3) + 7 = 25.$

Thus, the value is at most 25.

The upper bound is same as solution!

Proof of optimality!

Idea: Add up positive linear combination of inequalities to "get" upper bound on optimization function.
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Will this always work?

Idea: Add up positive linear combination of inequalities to "get" upper bound on optimization function.

Will this always work?

How to find best upper bound?

Best Upper Bound.

Multiplier Inequality

<i>Y</i> 1	<i>x</i> ₁	\leq 4
<i>y</i> ₂		<i>x</i> ₂ ≤ 3
y 3	<i>x</i> ₁ +	$2x_2 \le 7$

Adding equations thusly...

Best Upper Bound.

Multiplier Inequality

<i>Y</i> 1	<i>x</i> ₁	\leq 4
y 2		<i>x</i> ₂ ≤ 3
y 3	<i>x</i> ₁ +	$2x_2 \le 7$

Adding equations thusly... $(y_1 + y_3)x_1 + (y_2 + 2y_3)x_2 \le 4y_1 + 3y_2 + 7y_3.$

Best Upper Bound.

Multiplier Inequality

y 1	<i>x</i> ₁	\leq 4
y 2		<i>x</i> ₂ ≤ 3
y 3	<i>x</i> ₁ +	$2x_2 \le 7$

Adding equations thusly...

$$(y_1+y_3)x_1+(y_2+2y_3)x_2 \le 4y_1+3y_2+7y_3.$$

The left hand side should "dominate" optimization function:

Best Upper Bound.

Multiplier Inequality

y 1	<i>x</i> ₁	\leq 4
y 2		<i>x</i> ₂ ≤ 3
y 3	<i>x</i> ₁ +	$2x_2 \le 7$

Adding equations thusly...

 $(y_1+y_3)x_1+(y_2+2y_3)x_2 \le 4y_1+3y_2+7y_3.$

The left hand side should "dominate" optimization function: If $y_1, y_2, y_3 \ge 0$

Best Upper Bound.

Multiplier Inequality

y 1	<i>x</i> ₁	\leq 4
y 2		<i>x</i> ₂ ≤ 3
y 3	<i>x</i> ₁ +	$2x_2 \le 7$

Adding equations thusly...

 $(y_1+y_3)x_1+(y_2+2y_3)x_2 \le 4y_1+3y_2+7y_3.$

The left hand side should "dominate" optimization function:

If $y_1, y_2, y_3 \ge 0$ and $y_1 + y_3 \ge 1$ and $y_2 + 2y_3 \ge 8$ then..

Best Upper Bound.

Multiplier Inequality

y 1	<i>x</i> ₁	\leq 4
y 2		<i>x</i> ₂ ≤ 3
y 3	<i>x</i> ₁ +	$2x_2 \le 7$

Adding equations thusly...

 $(y_1+y_3)x_1+(y_2+2y_3)x_2 \leq 4y_1+3y_2+7y_3.$

The left hand side should "dominate" optimization function:

If $y_1, y_2, y_3 \ge 0$ and $y_1 + y_3 \ge 1$ and $y_2 + 2y_3 \ge 8$ then.. $x_1 + 8x_2 \le 4y_1 + 3y_2 + 7y_3$

Best Upper Bound.

Multiplier Inequality

y 1	<i>x</i> ₁	\leq 4
y 2		<i>x</i> ₂ ≤ 3
y 3	<i>x</i> ₁ +	$2x_2 \le 7$

Adding equations thusly...

 $(y_1+y_3)x_1+(y_2+2y_3)x_2 \leq 4y_1+3y_2+7y_3.$

The left hand side should "dominate" optimization function:

If $y_1, y_2, y_3 \ge 0$ and $y_1 + y_3 \ge 1$ and $y_2 + 2y_3 \ge 8$ then.. $x_1 + 8x_2 \le 4y_1 + 3y_2 + 7y_3$

Find best y_i's to minimize upper bound?

Find best y_i 's to minimize upper bound?

Find best y_i 's to minimize upper bound?

Again: If you find $y_1, y_2, y_3 \ge 0$

Find best y_i 's to minimize upper bound?

Again: If you find $y_1, y_2, y_3 \ge 0$ and $y_1 + y_3 \ge 1$ and $y_2 + 2y_3 \ge 8$ then..

Find best y_i 's to minimize upper bound?

Again: If you find $y_1, y_2, y_3 \ge 0$ and $y_1 + y_3 \ge 1$ and $y_2 + 2y_3 \ge 8$ then.. $x_1 + 8x_2 \le 4y_1 + 3y_2 + 7y_3$

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 $\begin{array}{l} \min 4y_1 + 3y_2 + 7y_3 \\ y_1 + y_3 \geq 1 \\ y_2 + 2y_3 \geq 8 \\ y_1, y_2, y_3 \geq 0 \end{array}$

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$$y_1 + y_3 \ge 1$$

 $y_2 + 2y_3 \ge 8$
 $y_1, y_2, y_3 \ge 0$

A linear program.

Find best y_i 's to minimize upper bound?

Again: If you find
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A linear program. The Dual linear program.

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 $y_1 + y_3 \ge 1$ $y_2 + 2y_3 \ge 8$ $y_1, y_2, y_3 \ge 0$

A linear program. The Dual linear program.

Primal: $(x_1, x_2) = (1,3)$; Dual: $(y_1, y_2, y_3) = (0,6,1)$.

Find best y_i's to minimize upper bound?

Again: If you find
$$y_1, y_2, y_3 \ge 0$$

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 $y_1 + y_3 \ge 1$ $y_2 + 2y_3 \ge 8$ $y_1, y_2, y_3 \ge 0$

A linear program. The Dual linear program.

Primal: $(x_1, x_2) = (1,3)$; Dual: $(y_1, y_2, y_3) = (0,6,1)$. Value of both is 25!

Find best y_i 's to minimize upper bound?

Again: If you find
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A linear program. The Dual linear program.

Primal: $(x_1, x_2) = (1,3)$; Dual: $(y_1, y_2, y_3) = (0,6,1)$.

Value of both is 25!

Primal is optimal

Find best y_i 's to minimize upper bound?

Again: If you find
$$y_1, y_2, y_3 \ge 0$$

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 $x_1 + 8x_2 \le 4y_1 + 3y_2 + 7y_3$
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A linear program. The Dual linear program.

Primal: $(x_1, x_2) = (1,3)$; Dual: $(y_1, y_2, y_3) = (0,6,1)$.

Value of both is 25!

Primal is optimal ... and dual is optimal!

In general.

Primal LP	<u>Dual LP</u>
max c · x	min y ^T b
$Ax \leq b$	$y^T A \ge c$
$x \ge 0$	$y \ge 0$

In general.

Primal LP	<u>Dual LP</u>
max <i>c</i> ⋅ <i>x</i>	min $y^T b$
$Ax \leq b$	$y^T A \ge c$
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Theorem: If a linear program has a bounded value, then its dual is bounded and has the same value.

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Weak Duality: primal (P) \leq dual (D)

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Weak Duality: primal (P) \leq dual (D) Feasible (x, y)

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Primal LP	<u>Dual LP</u>
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Weak Duality: primal (P) \leq dual (D)
Feasible (x, y)
P(x)
```

In general.

<u>Primal LP</u>	<u>Dual LP</u>
max <i>c</i> ⋅ <i>x</i>	min $y^T b$
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Theorem: If a linear program has a bounded value, then its dual is bounded and has the same value.

Weak Duality: primal (P) \leq dual (D) Feasible (x, y) $P(x) = c \cdot x$

In general.

Primal LP	<u>Dual LP</u>
max <i>c</i> ⋅ <i>x</i>	min $y^T b$
$Ax \leq b$	$y^T A \ge c$
$x \ge 0$	<i>y</i> ≥ 0

Theorem: If a linear program has a bounded value, then its dual is bounded and has the same value.

Weak Duality: primal (P) \leq dual (D)

Feasible (x, y) $P(x) = c \cdot x \le y^T A x$

In general.

Primal LP	<u>Dual LP</u>
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Theorem: If a linear program has a bounded value, then its dual is bounded and has the same value.

Weak Duality: primal (P) \leq dual (D)

Feasible
$$(x, y)$$

 $P(x) = c \cdot x \le y^T A x \le y^T b \cdot x = D(y).$

In general.

Primal LP	<u>Dual LP</u>
max <i>c</i> ⋅ <i>x</i>	min $y^T b$
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Feasible
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 $P(x) = c \cdot x \le y^T A x \le y^T b \cdot x = D(y).$

Strong Duality: next lecture.

Primal LP	<u>Dual LP</u>
max <i>c</i> ⋅ <i>x</i>	min $y^T b$
$Ax \leq b$	$y^T A \ge c$
$x \ge 0$	<i>y</i> ≥ 0

Given *A*, *b*, *c*, and feasible solutions *x* and *y*.

Primal LP	<u>Dual LP</u>
max c · x	min $y^T b$
$Ax \leq b$	$y^T A \ge c$
$x \ge 0$	<i>y</i> ≥ 0

Given *A*, *b*, *c*, and feasible solutions *x* and *y*.

Solutions *x* and *y* are both optimal if and only if $x_i(c_i - (y^T A)_i) = 0$, and $y_j(b_j - (Ax)_j)$.

Primal LP	<u>Dual LP</u>
max <i>c</i> ⋅ <i>x</i>	min $y^T b$
$Ax \leq b$	$y^T A \ge c$
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Primal LP	<u>Dual LP</u>
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Solutions *x* and *y* are both optimal if and only if $x_i(c_i - (y^T A)_i) = 0$, and $y_j(b_j - (Ax)_j)$. $x_i(c_i - (y^T A)_i) = 0 \rightarrow \sum_i (c_i - (y^T A)_i) x_i$

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Primal LP	<u>Dual LP</u>
max <i>c</i> ⋅ <i>x</i>	min $y^T b$
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Primal LP	<u>Dual LP</u>
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Primal LP	<u>Dual LP</u>
max c·x	min $y^T b$
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Solutions x and y are both optimal if and only if $x_i(c_i - (y^T A)_i) = 0$, and $y_j(b_j - (Ax)_j)$. $x_i(c_i - (y^T A)_i) = 0 \rightarrow$ $\sum_i (c_i - (y^T A)_i)x_i = cx - y^T Ax \rightarrow cx = y^T Ax$. $y_j(b_j - (Ax)_j) = 0 \rightarrow$ $\sum_i y_j(b_j - (Ax)_j) = yb - y^T Ax \rightarrow by = y^T Ax$. cx = by.

Primal LP	<u>Dual LP</u>
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If both are feasible, $cx \leq by$, so must be optimal.

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If both are feasible, $cx \leq by$, so must be optimal.

In words: nonzero dual variables only for tight constraints!



 $max 4x_1 + 2x_2$ $3x_1 \le 60$ $3x_2 \le 75$ $3x_1 + 2x_2 \le 100$ $x_1, x_2 \ge 0$





Simplex: Start at vertex. Move to better neighboring vertex. Until no better neighbor.



Simplex: Start at vertex. Move to better neighboring vertex. Until no better neighbor. Duality:



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Add blue equations to get objective function?



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1/3 times first plus second.



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1/3 times first plus second.

Get $4x_1 + 2x_2 \le 120$.



Simplex: Start at vertex. Move to better neighboring vertex.

Until no better neighbor. Duality:

Add blue equations to get objective function?

1/3 times first plus second.

Get $4x_1 + 2x_2 \le 120$. Every solution must satisfy this inequality!



Simplex: Start at vertex. Move to better neighboring vertex.

Until no better neighbor. Duality:

Add blue equations to get objective function?

1/3 times first plus second.

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Simplex: Start at vertex. Move to better neighboring vertex.

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1/3 times first plus second.

Get $4x_1 + 2x_2 \le 120$. Every solution must satisfy this inequality! Geometrically and Complementary slackness:



Simplex: Start at vertex. Move to better neighboring vertex.

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1/3 times first plus second.

Get $4x_1 + 2x_2 \le 120$. Every solution must satisfy this inequality! Geometrically and Complementary slackness:

Add tight constraints to "dominate objective function."



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Until no better neighbor. Duality:

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1/3 times first plus second.

Get $4x_1 + 2x_2 \le 120$. Every solution must satisfy this inequality! Geometrically and Complementary slackness:

Add tight constraints to "dominate objective function."

Don't add this equation!



Simplex: Start at vertex. Move to better neighboring vertex.

Until no better neighbor. Duality:

Add blue equations to get objective function?

1/3 times first plus second.

Get $4x_1 + 2x_2 \le 120$. Every solution must satisfy this inequality! Geometrically and Complementary slackness:

Add tight constraints to "dominate objective function."

Don't add this equation! Shifts.

Example: review.

$$\begin{array}{ll} \max x_1 + 8x_2 & \min 4y_1 + 3y_2 + 7y_3 \\ x_1 \leq 4 & y_1 + y_3 \geq 1 \\ x_2 \leq 3 & y_2 + 2y_3 \geq 8 \\ x_1 + 2x_2 \leq 7 & x_1, x_2 \geq 0 \\ y_1, y_2, y_3 \geq 0 \end{array}$$

"Matrix form"

$$\begin{split} \max[1,8] \cdot [x_1,x_2] & \min[4,3,7] \cdot [y_1,y_2,y_3] \\ \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 2 \end{pmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \leq \begin{bmatrix} 4 \\ 3 \\ 7 \end{bmatrix} & [y_1,y_2,y_3] \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 2 \end{pmatrix} \geq \begin{bmatrix} 1 \\ 8 \end{bmatrix} \\ [x_1,x_2] \geq 0 & [y_1,y_2,y_3] \geq 0 \end{split}$$

Matrix equations.

$$\max[1,8] \cdot [x_1, x_2] \qquad \min[4,3,7] \cdot [y_1, y_2, y_3] \\ \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 2 \end{pmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \le \begin{bmatrix} 4 \\ 3 \\ 7 \end{bmatrix} \qquad [y_1, y_2, y_3] \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 2 \end{pmatrix} \ge \begin{bmatrix} 1 \\ 8 \end{bmatrix} \\ [x_1, x_2] \ge 0 \qquad [y_1, y_2, y_3] \ge 0$$

We can rewrite the above in matrix form.

$$A = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 2 \end{pmatrix} \qquad \qquad c = [1,8]b = [4,3,7]$$

The primal is $Ax \le b$, max $c \cdot x$, $x \ge 0$. The dual is $y^T A \ge c$, min $b \cdot y$, $y \ge 0$.

or..."Rules for taking duals"

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Standard:

 $Ax \leq b, \max cx, x \geq 0 \leftrightarrow y^T A \geq c, \min by, y \geq 0.$

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Standard:

 $Ax \leq b, \max cx, x \geq 0 \leftrightarrow y^T A \geq c, \min by, y \geq 0.$ min $\leftrightarrow \max$

or..."Rules for taking duals"

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 $Ax \le b, \max cx, x \ge 0 \leftrightarrow y^T A \ge c, \min by, y \ge 0.$ min \leftrightarrow max

 $\geq \leftrightarrow \leq$

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Standard:

 $Ax \le b, \max cx, x \ge 0 \leftrightarrow y^T A \ge c, \min by, y \ge 0.$ min $\leftrightarrow \max$

 $\geq \leftrightarrow \leq$

"inequalities" \leftrightarrow "nonnegative variables"

or..."Rules for taking duals"

Standard:

 $Ax \leq b, \max cx, x \geq 0 \leftrightarrow y^T A \geq c, \min by, y \geq 0.$

 $\mathsf{min} \leftrightarrow \mathsf{max}$

 $\geq \leftrightarrow \leq$

"inequalities" \leftrightarrow "nonnegative variables"

"nonnegative variables" \leftrightarrow "inequalities"

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Standard:

 $Ax \leq b, \max cx, x \geq 0 \leftrightarrow y^T A \geq c, \min by, y \geq 0.$

 $\text{min} \leftrightarrow \text{max}$

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"inequalities" \leftrightarrow "nonnegative variables"

"nonnegative variables" \leftrightarrow "inequalities"

One more useful trick: Equality constraints.

or..."Rules for taking duals"

Standard:

 $Ax \leq b, \max cx, x \geq 0 \leftrightarrow y^T A \geq c, \min by, y \geq 0.$

 $\text{min} \leftrightarrow \text{max}$

 $\geq \leftrightarrow \leq$

"inequalities" \leftrightarrow "nonnegative variables"

"nonnegative variables" \leftrightarrow "inequalities"

One more useful trick: Equality constraints.

```
"equalities" \leftrightarrow "unrestricted variables."
```

Bipartite Graph $G = (V, E), w : E \rightarrow Z$.

Bipartite Graph G = (V, E), $w : E \to Z$. Find maximum weight perfect matching.

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$$\max \sum_{e} w_{e} x_{e}$$
$$\forall v : \sum_{e=(u,v)} x_{e} = 1$$
$$x_{e} \ge 0$$
Bipartite Graph G = (V, E), $w : E \to Z$. Find maximum weight perfect matching. Solution: x_e indicates whether edge e is in matching.

$$\max \sum_{e} w_{e} x_{e}$$
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$$x_{e} \ge 0$$

Dual.

Bipartite Graph G = (V, E), $w : E \to Z$. Find maximum weight perfect matching. Solution: x_e indicates whether edge e is in matching.

$$\max \sum_{e} w_{e} x_{e}$$
$$\forall v : \sum_{e=(u,v)} x_{e} = 1$$

 $x_e \ge 0$

Dual.

Variable for each constraint.

Bipartite Graph G = (V, E), $w : E \to Z$. Find maximum weight perfect matching. Solution: x_e indicates whether edge e is in matching.

$$\max \sum_{e} w_{e} x_{e}$$
$$\forall v : \sum_{e=(u,v)} x_{e} = 1 \qquad p_{v}$$
$$x_{e} \ge 0$$

Dual.

Variable for each constraint. p_v

Bipartite Graph G = (V, E), $w : E \to Z$. Find maximum weight perfect matching. Solution: x_e indicates whether edge e is in matching.

$$\max \sum_{e} w_{e} x_{e}$$
$$\forall v : \sum_{e=(u,v)} x_{e} = 1 \qquad p_{v}$$
$$x_{e} > 0$$

Dual.

Variable for each constraint. p_v unrestricted. Constraint for each variable.

Bipartite Graph G = (V, E), $w : E \to Z$. Find maximum weight perfect matching. Solution: x_e indicates whether edge e is in matching.

$$\max \sum_{e} \frac{w_e x_e}{w_e x_e}$$
$$\forall v : \sum_{e=(u,v)} \frac{x_e}{x_e} = 1 \qquad p_v$$
$$x_e \ge 0$$

Dual.

Variable for each constraint. p_v unrestricted. Constraint for each variable. Edge e, $p_u + p_v \ge w_e$ Objective function from right hand side.

Bipartite Graph G = (V, E), $w : E \to Z$. Find maximum weight perfect matching. Solution: x_e indicates whether edge e is in matching.

$$\max \sum_{e} w_{e} x_{e}$$
$$\forall v : \sum_{e=(u,v)} x_{e} = 1 \qquad p_{v}$$
$$x_{e} \ge 0$$

Dual.

Variable for each constraint. p_v unrestricted. Constraint for each variable. Edge e, $p_u + p_v \ge w_e$ Objective function from right hand side. min $\sum_v p_v$

Bipartite Graph G = (V, E), $w : E \to Z$. Find maximum weight perfect matching. Solution: x_e indicates whether edge e is in matching.

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Variable for each constraint. p_v unrestricted. Constraint for each variable. Edge e, $p_u + p_v \ge w_e$ Objective function from right hand side. min $\sum_v p_v$

$$\min \sum_{v} p_{v}$$

$$\forall e = (u, v): \quad p_{u} + p_{v} \ge w_{e}$$

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Weak duality?

Bipartite Graph G = (V, E), $w : E \to Z$. Find maximum weight perfect matching. Solution: x_e indicates whether edge e is in matching.

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Variable for each constraint. p_v unrestricted. Constraint for each variable. Edge e, $p_u + p_v \ge w_e$ Objective function from right hand side. min $\sum_v p_v$

$$\min \sum_{v} p_{v}$$

$$\forall \boldsymbol{e} = (\boldsymbol{u}, \boldsymbol{v}) : \quad \boldsymbol{p}_{u} + \boldsymbol{p}_{v} \geq \boldsymbol{w}_{\boldsymbol{e}}$$

Weak duality? Price function upper bounds matching.

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$$\min \sum_{\boldsymbol{v}} \boldsymbol{p}_{\boldsymbol{v}} \\ \forall \boldsymbol{e} = (\boldsymbol{u}, \boldsymbol{v}) : \quad \boldsymbol{p}_{\boldsymbol{u}} + \boldsymbol{p}_{\boldsymbol{v}} \geq \boldsymbol{w}_{\boldsymbol{e}}$$

Weak duality? Price function upper bounds matching. $\sum_{e \in M} w_e x_e \leq \sum_{e=(u,v) \in M} p_u + p_v \leq \sum_v p_u.$

Bipartite Graph G = (V, E), $w : E \to Z$. Find maximum weight perfect matching. Solution: x_e indicates whether edge e is in matching.

$$\max \sum_{e} w_{e} x_{e}$$
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Dual.

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$$\min \sum_{\boldsymbol{v}} \boldsymbol{p}_{\boldsymbol{v}} \\ \forall \boldsymbol{e} = (\boldsymbol{u}, \boldsymbol{v}) : \quad \boldsymbol{p}_{\boldsymbol{u}} + \boldsymbol{p}_{\boldsymbol{v}} \geq \boldsymbol{w}_{\boldsymbol{e}}$$

Weak duality? Price function upper bounds matching.

$$\sum_{e\in M} w_e x_e \leq \sum_{e=(u,v)\in M} p_u + p_v \leq \sum_v p_u.$$

Strong Duality?

Bipartite Graph G = (V, E), $w : E \to Z$. Find maximum weight perfect matching. Solution: x_e indicates whether edge e is in matching.

$$\max \sum_{e} w_{e} x_{e}$$
$$\forall v : \sum_{e=(u,v)} x_{e} = 1 \qquad p_{v}$$
$$x_{e} \ge 0$$

Dual.

Variable for each constraint. p_v unrestricted. Constraint for each variable. Edge e, $p_u + p_v \ge w_e$ Objective function from right hand side. min $\sum_v p_v$

$$\min \sum_{v} p_{v}$$

$$\forall \boldsymbol{e} = (\boldsymbol{u}, \boldsymbol{v}) : \quad \boldsymbol{p}_{u} + \boldsymbol{p}_{v} \geq \boldsymbol{w}_{\boldsymbol{e}}$$

Weak duality? Price function upper bounds matching.

$$\sum_{e\in M} w_e x_e \leq \sum_{e=(u,v)\in M} p_u + p_v \leq \sum_v p_u.$$

Strong Duality? Same value solutions.

Bipartite Graph G = (V, E), $w : E \to Z$. Find maximum weight perfect matching. Solution: x_e indicates whether edge e is in matching.

$$\max \sum_{e} w_{e} x_{e}$$
$$\forall v : \sum_{e=(u,v)} x_{e} = 1 \qquad p_{v}$$
$$x_{e} \ge 0$$

Dual.

Variable for each constraint. p_v unrestricted. Constraint for each variable. Edge e, $p_u + p_v \ge w_e$ Objective function from right hand side. min $\sum_v p_v$

$$\min \sum_{v} p_{v}$$

$$\forall \boldsymbol{e} = (\boldsymbol{u}, \boldsymbol{v}) : \quad \boldsymbol{p}_{u} + \boldsymbol{p}_{v} \geq \boldsymbol{w}_{\boldsymbol{e}}$$

Weak duality? Price function upper bounds matching.

$$\sum_{e \in M} w_e x_e \leq \sum_{e=(u,v) \in M} p_u + p_v \leq \sum_v p_u.$$

Strong Duality? Same value solutions. Hungarian algorithm

Bipartite Graph G = (V, E), $w : E \to Z$. Find maximum weight perfect matching. Solution: x_e indicates whether edge e is in matching.

$$\max \sum_{e} w_{e} x_{e}$$
$$\forall v : \sum_{e=(u,v)} x_{e} = 1 \qquad p_{v}$$
$$x_{e} \ge 0$$

Dual.

Variable for each constraint. p_v unrestricted. Constraint for each variable. Edge e, $p_u + p_v \ge w_e$ Objective function from right hand side. min $\sum_v p_v$

$$\min \sum_{v} p_{v}$$

$$\forall \boldsymbol{e} = (\boldsymbol{u}, \boldsymbol{v}) : \quad \boldsymbol{p}_{u} + \boldsymbol{p}_{v} \geq \boldsymbol{w}_{\boldsymbol{e}}$$

Weak duality? Price function upper bounds matching.

$$\sum_{e \in M} w_e x_e \leq \sum_{e=(u,v) \in M} p_u + p_v \leq \sum_v p_u.$$

Strong Duality? Same value solutions. Hungarian algorithm !!!

Complementary Slackness.

$$\max \sum_{e} \frac{w_e x_e}{w_e x_e}$$
$$\forall v : \sum_{e=(u,v)} \frac{x_e}{x_e} = 1 \qquad p_v$$
$$x_e \ge 0$$



$$\forall \boldsymbol{e} = (\boldsymbol{u}, \boldsymbol{v}): \quad \boldsymbol{p}_{\boldsymbol{u}} + \boldsymbol{p}_{\boldsymbol{v}} \geq \boldsymbol{w}_{\boldsymbol{e}}$$

Complementary Slackness.

$$\max \sum_{e} w_{e} x_{e}$$
$$\forall v : \sum_{e=(u,v)} x_{e} = 1 \qquad p_{v}$$
$$x_{e} \ge 0$$

Dual:

$$\min \sum_{v} p_{v}$$
$$\forall e = (u, v) : p_{u} + p_{v} \ge w_{e}$$

Complementary slackness:

Complementary Slackness.

$$\max \sum_{e} w_{e} x_{e}$$
$$\forall v : \sum_{e=(u,v)} x_{e} = 1 \qquad p_{v}$$
$$x_{e} \ge 0$$

Dual:

$$\min \sum_{v} p_{v}$$

$$\forall \boldsymbol{e} = (\boldsymbol{u}, \boldsymbol{v}) : \quad \boldsymbol{p}_{u} + \boldsymbol{p}_{v} \geq \boldsymbol{w}_{\boldsymbol{e}}$$

Complementary slackness: Only match on tight edges. Nonzero p_u on matched u.

Given G = (V, E), and capacity function $c : E \to Z$, and pairs $(s_1, t_1), \ldots, (s_k, t_k)$ with demands d_1, \ldots, d_k .

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variables: f_p flow on path p.

 P_i -set of paths with endpoints s_i, t_i .

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variables: f_p flow on path p. P_i -set of paths with endpoints s_i, t_i .

 $\min \mu$ $\forall \boldsymbol{e} : \sum_{\boldsymbol{p} \geq \boldsymbol{e}} f_{\boldsymbol{p}} \leq \mu c_{\boldsymbol{e}}$ $\forall \boldsymbol{i} : \sum_{\boldsymbol{p} \in P_{i}} f_{\boldsymbol{p}} = D_{i}$ $f_{\boldsymbol{p}} \geq 0$

Take the dual.

$$\min \mu$$
$$\forall e : \sum_{p \ni e} f_p \le \mu c_e$$
$$\forall i : \sum_{p \in P_i} f_p = D_i$$
$$f_p \ge 0$$

Modify to make it \geq , which "go with min.

Take the dual.

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Modify to make it \geq , which "go with min. And only constants on right hand side.

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Modify to make it \geq , which "go with min. And only constants on right hand side.

$$\begin{aligned} \min \mu \\ \forall \boldsymbol{e} : \mu \boldsymbol{c}_{\boldsymbol{e}} - \sum_{\boldsymbol{p} \ni \boldsymbol{e}} f_{\boldsymbol{p}} \geq \boldsymbol{0} \\ \forall \boldsymbol{i} : \sum_{\boldsymbol{p} \in \boldsymbol{P}_{i}} f_{\boldsymbol{p}} = \boldsymbol{D}_{i} \\ f_{\boldsymbol{p}} \geq \boldsymbol{0} \end{aligned}$$

 $\min \mu$

$$\forall \boldsymbol{e} : \boldsymbol{\mu} \boldsymbol{c}_{\boldsymbol{e}} - \sum_{\boldsymbol{p} \ni \boldsymbol{e}} f_{\boldsymbol{p}} \ge 0$$
$$\forall \boldsymbol{i} : \sum_{\boldsymbol{p} \in \boldsymbol{P}_{i}} f_{\boldsymbol{p}} = \boldsymbol{D}_{i}$$
$$f_{\boldsymbol{p}} \ge 0$$

 $\min \mu$

$$\forall \boldsymbol{e} : \boldsymbol{\mu} \boldsymbol{c}_{\boldsymbol{e}} - \sum_{\boldsymbol{p} \ni \boldsymbol{e}} f_{\boldsymbol{p}} \ge \boldsymbol{0} \qquad \boldsymbol{d}_{\boldsymbol{e}}$$
$$\forall \boldsymbol{i} : \sum_{\boldsymbol{p} \in \boldsymbol{P}_{i}} f_{\boldsymbol{p}} = \boldsymbol{D}_{i} \qquad \boldsymbol{d}_{i}$$

 $f_{p} \geq 0$ Introduce variable for each constraint.

 $\min \mu$

$$\forall e : \mu c_e - \sum_{p \ni e} f_p \ge 0 \qquad d_e$$
$$\forall i : \sum_{p \in P_i} f_p = D_i \qquad d_i$$

 $f_p \ge 0$ Introduce variable for each constraint. Introduce constraint for each var:

min µ

$$\forall e : \mu c_e - \sum_{p \ge e} f_p \ge 0 \qquad d_e$$
$$\forall i : \sum_{p \in P_i} f_p = D_i \qquad d_i$$

 $f_{p} \geq 0$ Introduce variable for each constraint. Introduce constraint for each var:

μ

min µ

$$\forall e : \mu c_e - \sum_{p \ge e} f_p \ge 0 \qquad d_e$$
$$\forall i : \sum_{p \in P_i} f_p = D_i \qquad d_i$$

 $f_{p} \geq 0$ Introduce variable for each constraint. Introduce constraint for each var: $\mu \rightarrow \sum_{e} c_{e} d_{e} = 1.$

 $\min \mu$

$$\forall e : \mu c_e - \sum_{p \ni e} f_p \ge 0 \qquad d_e$$

$$\forall i : \sum_{p \in P_i} f_p = D_i \qquad d_i$$

 $f_p \ge 0$ Introduce variable for each constraint. Introduce constraint for each var: $\mu \rightarrow \sum_e c_e d_e = 1$. f_p

 $\min \mu$

$$\forall e : \mu c_e - \sum_{p \ni e} f_p \ge 0 \qquad d_e$$

$$\forall i : \sum_{p \in P_i} f_p = D_i \qquad d_i$$

 $f_{\rho} \geq 0$ Introduce variable for each constraint. Introduce constraint for each var:

 $\mu \
ightarrow \sum_e c_e d_e = 1. \ f_p \
ightarrow orall p \in P_i \ d_i - \sum_{e \in p} d_e \leq 0.$

 $\min \mu$

$$\forall e : \mu c_e - \sum_{p \ni e} f_p \ge 0 \qquad d_e$$

$$\forall i : \sum_{p \in P_i} f_p = D_i \qquad d_i$$

 $f_p \ge 0$ Introduce variable for each constraint. Introduce constraint for each var:

$$\mu \
ightarrow \sum_e c_e d_e = 1. \ f_{\rho} \
ightarrow orall p \in P_i \ d_i - \sum_{e \in \rho} d_e \leq 0.$$

 $\min \mu$

$$\forall e : \mu c_e - \sum_{p \ni e} f_p \ge 0 \qquad d_e$$
$$\forall i : \sum_{p \in P_i} f_p = D_i \qquad d_i$$

 $f_p \ge 0$ Introduce variable for each constraint. Introduce constraint for each var: $\mu \rightarrow \sum_e c_e d_e = 1$. $f_p \rightarrow \forall p \in P_i \ d_i - \sum_{e \in p} d_e \le 0$. Objective: right hand sides.

 $\min \mu$

$$\forall e : \mu c_e - \sum_{p \ni e} f_p \ge 0 \qquad d_e \\ \forall i : \sum_{p \in P_i} f_p = D_i \qquad d_i$$

 $f_p \ge 0$ Introduce variable for each constraint. Introduce constraint for each var: $\mu \rightarrow \sum_e c_e d_e = 1$. $f_p \rightarrow \forall p \in P_i \ d_i - \sum_{e \in p} d_e \le 0$. Objective: right hand sides. max $\sum_i D_i d_i$

$$\max \sum_{i} D_{i}d_{i}$$
 $orall p \in P_{i}: d_{i} \leq \sum_{e \in p} d(e)$
 $\sum_{e} c_{e}d_{e} = 1$
$\min \mu$

$$\forall e : \mu c_e - \sum_{p \ni e} f_p \ge 0 \qquad d_e$$
$$\forall i : \sum_{p \in P_i} f_p = D_i \qquad d_i$$

 $f_p \ge 0$ Introduce variable for each constraint. Introduce constraint for each var: $\mu \rightarrow \sum_e c_e d_e = 1$. $f_p \rightarrow \forall p \in P_i \ d_i - \sum_{e \in p} d_e \le 0$. Objective: right hand sides. max $\sum_i D_i d_i$

$$\max \sum_i D_i d_i$$
 $orall p \in P_i : d_i \leq \sum_{e \in p} d(e)$

 d_i - shortest s_i, t_i path $\lim_e g_e d_e = 1$

 $\min \mu$

$$\forall e : \mu c_e - \sum_{p \ni e} f_p \ge 0 \qquad d_e$$
$$\forall i : \sum_{p \in P_i} f_p = D_i \qquad d_i$$

 $f_p \ge 0$ Introduce variable for each constraint. Introduce constraint for each var: $\mu \rightarrow \sum_e c_e d_e = 1$. $f_p \rightarrow \forall p \in P_i \ d_i - \sum_{e \in p} d_e \le 0$. Objective: right hand sides. max $\sum_i D_i d_i$

$$egin{aligned} \max \sum_i D_i d_i \ orall oldsymbol{p} \in oldsymbol{P}_i : oldsymbol{d}_i \leq \sum_{oldsymbol{e} \in oldsymbol{p}} d(oldsymbol{e}) \end{aligned}$$

 d_i - shortest s_i, t_i path $\lim_e g_e d_e$ Toll problem!

 $\min \mu$

$$\forall e : \mu c_e - \sum_{p \ni e} f_p \ge 0 \qquad d_e$$
$$\forall i : \sum_{p \in P_i} f_p = D_i \qquad d_i$$

 $f_p \ge 0$ Introduce variable for each constraint. Introduce constraint for each var: $\mu \rightarrow \sum_e c_e d_e = 1$. $f_p \rightarrow \forall p \in P_i \ d_i - \sum_{e \in p} d_e \le 0$. Objective: right hand sides. max $\sum_i D_i d_i$

$$\max \sum_{i} D_{i} d_{i}$$

 $orall p \in P_{i} : d_{i} \leq \sum_{e \in p} d(e)$

 $\min \mu$

$$\forall e : \mu c_e - \sum_{p \ni e} f_p \ge 0 \qquad d_e$$
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$$ext{max} \sum_i D_i d_i$$
 $orall p \in P_i: d_i \leq \sum_{e \in p} d(e)$

d_i - shortest s_i, t_i path langer de Tell¹problem! Weak duality: toll lower bounds routing. Strong Duality.

 $\min \mu$

$$\forall e : \mu c_e - \sum_{p \ni e} f_p \ge 0 \qquad d_e$$
$$\forall i : \sum_{p \in P_i} f_p = D_i \qquad d_i$$

 $f_p \ge 0$ Introduce variable for each constraint. Introduce constraint for each var: $\mu \rightarrow \sum_e c_e d_e = 1$. $f_p \rightarrow \forall p \in P_i \ d_i - \sum_{e \in p} d_e \le 0$. Objective: right hand sides. max $\sum_i D_i d_i$

$$ext{max} \sum_i D_i d_i$$
 $orall oldsymbol{p} \in oldsymbol{P}_i : oldsymbol{d}_i \leq \sum_{oldsymbol{e} \in oldsymbol{p}} oldsymbol{d}(oldsymbol{e})$

d_i - shortest s_i, t_i path length lengt

 $\min \mu$

$$\forall e : \mu c_e - \sum_{p \ni e} f_p \ge 0 \qquad d_e$$
$$\forall i : \sum_{p \in P_i} f_p = D_i \qquad d_i$$

 $f_p \ge 0$ Introduce variable for each constraint. Introduce constraint for each var: $\mu \rightarrow \sum_e c_e d_e = 1$. $f_p \rightarrow \forall p \in P_i \ d_i - \sum_{e \in p} d_e \le 0$. Objective: right hand sides. max $\sum_i D_i d_i$

$$ext{max}\sum_{i} D_i d_i$$
 $orall p \in P_i$: $d_i \leq \sum_{e \in p} d(e)$

d_i - shortest *s_i*, *t_i* path long *d* = Tell¹ problem! Weak duality: toll lower bounds routing. Strong Duality. Tight lower bound. First lecture.

 $\min \mu$

$$\forall e : \mu c_e - \sum_{p \ni e} f_p \ge 0 \qquad d_e$$
$$\forall i : \sum_{p \in P_i} f_p = D_i \qquad d_i$$

 $f_p \ge 0$ Introduce variable for each constraint. Introduce constraint for each var: $\mu \rightarrow \sum_e c_e d_e = 1$. $f_p \rightarrow \forall p \in P_i \ d_i - \sum_{e \in p} d_e \le 0$. Objective: right hand sides. max $\sum_i D_i d_i$

$$ext{max}\sum_{i} D_i d_i$$
 $orall oldsymbol{p} \in oldsymbol{P}_i$: $oldsymbol{d}_i \leq \sum_{oldsymbol{e} \in oldsymbol{p}} oldsymbol{d}(oldsymbol{e})$

d_i - shortest *s_i*, *t_i* path long *d* and and *d* an

 $\min \mu$

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$$ext{max}\sum_{i} D_i d_i$$
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d_i - shortest *s_i*, *t_i* path long *d_e*Tell¹problem! Weak duality: toll lower bounds routing. Strong Duality. Tight lower bound. First lecture. Or Experts. Complementary Slackness:

 $\min \mu$

$$\forall e : \mu c_e - \sum_{p \ni e} f_p \ge 0 \qquad d_e$$
$$\forall i : \sum_{p \in P_i} f_p = D_i \qquad d_i$$

 $f_p \ge 0$ Introduce variable for each constraint. Introduce constraint for each var: $\mu \rightarrow \sum_e c_e d_e = 1$. $f_p \rightarrow \forall p \in P_i \ d_i - \sum_{e \in p} d_e \le 0$. Objective: right hand sides. max $\sum_i D_i d_i$

$$ext{max}\sum_{i} D_i d_i$$
 $orall oldsymbol{p} \in oldsymbol{P}_i$: $oldsymbol{d}_i \leq \sum_{oldsymbol{e} \in oldsymbol{p}} oldsymbol{d}(oldsymbol{e})$

d_i - shortest *s_i*, *t_i* path long *d_e*Tell¹problem! Weak duality: toll lower bounds routing. Strong Duality. Tight lower bound. First lecture. Or Experts. Complementary Slackness: only route on shortest paths

 $\min \mu$

$$\forall e : \mu c_e - \sum_{p \ni e} f_p \ge 0 \qquad d_e$$
$$\forall i : \sum_{p \in P_i} f_p = D_i \qquad d_i$$

 $f_p \ge 0$ Introduce variable for each constraint. Introduce constraint for each var: $\mu \rightarrow \sum_e c_e d_e = 1$. $f_p \rightarrow \forall p \in P_i \ d_i - \sum_{e \in p} d_e \le 0$. Objective: right hand sides. max $\sum_i D_i d_i$

$$ext{max}\sum_{i} D_i d_i$$
 $orall oldsymbol{p} \in oldsymbol{P}_i$: $oldsymbol{d}_i \leq \sum_{oldsymbol{e} \in oldsymbol{p}} oldsymbol{d}(oldsymbol{e})$

Multicommodity flow.

 $\min \mu$

$$orall e: \mu c_e - \sum_{p \ni e} f_p \ge 0$$

 $orall i: \sum_{p \in P_i} f_p = d_i$
 $f_p \ge 0$

Multicommodity flow.

 $\min \mu$ $\forall e : \mu c_e - \sum_{p \ni e} f_p \ge 0$ $\forall i : \sum_{p \in P_i} f_p = d_i$ $f_p \ge 0$

Dual is.

$$\max \sum_i D_i d_i$$

 $orall p \in P_i : d_i \leq \sum_{e \in p} d(e)$

Multicommodity flow.

 $\min \mu$ $\forall \boldsymbol{e} : \mu \boldsymbol{c}_{\boldsymbol{e}} - \sum_{\boldsymbol{p} \ni \boldsymbol{e}} f_{\boldsymbol{p}} \ge 0$ $\forall \boldsymbol{i} : \sum_{\boldsymbol{p} \in \boldsymbol{P}_{i}} f_{\boldsymbol{p}} = \boldsymbol{d}_{i}$ $f_{\boldsymbol{p}} \ge 0$

Dual is.

$$\max \sum_{i} D_i d_i$$
 $orall p \in P_i: d_i \leq \sum_{e \in p} d(e)$

Exponential sized programs?

Multicommodity flow.

 $\min \mu$ $\forall \boldsymbol{e} : \boldsymbol{\mu} \boldsymbol{c}_{\boldsymbol{e}} - \sum_{\boldsymbol{\rho} \ni \boldsymbol{e}} \boldsymbol{f}_{\boldsymbol{\rho}} \geq \boldsymbol{0}$

$$\forall i : \sum_{p \in P_i} f_p = d_i$$
$$f_p \ge 0$$

Dual is.

$$\max \sum_{i} D_i d_i$$
 $orall p \in P_i: d_i \leq \sum_{e \in p} d(e)$

Exponential sized programs? Answer 1:

Multicommodity flow.

 $\min \mu$ $\forall \boldsymbol{e} : \mu \boldsymbol{c}_{\boldsymbol{e}} - \sum_{\boldsymbol{p} \ni \boldsymbol{e}} f_{\boldsymbol{p}} \ge 0$ $\forall \boldsymbol{i} : \sum_{\boldsymbol{p} \in \boldsymbol{P}_{i}} f_{\boldsymbol{p}} = \boldsymbol{d}_{i}$ $f_{\boldsymbol{p}} \ge 0$

Dual is.

$$\max \sum_{i} D_i d_i$$
 $orall p \in P_i: d_i \leq \sum_{e \in p} d(e)$

Exponential sized programs? Answer 1: We solved anyway!

Multicommodity flow.

 $\min \mu$ $\forall e: \mu c_e - \sum_{p \ni e} f_p \ge 0$

$$\forall i : \sum_{p \in P_i} f_p = d_i$$
$$f_p \ge 0$$

Dual is.

$$\max \sum_{i} D_i d_i$$
 $orall p \in P_i: d_i \leq \sum_{e \in p} d(e)$

Exponential sized programs? Answer 1: We solved anyway! Answer 2:

Multicommodity flow.

 $\min \mu$ $\forall \boldsymbol{e} : \mu \boldsymbol{c}_{\boldsymbol{e}} - \sum_{\boldsymbol{p} \ni \boldsymbol{e}} f_{\boldsymbol{p}} \ge 0$ $\forall \boldsymbol{i} : \sum_{\boldsymbol{p} \in \boldsymbol{P}_{i}} f_{\boldsymbol{p}} = \boldsymbol{d}_{i}$ $f_{\boldsymbol{p}} \ge 0$

Dual is.

$$\max \sum_{i} D_i d_i$$
 $orall p \in P_i: d_i \leq \sum_{e \in p} d(e)$

Exponential sized programs? Answer 1: We solved anyway! Answer 2: Ellipsoid algorithm.

Multicommodity flow.

 $\min \mu$

$$\forall \boldsymbol{e} : \boldsymbol{\mu} \boldsymbol{c}_{\boldsymbol{e}} - \sum_{\boldsymbol{p} \ge \boldsymbol{e}} f_{\boldsymbol{p}} \ge \boldsymbol{0}$$
$$\forall \boldsymbol{i} : \sum_{\boldsymbol{p} \in \boldsymbol{P}_{i}} f_{\boldsymbol{p}} = \boldsymbol{d}_{i}$$
$$f_{\boldsymbol{p}} \ge \boldsymbol{0}$$

Dual is.

$$\max \sum_{i} D_i d_i$$
 $orall p \in P_i: d_i \leq \sum_{e \in p} d(e)$

Exponential sized programs?

Answer 1: We solved anyway!

Answer 2: Ellipsoid algorithm.

Find violated constraint \rightarrow poly time algorithm.

Multicommodity flow.

 $\min \mu$

$$\forall \boldsymbol{e} : \boldsymbol{\mu} \boldsymbol{c}_{\boldsymbol{e}} - \sum_{\boldsymbol{p} \ge \boldsymbol{e}} f_{\boldsymbol{p}} \ge \boldsymbol{0}$$
$$\forall \boldsymbol{i} : \sum_{\boldsymbol{p} \in \boldsymbol{P}_{i}} f_{\boldsymbol{p}} = \boldsymbol{d}_{i}$$
$$f_{\boldsymbol{p}} \ge \boldsymbol{0}$$

Dual is.

 $\max \sum_{i} D_{i} d_{i}$ $orall p \in P_{i}: d_{i} \leq \sum_{e \in p} d(e)$

Exponential sized programs?

Answer 1: We solved anyway!

Answer 2: Ellipsoid algorithm.

Find violated constraint \rightarrow poly time algorithm.

Answer 3: there is polynomial sized formulation.

Multicommodity flow.

 $\min \mu$

$$\forall \boldsymbol{e} : \boldsymbol{\mu} \boldsymbol{c}_{\boldsymbol{e}} - \sum_{\boldsymbol{p} \ge \boldsymbol{e}} f_{\boldsymbol{p}} \ge \boldsymbol{0}$$
$$\forall \boldsymbol{i} : \sum_{\boldsymbol{p} \in \boldsymbol{P}_{i}} f_{\boldsymbol{p}} = \boldsymbol{d}_{i}$$
$$f_{\boldsymbol{p}} \ge \boldsymbol{0}$$

Dual is.

 $\max \sum_{i} D_{i} d_{i}$ $orall p \in P_{i}: d_{i} \leq \sum_{e \in p} d(e)$

Exponential sized programs?

Answer 1: We solved anyway!

Answer 2: Ellipsoid algorithm.

Find violated constraint \rightarrow poly time algorithm.

Answer 3: there is polynomial sized formulation. Question: what is it?

 $\begin{array}{c} \circ \\ \circ \\ \circ \\ \circ \\ \end{array}$





































Blue constraints redundant.









Blue constraints tight.



Blue constraints tight.



Blue constraints tight.




V



V



V



V



V



V



V









V



Sum: x + 2z + y.







See you on Thursday.