

LP Duality and Optimum Matching

Seth Pettie

EECS 477

Aaron Bernstein: Negative-Weight Single-Source Shortest Paths in Near-linear Time

Aaron Berstein

Rutgers University

WHERE: 3725 Beyster Building



WHEN: Friday, November 4, 2022 @ 3:00 pm - 4:00 pm

This event is free and open to the public



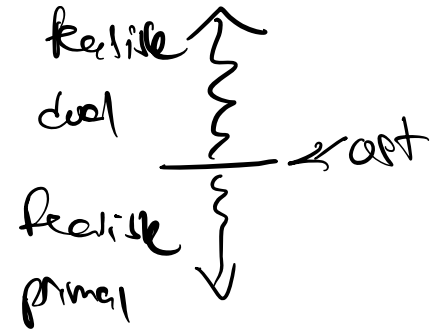
WEB: [Event Website](#)

SHARE: [f](#) [t](#) [in](#) [✉](#)

Zoom link

LPs and Duality.

- Primal LP: $x \in \mathbb{R}^n, A \in \mathbb{R}^{m \times n}, b \in \mathbb{R}^m, c \in \mathbb{R}^n$
 - **Maximize** $c^T x$ ($\sum_{i=1}^n c_i x_i$)
 - Subject to: $Ax \leq b, x \geq 0$.
- Dual LP: $y \in \mathbb{R}^m$
 - **Minimize** $y^T b$
 - Subject to: $c^T \leq y^T A, y \geq 0$.
- Assume Primal is feasible and bounded (objective not ∞).
- **Weak Duality Theorem**. If x, y are feasible, $c^T x \leq y^T b$.
- **Strong Duality Theorem**. If x^*, y^* are opt., $c^T x^* = y^{*T} b$



Duality Cheatsheet

Primal

Dual

<i>Maximize</i>	<i>Minimize</i>
i th constraint \leq	$y_i \geq 0$
i th constraint \geq	$y_i \leq 0$
i th constraint $=$	y_i unconstrained
$x_j \geq 0$	j th constraint \geq
$x_j \leq 0$	j th constraint \leq
x_j unconstrained	j th constraint $=$

If the primal is in
"standard" format...

... the dual is also in
"standard" format...

LPs and Duality.

- **Strong Duality Theorem**. If x^*, y^* are opt., $c^T x^* = y^{*T} b$
- An implication of Strong Duality.
 - Suppose j^{th} constraint is “loose” $(\sum_{i=1}^n a_{i,j} x_i^*) < b_j$.
 - What does this tell us about y_j^* ? It must be 0.
- **Complementary Slackness Conditions:**
 - If $y_j \neq 0$ then the j^{th} constraint in the primal is *tight* (=).
 - If the j^{th} constraint in the primal is *loose* then $y_j = 0$.
 - If $x_i \neq 0$ then the i^{th} constraint in the dual is *tight* (=).
 - If the i^{th} constraint in the dual is *loose* then $x_i = 0$.

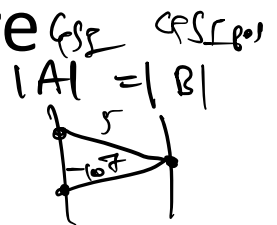
Max Weight Perfect Matching

- Given **bipartite** graph $G = (A \cup B, E)$, $E = A \times B$, and **weight function** $w: E \rightarrow \mathbb{R}$, find the maximum weight perfect matching.

– A **matching** is a set of edges that share no endpoints.

– A **perfect matching** is one in which all vertices are adjacent to a matching edge.

bipartite graph



$A \times B$

LP max

$$x(u,v) = \begin{cases} 0 & (u,v) \in \text{matching} \\ 1 & \notin \text{matching} \end{cases}$$

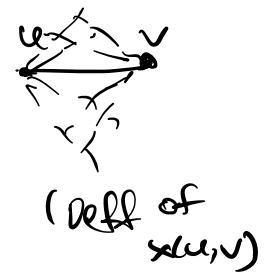
$$\max \sum_{(u,v)} x(u,v) \cdot w(u,v)$$

$$\forall v \sum_u x(u,v) = 1 \quad (\text{only 1 edge selected})$$

$$x \geq 0$$

Dual

$$\text{Min} \sum_v 1 \cdot y(v)$$



$\longrightarrow y(v)$ unconstrained

$$y(u) + y(v) \geq w(u,v)$$

Complementary Slackness

$$x(u,v) > 0 \Rightarrow y(u) + y(v) = w(u,v)$$

Max Weight Perfect Matching

- Given **bipartite** graph $G = (A \cup B, E)$, $E = A \times B$, and **weight function** $w: E \rightarrow \mathbb{R}$, find the maximum weight perfect matching.
- LP:
 - **Maximize** $\sum_{e \in E} w(e)x(e)$ subject to:
 - $\forall v \in V: \sum_{u \in V} x(v, u) = 1$
 - $x(u, v) \geq 0$
- Dual LP:
 - **Minimize** $\sum_{v \in V} y(v)$ subject to:
 - $\forall (u, v) \in E: y(u) + y(v) \geq w(u, v)$
 - y unconstrained.

The Hungarian Algorithm

THE HUNGARIAN METHOD FOR THE ASSIGNMENT PROBLEM¹

H. W. Kuhn
Bryn Mawr College

1955

Kőnig, Dénes (1931), "Gráfok és mátrixok", *Matematikai és Fizikai Lapok*, **38**: 116–119.
Egerváry, Jenő (1931), "Matrixok kombinatorius tulajdonságairól", *Matematikai és Fizikai Lapok*, **38**: 16–28

DE AEQUATIONUM DIFFERENTIALIUM SYSTEMATE NON NORMALI AD FORMAM NORMALEM REVOCANDO

AUCTORE

C. G. J. JACOBI,
PROF. ORD. MATH. REGIOM.

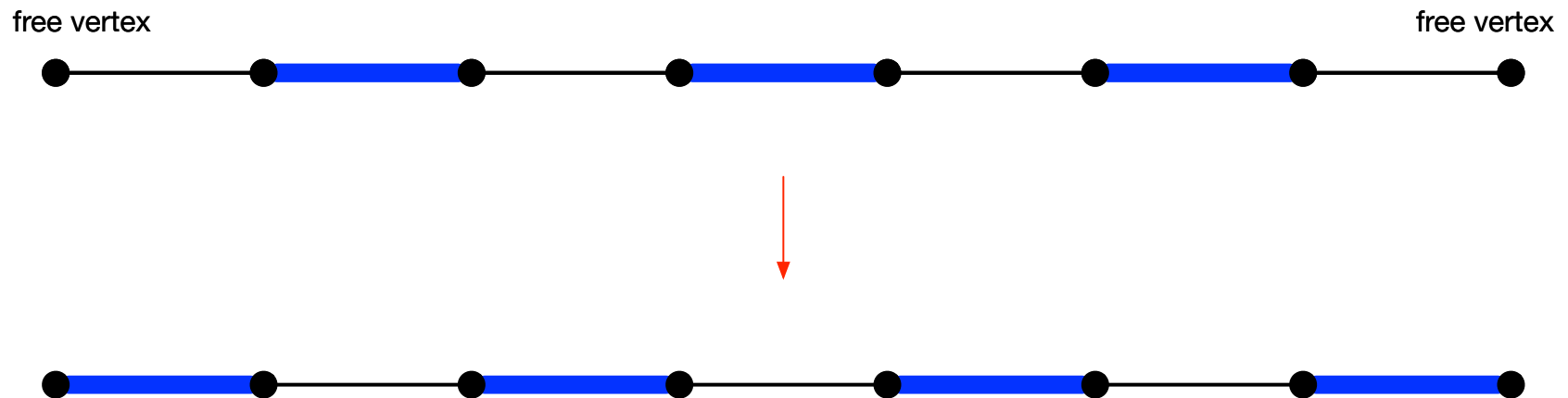
Written 1836, left on a shelf,
published posthumously in 1890,
discovered & translated in 2006.

The Hungarian Algorithm

- A *primal-dual* algorithm:
 - Maintain a *feasible dual solution* at all times.
 - Maintain a primal solution, that is *usually infeasible*.
 - In each step do one of two things:
 - Do a “*dual adjustment*” — reduce the dual objective.
 - Do a “*primal adjustment*” — make the primal *less infeasible*.
 - Halt when you have a feasible dual & feasible primal with the same objective. By ***Strong Duality***, they are both optimal.

Augmenting Paths

- An *augmenting path* is one that
 - Starts and ends at free (unmatched) vertices
 - Alternates between unmatched & matched edges



- **Observation.** If M is a matching and P an augmenting path wrt M , then $M' = (M \cup P) - (M \cap P)$ is also a matching and $|M'| = |M| + 1$.

M matching $e \in M \Leftrightarrow x(e) = 1$

If invariant \star holds and M is perfect, then M is a max weight perfect matching.

$$w(M) = \sum_{e \in M} w(e) \quad (\text{count every } y \text{ value exactly once})$$

$$= \sum_{u \in V} y(u) \quad (\text{"tightness" from } \odot)$$

$$\geq \sum_{e \in M'} w(e) = w(M')$$

Invariant \star

Maintain (M, y)

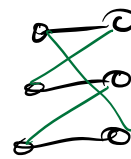
↑
matching

1) $(u, v) \in M \Rightarrow$

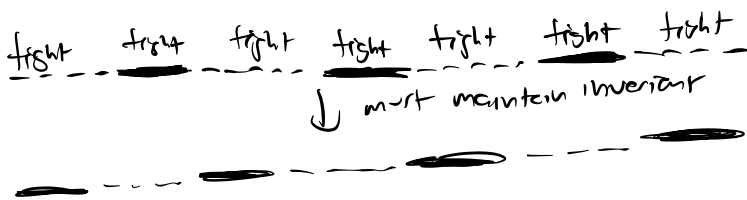
$$y(u) + y(v) = w(u, v)$$

2) $(u, v) \notin M \Rightarrow$

$$y(u) + y(v) \geq w(u, v)$$



M' ← some other matching



Hungarian (G, w) :

Initialize $M = \emptyset$

$u \in A \quad y(u) = \max_z w(u, z)$

$u \in B \quad y(u) = 0$

Repeat

F_A, F_B unmatched vertices in A, B .

$$E_+ = \{(u, v) \mid y(u) + y(v) = w(u, v)\}$$

$V_+ =$ all vertices reachable from F_A by alt. path

if $(V_+ \cap F_B) \neq \emptyset \Rightarrow$ aug path P . Set

else

$$M = (M \cup P) / (M \cap P)$$

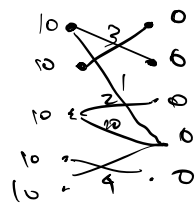
"primal only" →
"dual only"

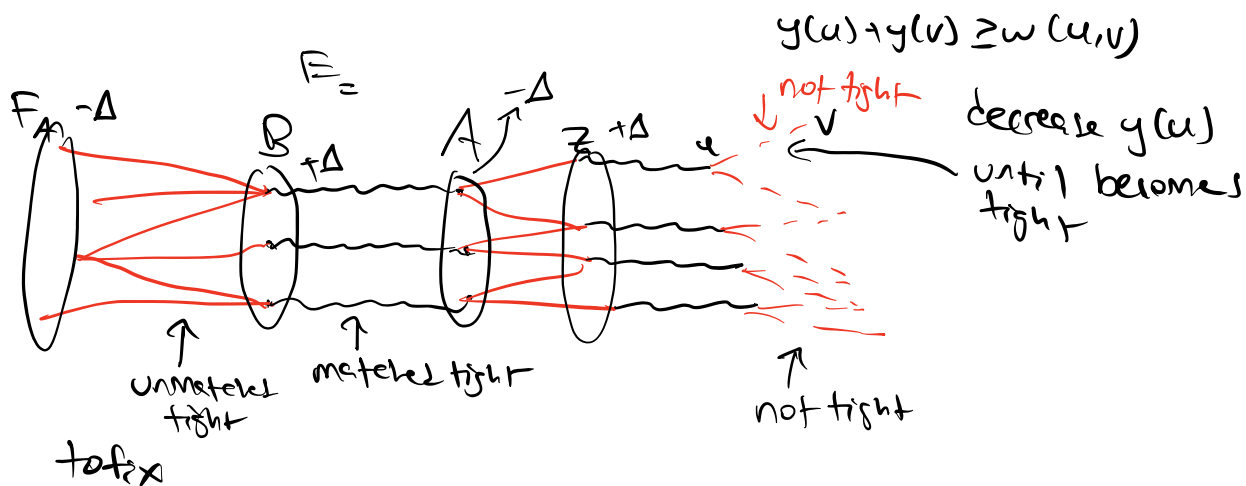
$O(n^2 \cdot m)$

1 Dijkstra's

$\leq n^2$ times

$\leq |A|$ times





- everything in A , $-\Delta$ \rightarrow all y values.
 - " B , $+\Delta$

$$\Delta = \min_{\substack{u \in (V \setminus A) \\ v \in V =}} y(u) + y(v) - w(u, v)$$

Bipartite graph assumption

$$y(u) -= \Delta \text{ if } u \in A \cap V =$$

$$y(v) += \Delta \text{ if } v \in B \cap V =$$

means this algorithm finishes. Not true for a general graph.

$\frac{1}{2}$ $\frac{1}{2}$ fractional soln for general graph.

Edmond's LP:
max _____

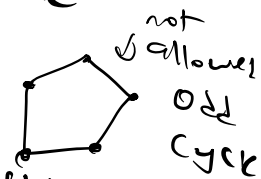
Still polynomial time

$$\sum x(u, v) = 1$$

$\forall s$

$$|S| \text{ odd, } \sum_{(u, v) \in S} x(u, v) \leq \lfloor \frac{|S|}{2} \rfloor$$

2^n constraints



Extra