

Primal/dual Method for weighted Matching

We see a different Primal/dual based algorithm, called Hungarian method, for weighted bipartite matching.

$$\begin{array}{ll} \max \sum_e w_e x_e & \min \sum_{v \in A \cup B} y_v \\ \forall a \in A \quad \sum_{a \in e} x_e \leq 1 & (P) \quad \forall e = ab \quad y_a + y_b \geq w_e \quad (D) \\ \forall b \in B \quad \sum_{b \in e} x_e \leq 1 & \\ x_e \geq 0 & y_v \geq 0 \end{array}$$

By duality, if $\vec{x}$ and $\vec{y}$ are feasible solutions to (P) and (D) then $w(y) \geq w(M)$ (where y is the vertex cover & M is the matching defined by $\vec{x}$ & $\vec{y}$)

The P/D algorithm starts with a trivial Primal solution i.e. $M = \emptyset$ (so $x_e = 0$) and a trivial dual solution

$$\begin{array}{ll} \forall a \in A: & y_a = \max_{ab \in E} w(ab) \\ \forall b \in B: & y_b = 0 \end{array}$$

- At each iteration of the algorithm, we build an equality graph G_y as defined below

$$G_y = (A \cup B, E_y) : \quad ab \in E_y \iff y_a + y_b = w_e$$

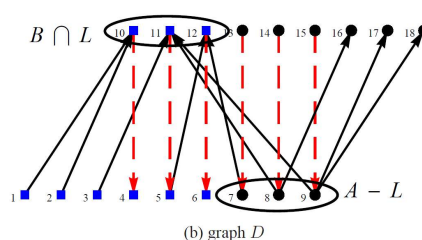
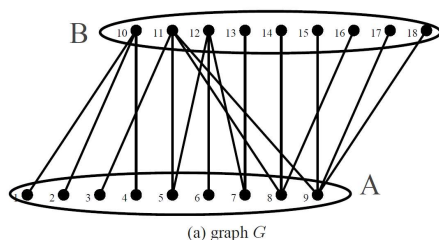
let us call $(y_a + y_b - w_e)$ as the excess of ab .

Observation: if M is a perfect matching in G_y then

$$W(M) = \sum_{a \in A} y_a + \sum_{b \in B} y_b \quad \text{and } M \text{ is an optimum matching}$$

- So the goal of the algorithm will be to find a maximum in equality graph G_y .

- So the goal of the algorithm will be to find a matching in equality graph G_y .
- We update y so that more edges go tight to be added to G_y (until it has a perfect matching) while ensuring y remains a vertex cover.
- Suppose we are at some iteration of the algorithm and M is a matching in G_y but not perfect
- We construct the digraph D as before
 - $e \in M$: directed B to A
 - $e \notin M$: directed A to B



- let L be the set of nodes accessible from any exposed node in A .
- Recall $C^* = (A - L) \cup (B \cap L)$ is a vertex-cover
 - $\rightarrow$ $\nexists$ no edge between $A \cap L$ and $B - L$
- But G was a complete graph; so there are edges in G but not in G_y
 - $\rightarrow$ those edges have positive excess.
- We update y s.t. one of these edges goes tight
 - let $\epsilon = \min \{ y_a + y_b - w_e \text{ s.t. } a \in A \cap L, b \in B - L \}$
 - be the min excess of all such edges (these are those that can be added to G_y)

those that can be added to G_y

- update all y 's by ε :

$$y_a = \begin{cases} y_a & a \in A-L \\ y_a - \varepsilon & a \in A \cap L \end{cases} \quad y_b = \begin{cases} y_b & b \in B-L \\ y_b + \varepsilon & b \in B \cap L \end{cases}$$

Note: all edges in G_y remain tight.

No constraint goes violated.

and at least one edge between $A \cap L$ & $B-L$ goes tight $\rightarrow$ added to G_y .

- repeat this until G_y has a perfect matching or there is no edges left between $A \cap L$ & $B-L$.

- The latter can happen only if both $A \cap L$ & $B-L = \emptyset$

$\rightarrow$ We eventually find a perfect matching in G_y

Max-Weight bip. (perfect) matching: P/D

For each $a \in A$ let $y_a = \max_{b \in B} w(ab)$

For each $b \in B$ let $y_b = 0$

Build G_y and let M be a max matching in G_y

Construct D

repeat

let L be the nodes reachable from an exposed node of A

let $\varepsilon = \min \{ y_a + y_b - w(ab) \mid a \in A \cap L, b \in B-L \}$

Decrease y_a for all $a \in A \cap L$ by ε

increase y_b " " $b \in B-L$ " "

Add tight edges to G_y and recompute M

until M is a perfect matching

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Runtime : At each iteration one edge is added to G_y $\rightarrow O(n^2)$ iterations. To compute L & recompute M we spend at most $O(n^2) \rightarrow$ total time $O(n^4)$

- with more careful analysis the time can be improved to $O(n^3)$.

Primal / Dual for Bipartite Matching

- We have already seen a number of algorithms for bip. matching

- One can design algorithm using LP and dual LP that avoid solving the LP (are combinatorial & fast).

- Hungarian method is one such alg.

- We see another method that finds a $(1-\epsilon)$ -approximate matching in time $O(m/\epsilon)$; it also finds an approximate vertex cover.

- Suppose we have a bipartite graph $G = (A \cup B, E)$

$$\max \sum_{e \in E} x_e$$

$$\min \sum_{u \in A \cup B} y_u$$

$$\forall a \in A: \sum_{a \in e} x_e \leq 1$$

$$\forall a, b \in E: y_a + y_b \geq 1$$

$$\forall b \in B: \sum_{b \in e} x_e \leq 1$$

$$y_u \geq 0$$

$$x_e \geq 0$$

- The algorithm maintains a primal feasible solution (i.e. a matching) and a dual infeasible solution.

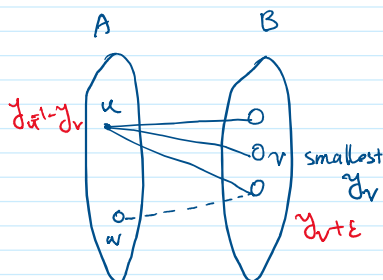
- At each iteration, we try to improve optimality of the primal and feasibility of the dual.

(1-ε)-approximate Primal/Dual Matching

- Start from $M = \emptyset$ and $y_u = 0$ (for all u)
- let $U = A$ be the set of unmatched node of A
- while $U \neq \emptyset$ do
 - * pick a node $a \in U$ and remove from U
 - * let $b \in B$ be a neighbor of a with smallest y_b
 - * if $y_b = 1$ skip to next iteration
 - * else let a' be the matched neighbor of b if it exists
 - * Add ab to M (a' becomes unmatched and add to U)
 - * update $y_b \leftarrow y_b + \epsilon$, $y_a = 1 - y_b$, $y_{a'} = 0$
- output M as a matching and $(\frac{y}{1-\epsilon})$ as a V.C

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Fact 1: if $ab \in M$ then $y_a = 1 - y_b$, if a is not matched $y_a = 0$

Fact 2: $\forall b \in B : y_b > 0 \iff b$ is matched

proof: when b is matched for the first time y_v increases from 0 and it remains matched.

Fact 3: For any $ab \in M : y_a + y_b = 1$ (so dual const tight)

Lemma 1: At any iteration: $|M| = \sum_{a \in A} y_a + \sum_{b \in B} y_b$

proof:

$$\sum_{a \in A} y_a + \sum_{b \in B} y_b = \sum_{ab \in M} \underbrace{(y_a + y_b)}_1 + \underbrace{\sum_{a \text{ not matched}} y_a}_{0 \text{ by Fact 1}} + \underbrace{\sum_{b \text{ not matched}} y_b}_{0 \text{ by Fact 2}} = |M|$$

1. 2. 3. 4. 5. 6. 7. 8. 9. 10.

Lemma 2: $y/(1-\epsilon)$ will be a dual feasible solution

Corollary: The algorithm returns a matching M that is $(1-\epsilon)$ -approximate and $y/(1-\epsilon)$ is a fractional vertex cover.

proof: Note that $\text{opt}_D \geq \text{opt}_P$ P/D
 $\geq |M|$
 $\geq \sum_{a \in A} y_a + \sum_{b \in B} y_b$ lemma 1
 $\geq (1-\epsilon) \text{opt}_D$ lemma 2
 $\geq (1-\epsilon) \text{opt}_{LP}$

Lemma 2: $y/(1-\epsilon)$ will be a dual feasible solution

proof: We show $\forall ab \in E: y_a + y_b \geq 1-\epsilon$

- if a is not matched: $\rightarrow$ All its neighbors b have $y_b = 1$

$$\rightarrow y_a + y_b = 1$$

- if $ab \in M$: $y_a = 1 - y_b \rightarrow y_a + y_b = 1$

- if $ac \in M$ for some other node c :

when ac was chosen, y_c was smallest $\rightarrow y_c \leq y_b$
 at the time, so after ac was added y_c got $+\epsilon$; so $y_c \leq y_b + \epsilon$ at that time.

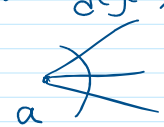
Since c remained assigned to a , y_c does not increase, y_b can only increase $\rightarrow$

at the end: $y_c \leq y_b + \epsilon$ but $\underbrace{y_a + y_c = 1}_{\text{by Fact 1}}$

$$\rightarrow y_a + y_b \geq 1 - \epsilon.$$

Time Complexity: Each iteration of while loop the y_b value of some $b \in B$ increases by ϵ , and at most n nodes $\rightarrow O(n/\epsilon)$ iterations.

Add/delete to U takes $O(1)$ and $O(n)$ to check y_b for neighbours of a to find $\min y_b$ $\rightarrow$ total time $O(n^2/\epsilon) \rightarrow O(m/\epsilon)$



Better time: For each $a \in A$ we keep a list $D(a)$ of size $\deg(a)$ and a threshold value $d(a)$ initially $= 0$.

Initially each $d(a)$ has all neighbours of a and to find $\min_{b \in N(a)} y_b$, we check nodes in $D(a)$ if $y_b > d(a)$ remove b from $D(a)$; else return b ; when $D(a)$ becomes empty we increase threshold $d(a)$ by ϵ and add back all neighbours of a to $D(a)$.

Note that each time we pick $b \in D(a)$ it has the smallest y_b value among $N(a)$.

Also, $D(a)$ resets to $N(a)$ at most $\frac{1}{\epsilon}$ times

$\rightarrow$ time for each node $a \in A$: $O(\deg(a)/\epsilon)$

$\rightarrow$ total time of algorithm: $O(m/\epsilon)$.

Change to Exact matching Algorithm

if we choose $\epsilon < \frac{1}{n}$, say $\epsilon = \frac{1}{n+1}$, then

we find a matching M with $|M| \geq (1 - \frac{1}{n+1}) \text{opt}$

Since $\text{opt} \leq n$. Since opt is $> \text{opt} - 1$
always integer this implies $|M| = \text{opt}$ and time
complexity is $O(mn)$.

Faster exact Algorithm:

- Phase 1: run the $(1-\epsilon)$ -approx alg. with $\epsilon = \frac{1}{\sqrt{n}}$
to obtain a matching of size $(1 - \frac{1}{\sqrt{n}})\text{opt} \geq \text{opt} - \sqrt{n}$
in time $O(m\sqrt{n})$.
 - Phase 2: run augmenting path algorithm at most
 $\sqrt{n}$ times to find a max matching in time
 $O(m\sqrt{n})$.
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