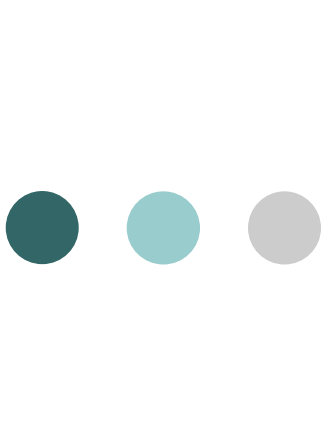

Matching and Covering

Matchings

- A *matching* of size *k* in a graph G is a set of *k* pairwise disjoint edges
  - The vertices belonging to the edges of a matching are *saturated* by the matching; the others are *unsaturated*
  - If a matching saturates every vertex of G, then it is a *perfect matching* or 1-*factor*

Alternating Paths

- Given a matching M, an *M-alternating path* is a path that alternates between the edges in M and the edges not in M
  - An M-alternating path P that begins and ends at unsaturated vertices is an *M-augmenting path*
  - Replacing $M \cap E(P)$ by $E(P) - M$ produces a new matching M' with one more edge than M

Symmetric Difference

- If G and H are graphs with vertex set V, then the *symmetric difference* $G \Delta H$ is the graph with vertex set V whose edges are all those edges appearing in exactly one of G and H

  - If M and M′ are matchings, then
    $M \Delta M' = (M \cup M') - (M \cap M')$

Key Result

- A matching M in a graph G is a *maximum matching* in G if and only if G has no M-augmenting path

Bipartite Matching

- G is a bipartite graph with bipartition *X,Y*
- Does G have a matching that saturates *X?*
  - *A matching of X into Y*

Results…

- [Hall's Theorem: 1935]

  If G is a bipartite graph with bipartition X, Y, then G has a matching of X into Y if and only if $|N(S)| \geq |S|$ for all $S \subseteq X$

- For $k > 0$, every k-regular bipartite graph has a perfect matching

Finding a Matching in Bipartite Graphs: Using Network Flows

- Direct all edges from X to Y
- Add a new vertex s with edges ***to*** all vertices in X
- Add a new vertex t with edges ***from*** all vertices in Y
- Give capacity of 1 to all vertices
- Solve the maximum flow problem from s to t
- The edges in the bipartite graph corresponding to edges in the new graph with flow > 0 is a matching of the bipartite graph

Augmenting Path Algorithm

Input:

- A bipartite graph G with a bipartition *X,Y,* a matching M in G, and the set U of all *M*-unsaturated vertices in *X*

Idea:

- Explore *M*-alternating paths from U, letting $S \subseteq X$ and $T \subseteq Y$ be the sets of vertices reached.
- *Mark* vertices of *S* that have been explored for extending paths.
- For each $x \in (S \cup T) - U$, record the vertex before *x* on some *M*-alternating path from U

Augmenting Path Algorithm

Initialization: Set $S=U$ and $T=\phi$

Iteration:

- If S has no unmarked vertex, stop and report M as a maximum matching.
- Otherwise, select an unmarked $x \in S$.
- Consider each $y \in N(x)$ such that $xy \notin M$. If y is unsaturated, terminate and trace back from y to report an M-augmenting path from U to y. Otherwise, y is matched to some $w \in X$ by M. In this case, include y in T and w in S.
- After exploring all such edges incident to x, mark x and iterate.

Augmenting Path Algorithm

- Repeated application of the Augmenting Path Algorithm to a bipartite graph produces a maximum matching

- The complexity of the algorithm is $O(n^3)$.
  - Since matchings have at most n/2 edges, we apply the augmenting path algorithm at most n/2 times.
  - In each iteration, we search from a vertex of X at most once, before we mark it. Hence each iteration is O(e(G)), which is $O(n^2)$

Vertex Cover & Bipartite Matching

- A *vertex cover* of G is a set *S* of vertices such that *S* contains at least one endpoint of every edge of G
  - The vertices in S *cover* the edges of G

- If G is a bipartite graph, then the maximum size of a matching in G equals the minimum size of a vertex cover of G

[König and Egerváry: 1931]

Computing Minimum Vertex Cover

- Can be computed by the same augmenting path algorithm for computing maximum matching
  - When the augmenting path algorithm finds a maximum matching (no augmenting path found and all vertices in set S are marked), the set $T \cup (X - S)$ gives the minimum vertex cover

Edge Cover

- An *edge cover* of G is a set of edges that cover the vertices of G
  - only graphs without isolated vertices have edge covers
- Minimum edge cover – edge cover of minimum size

Independent Set

- An *independent set* is a set of vertices such that no two of them are adjacent to each other
- Maximum independent set – independent set of maximum size

Notation…

- We will use the following notation for independence and covering problems

$\alpha(G)$: maximum size of independent set

$\alpha'(G)$: maximum size of matching

$\beta(G)$: minimum size of vertex cover

$\beta'(G)$: minimum size of edge cover

Min-max Theorems

- In a graph G, $S \subseteq V(G)$ is an independent set if and only if S' is a vertex cover, and hence

  $\alpha(G) + \beta(G) = n(G)$

- If G has no isolated vertices, then

  $\alpha'(G) + \beta'(G) = n(G)$

- If G is a bipartite graph with no isolated vertices, then $\alpha(G) = \beta'(G)$
  - (max independent set = min edge cover)

Weighted Bipartite Matching

- A *transversal* of an *n X n* matrix *A* consists of *n* positions – one in each row and each column
  - Finding a transversal of *A* with maximum sum is the *assignment problem*
  - This is the matrix formulation of the *maximum weighted matching problem,* where *A* is the matrix of weights w_{ij} assigned to the edges $x_i y_j$ of $K_{n,n}$ and we seek a perfect matching *M* with maximum total weight *w(M)*

Minimum Weighted Cover

- Given the weights $\{w_{ij}\}$, a *weighted cover* is a choice of labels $\{u_i\}$ and $\{v_j\}$ such that

$$u_i + v_j \geq w_{ij} \quad \text{for all } i, j$$

- The *cost c(u,v)* of a cover *u,v* is $\Sigma\, u_i + \Sigma\, v_j$

- The *minimum weighted cover problem* is the problem of finding a cover of minimum cost

Min Cover & Max Matching

- If M is a perfect matching in a weighted bipartite graph G and *u,v* is a cover, then $c(u,v) \geq w(M)$
  - Furthermore, $c(u,v) = w(M)$ if and only if M consists of edges $x_i y_j$ such that $u_i + v_j = w_{ij}$. In this case, M is a maximum weight matching and *u,v* is a minimum weight cover

Equality Subgraph

- Equality subgraph of a bipartite graph G = (V, E) corresponding to a cover (u,v) is a graph $G_{u,v}$ = (V, E'), where (x_i,y_j) in E' iff $u_i + v_j = w_{ij}$

Hungarian Algorithm

Input: A matrix of weights on the edges of $K_{n,n}$ with bipartition X, Y.

Idea: Maintain a cover u,v, iteratively reducing the cost of the cover until the equality subgraph $G_{u,v}$ has a perfect matching.

Initialization: Let u,v be a feasible labeling, such as $u_i = max_j\ w_{ij}$ and $v_j = 0$, and find a maximum matching M in $G_{u,v}$.

Hungarian Algorithm

Iteration:

- If M is a perfect matching, stop and report M as a maximum weight matching.
- Otherwise, find a minimum vertex cover Q
- Let $R = Q \cap X$ and $T = Q \cap Y$
- Let

  $\varepsilon = \min\{u_i + v_j - w_{ij}: x_i \in X - R, y_j \in Y - T\}$
- Decrease u_i by ε for all $x_i \in X\text{-}R$, and increase v_j by ε for all $y_j \in T$. If the new equality subgraph G' contains an M-augmenting path, replace M by a maximum matching in G' and iterate. Otherwise, iterate without changing M.

Hungarian Algorithm

- The Hungarian Algorithm finds a maximum weight matching and a minimum cost cover

Stable Matchings

- Given *n* men and *n* women, we wish to establish *n* stable marriages.
  - If man *x* and woman *a* prefers each other over their existing partners, then they might leave their current partners and switch to each other.
  - In this case we say that the unmatched pair *(x,a)* is an unstable pair.
  - A perfect matching is a *stable matching* if it yields no unstable matched pair.

Gale-Shapley Proposal Algorithm

Input: Preference rankings by each of *n* men and *n* women.

Iteration:

- Each man proposes to the highest woman on his preference list who has not previously rejected him.
- If each woman receives exactly one proposal, stop and use the resulting matching.
- Otherwise, every woman receiving more than one proposal rejects all of them except the one that is highest on her preference list.
- Every woman receiving a proposal says "*maybe*" to the most attractive proposal received.

The algorithm produces a stable matching.

Matchings in General Graphs

- A *factor* of a graph G is a spanning sub-graph of G
  - A *k-factor* is a spanning *k-regular* sub-graph.
  - An *odd component* of a graph is a component of odd order; the number of odd components of H is o(H)

- [Tutte 1947]: A graph G has a 1-factor if and only if $o(G - S) \leq |S|$ for every $S \subseteq V(G$

Edmond's Blossom Algorithm

- Let M be a matching in a graph G, and let *u* be an M-unsaturated vertex
- A *flower* is the union of two *M*-alternating paths from *u* that reach a vertex *x* on steps of opposite parity
- The *stem* of the flower is the maximal common initial path
- The *blossom* of the flower is the odd cycle obtained by deleting the stem

Edmond's Blossom Algorithm

Input: A graph G, a matching M in G, and an M-unsaturated vertex *u.*

Initialization: S = {*u*} and T = { }

Iteration:

- If S has no unmarked vertex, stop
- Otherwise, select an unmarked vertex $v \in S$. To explore from *v,* successively consider each $y \in N(v)$ such that $y \notin T$.
- If *y* is unsaturated by M, then trace back from *y* to report an M-augmenting *u,y*-path.
- If $y \in S$, then a blossom has been found. Contract the blossom and continue the search from this vertex in the smaller graph.
- Otherwise, *y* is matched to some *w* by M. Include *y* in *T* (reached from *v*), and include *w* in *S*.
- After exploring all such neighbors of *v,* mark *v* and iterate.