Graphs

- (p 186) Graphs G = (V,E)
 - set V of vertices, each with a unique name
 - Note: book calls vertices as nodes
 - set E of edges between vertices, each encoded as tuple of 2 vertices as in (u,v)
- Edges may be **directed** (from u to v) or **undirected**
 - Undirected edge eqvt to pair of directed edges
- Example of undirected graph

 $G = (4) \quad \langle G \rangle = (1,2,3,4) ((1,2), (2,3), (3,1), (1,4))$

- Labeled graph: each edge has a "name"
- Weighted graph: each edge has numerical value
 - E.g. in graph of V=cities, weight on (u,v) is distance from u to v

Terminology about Graphs (see P. 12)

- Outdegree of a vertex u: # of edges leaving it
 - i.e. |{(u, v)}| for some v
- Indegree of a vertex v: # of edges entering it
 - i.e. |{(u, v)}| for some u
 - For undirected graph outdegree(u) = indegree(u) = degree(u)
- K-regular: every vertex has degree k (p. 13)
- Subgraph: subsets V' and E' of V and E where all edges in E' are between vertices in V'
- Path of length k: from u to v if a set of k edges in G (u_i , v_i), $1 \le i \le k$, where $u_1 = u$, $v_i = u_{i+1}$, and $v_k = v$
- Simple path: no vertices are repeated
- Shortest path: between u & v is simple path of shortest lengthg
- Diameter: longest shortest path between any 2 vertices
- Hamiltonian Path: goes thru every vertex once
- Cycle: a path exists from u back to u

- Connected iff every vertex can be reached from every other vertex by some path
- **Strongly connected** iff a directed path from each vertex to every other
- Tree: graph with no simple paths: Has root & leaves
- Vertex Cover of size k: subset of k vertices where every edge touches at least one of them (p. 312)
- K-Clique: subset of k vertices where there is an edge between every pair in it.
- Two graphs G & H are **isomorphic** if vertices of one can be reordered so that graphs are identical
- **Spanning Tree**: subgraph that forms a tree that includes all vertices, but with minimum # of edges
- Flow Network: directed weighted graph where:
 - Each edge can carry a "flow" (a number)
 - Each edge has a "capacity" (max possible flow value)
 - For each vertex, \sum incoming flow = \sum outgoing flow
 - Except for some **source** that generates out flow
 - And some **sink** which has no outgoing flow

Graph Data Structures

- Assume |V| = N, |E| = M
- Adjacency Matrix: NxN Boolean matrix A where
 - A[u,v] = 1 if (u,v) in E
 - A[u,v] = 0 otherwise
 - If graph is weighted, A[u,v] = weight on (u,v)
- CSR (Compressed Sparse Row): 3 vectors
 - A: M vector of weights (one per edge)
 - JA: M vector of vertex indecies
 - IA: N+1 vector of indices into A, JA
 - IA[u] = index into A, JA for 1st edge from u
 - JA[IA[u]] thru JA[IA[u+1]-1] are the v's for edges (u,v)
 - Matching elements in A are weights
- Lots of variations

Spanning Trees & BFS

- Common problem: starting at some vertex u, find tree that reaches as many vertices as possible
 - If all vertices reachable, then a "spanning tree"
- Common algorithm: **BFS** (**Breadth First Search**)
 - Frontier: set of all vertices that have been reached "for 1st time"
 - At start, just u
 - Also each vertex can be marked as "touched" or not
 - At start only u so marked
 - For each vertex in current frontier:
 - Follow each outgoing edge
 - If other vertex is not touched:
 - Mark as touched
 - Add to new frontier
 - When frontier empty, swap with new frontier and repeat
- Option: when first touched, mark vertex with "level"
 - Level = # of edges from root u
 - At end, final level = diameter of reached subgraph
- Clearly polynomial time

- Basis for the **GRAPH500** benchmark
 - www.graph500.org
 - Search graphs with up to trillions of vertices
 - Literally thousands of different implementations on different computers, esp. parallel
 - Established by an ND quad-domer
- Beamer's Algorithm:
 - When frontier too large, instead explore all non-touched
 - If any of them have an edge to current touched, mark them as touched
 - Can reduce time by 10X

Maximum Flow Problems

- Max flow: given a flow network, find largest flow from source to sink where no edge exceeds its capacity
- Ford-Fulkerson Algorithm
 - https://en.wikipedia.org/wiki/Ford%E2%80%93Fulkerson_algorithm
 - Define:
 - c(u,v) = capacity of edge (u, v)
 - f(u,v) = flow on edge (u,v)
 - Residual Network G_f(V,E_f) = network with capacity c_f(u,v) = c(u,v) - f(u,v) ("residual flow")

Output Compute a flow *f* from *s* to *t* of maximum value

1. $f(u, v) \leftarrow 0$ for all edges (u, v)

2. While there is a path p from s to t in G_f , such that $c_f(u, v) > 0$ for all edges $(u, v) \in p$:

- 1. Find $c_f(p) = \min\{c_f(u,v) : (u,v) \in p\}$
- 2. For each edge $(u, v) \in p$
 - 1. $f(u, v) \leftarrow f(u, v) + c_f(p)$ (Send flow along the path)
 - 2. $f(v, u) \leftarrow f(v, u) c_f(p)$ (The flow might be "returned" later)

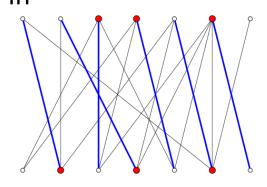
■ "←" denotes <u>assignment</u>. For instance, "largest ← item" means that the value of largest changes to the value of item.

"return" terminates the algorithm and outputs the following value.

- "Finding paths" can use BFS
- Each new path chosen is called "augmenting path"

Bipartite Graphs

- References:
 - https://en.wikipedia.org/wiki/Bipartite_graph
- Bipartite Graph:
 - 2 disjoint sets of vertices U and V, called "parts"
 - Every edge connects vertex from U with one from V
- Matching: subset of edges where no two edges share an endpoint
 - Maximal Matching: edge set is largest possible matching
 - **Perfect Matching**: |U|=|V|=|matching set|
- Examples of Bipartite Graphs:
 - Athletes and Teams they played with
 - Actors and Movies they acted in
 - Trains and Stations
 - Social networks
 - All graphs that are trees
 - Graphs that form single cycles with even # vertices
 - Graphs that can be written
 "on a 2D plane" without edges crossing



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- Properties
 - No odd cycles
 - 2-colorable
- Kőnig's theorem: # of edges in maximum matching = # of vertices in a minimum vertex cover
- O(VE) algorithm: Hopcroft-Karp
 - https://en.wikipedia.org/wiki/Hopcroft%E2%80%93Karp_algorithm
 - Similar to Ford-Fulkerson

```
Input: Bipartite graph G(U \cup V, E)

Output: Matching M \subseteq E

M \leftarrow \emptyset

repeat

\mathcal{P} \leftarrow \{P_1, P_2, \dots, P_k\} maximal set of vertex-disjoint shortest

augmenting paths

M \leftarrow M \oplus (P_1 \cup P_2 \cup \dots \cup P_k)

until \mathcal{P} = \emptyset
```

- P_k = augmenting path of length k
- "Free vertex" does not appear in current matching M
- Iteratively find "augmenting path"
 - Use BFS to partition vertices into layers
 - Current free vertices from U are the 1st layer
 - Pick a free vertex from 1st level
 - Use BFS to create a tree
 - Alternate edges between not in M and in M
 - Stop tree if reach a free vertex u from U

- Ends at level k when ≥1 free vertices in V are reached
- Define free vertices v from V into a set F
- Choose some shortest tree and then shortest path
 - Called augmenting path (of length k)
- Remove all edges from M that are in the path
- Add in all edges from path that are not in M
- Repeat until no free vertex from U has an augmenting path to a free vertex in V
- Why does this work?
 - At each iteration, always adding 1 more edge to M than deleting
 - Guaranteed to stop at min(|U|, |V|)
- Complexity: for each vertex in V may look at each edge in E

