- For two positive functions f, g defined for positive ints:  $-f = O(g) \text{ iff } \exists N, C \text{ s.t. } f(n) \le Cg(n) \forall n \ge N$   $-f = \Omega(g) \text{ if } g = O(f)$   $-f = \Theta(g) \text{ if } f = O(g) \text{ and } g = O(f)$
- $\lim_{n \to \infty} \frac{f(n)}{g(n)} = \begin{cases} \text{finite} \implies f(n) = O(g(n)) \\ \text{non-zero} \implies f(n) = \Omega(g(n)) \end{cases}$
- $a^{\log_b n} = n^{\log_b a}$
- Euclid's Algorithm: if  $a \ge b > 0$ , then gcd(a, b) = $gcd(b, a \mod b)$
- Fermat's Little Theorem: If p is prime then for every  $a: 1 \le a < p, a^{p-1} \equiv 1 \mod p.$
- Euler's Theorem : If a, n coprime:  $a^{\phi(n)} = 1 \mod n$ • RSA

  - Primes p, q; N = pq-  $ed \equiv 1 \mod (p-1)(q-1)$  Public key: (N, e). Private key: d- Encryption  $x \mapsto x^e$ . Decryption  $y \mapsto y^d$
- H is a universal hash function if for every two items x and
- y, exactly |H|/n of the functions map x and y to the same bucket (n = number of buckets)
- Master Theorem: Given  $T(n) = aT(\frac{n}{h}) + O(n^d)$ •  $T(n) = \begin{cases} O(n^d) & \text{if } d > \log_b a \\ O(n^d \log n) & \text{if } d = \log_b a \end{cases}$

$$\begin{bmatrix}
O(n^{\log_b a}) & \text{if } d < \log_b a
\end{bmatrix}$$

- FFT-recursive:  $A(\omega^j) = A_e(\omega^{2j}) + \omega^j A_o(\omega^{2j}), \omega = e^{2\pi i/n}$
- FFT-Matrix:  $M_n(\omega)_{j,k} = \omega^{jk}$ , for  $0 \le j, k < n$ . Inverse:  $M_n(\omega)^{-1} = \frac{1}{n} M_n(\omega^{-1})$
- After DFS traversal, (u, v) is a:

  - tree edge iff  $[\operatorname{pre}(v), \operatorname{post}(v)] \in [\operatorname{pre}(u), \operatorname{post}(u)]$  back edge iff  $[\operatorname{pre}(u), \operatorname{post}(u)] \in [\operatorname{pre}(v), \operatorname{post}(v)]$  cross edge iff  $[\operatorname{pre}(v), \operatorname{post}(v)]$  then  $[\operatorname{pre}(u), \operatorname{post}(u)]$
- SCC: Vertex with highest post number is in a source component. Can run on  $G^R$  to identify sinks.
- Dijkstra /Bellman-Ford for paths. Linearize for DAGs.
- Greedy. MST: Kruskal's, Prim's. Huffman encoding, Horn SAT, set cover approximation.
- Runtimes:

Dijkstra:  $O((|V| + |E|) \log |V|)$  with binary heap Bellman-Ford: O(|V||E|)

- Kruskal:  $O(|E| \log |V|)$ ; if E sorted, then  $O(|E| \log^* |V|)$
- DP: Longest common subsequence, etc. Floyd-Warshall for all-pairs shortest paths (expand permissible intermediate nodes)
- LP: Simplex, max-flow, bipartite matching - Standard Form: min  $\vec{c} \cdot \vec{x}$ ,  $A\vec{x} \ge \vec{b}$ ,  $\vec{x} \ge \vec{0}$ - Dual LP: max  $y^T \vec{b}$ ,  $y^T A \ge \vec{c}$ ,  $\vec{y} \ge \vec{0}$

# Ford-Fulkerson (Runtime: $O(|E| \times F)$ )

- Find s-t path, p with positive residual capacity (w/ DFS)
- Let  $\delta$  be minimum r(e) of edge in p
- For each edge, e = (u, v) in path
  - if e in G,  $f(e) = f(e) + \delta$

- if 
$$(v, u)$$
 in  $G$ ,  $f(v, u) = f(v, u) - \delta$ 

- compute residual capacities
- repeat; or terminate if no s t path with pos. residual capacity

# Strategic / Two-Person Zero-Sum Games

- Row  $(\vec{x})$  maximizes, column  $(\vec{y})$  minimizes.
- Row 1<sup>st</sup>:  $\min_y \max_x (x^T A y) = \max_x \min_y (x^T A y)$  :Col 1<sup>st</sup>
- LP for Row:  $\max z$ ,  $\sum x_i = 1$ .  $z \leq$  the expected value for each pure strategy from column (e.g.  $z \leq x_2 - x_3$ )
- LP for Col:  $\min w$ ,  $\sum y_i = 1$ .  $w \ge$  the expected value for each pure strategy from row (e.g.  $w \ge y_1 - y_3$ )

#### Simplex

- Start at origin (assume feasible). If all coefficients  $c_i \leq 0$ in objective function (for maximization), then optimal
- Increase  $x_i$  with highest positive coefficient  $c_i$  in objective function until a constraint is hit
- Repeat with new coord system defining curr pt as origin  $-y_i$  is distance from constraint *i*.  $y_i = b_i - \vec{a_i} \cdot \vec{x}$  $(\vec{a_i} \text{ is row } i \text{ of } A, \text{ i.e. the coeffs for constraint } i)$ - Solve for  $x_i$ 's from sys of eq, sub into obj/constraints
- Origin not feasible? Use new LP. New vars  $z_i$ , subtract from LHS of each constraint.  $\min(\sum z_i), z_j \ge 0$ . Start at  $x_i = 0, z_i = \max(-b_i, 0)$ . Result x values are new start vertex for original LP.

## NP complete problems

- SAT: Boolean formula in conjunctive normal form (CNF): clauses containing OR of literals, AND of these clauses
- TSP: Is there a tour, which visits each node exactly once, within budget?
- RUDRATA PATH: Path passing thru each vertex exactly once
- BALANCED CUT: With budget b for cut, partition into two sets such that each has  $\geq 1/3$  of elems
- ILP: Linear programming, but constrain variables to be integer (ZOE: constrain to be binary,  $A\vec{x} = \vec{1}$ )
- 3D matching: Given m valid tuples, match n boys, girls, and pets: find n disjoint triples that get along
- INDEPENDENT SET: Graph, integer g: find g vertices that are independent (no 2 share edge)
- VERTEX COVER: b vertices that touch every edge
- DOMINATING SET: b vertices such that every edge is in the set, or has a neighbor in the set
- SET COVER: given subsets, select b subsets such that union is the entire set
- CLIQUE: g vertices, each is connected to every other
- LONGEST PATH: is there a simple path of length g or more from s to t?
- KNAPSACK: Given weights and values for items, find best knapsack value with weight at most W

## Reductions

- Any problem in NP  $\rightarrow$  SAT (through CSAT)
- SAT  $\rightarrow$  3SAT
- $3SAT \rightarrow \{Independent Set, 3D Matching\}$
- Independent Set  $\rightarrow$  {Vertex Cover, Clique}
- 3D Matching  $\rightarrow$  ZOE
- $ZOE \rightarrow \{Subset Sum, ILP, Rudrata Cycle\}$
- Rudrata Cycle  $\rightarrow$  TSP

#### Coping with NP Completeness

- Backtracking: Try an assignment, test if it meets our constraints. If not, then reject.
- Branch and Bound: To reject a subproblem in a minimization problem, we must be certain that its cost exceeds that of some other solution.
- Approximation Alg.: For a min. problem, find a solution with factor  $\alpha_A = \max_I \frac{A(I)}{OPT(I)}$  away from the optimum. • Local Search: Suppose we have a set of solutions. We
- define a neighborhood structure to relate these solutions, then we we seek the local optima.
- Matching subset of edges that have no vertices in common, any matching is a lower bound on the optimal solution for vertex cover