

Chapter 7

Network Flow



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* 7.13 Assignment Problem

Assignment Problem

Assignment problem.

- Input: weighted, complete bipartite graph $G = (L \cup R, E)$ with |L| = |R|.
- Goal: find a perfect matching of min weight.

| | 1' | 2' | 3' | 4' | 5' | |
|---|----|----|----|----|----|--|
| 1 | 3 | 8 | 9 | 15 | 10 | |
| 2 | 4 | 10 | 7 | 16 | 14 | |
| 3 | 9 | 13 | 11 | 19 | 10 | |
| 4 | 8 | 13 | 12 | 20 | 13 | |
| 5 | 1 | 7 | 5 | 11 | 9 | |

Applications

Natural applications.

- Match jobs to machines.
- Match personnel to tasks.
- Match PU students to writing seminars.

Non-obvious applications.

- Vehicle routing.
- Signal processing.
- Virtual output queueing.
- Multiple object tracking.
- Approximate string matching.
- Enhance accuracy of solving linear systems of equations.

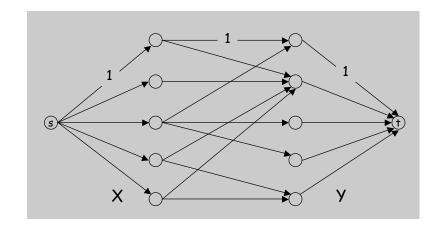
Bipartite Matching

Bipartite matching. Can solve via reduction to max flow.

Flow. During Ford-Fulkerson, all capacities and flows are 0/1. Flow corresponds to edges in a matching M.

Residual graph G_M simplifies to:

- If $(x, y) \notin M$, then (x, y) is in G_M .
- If $(x, y) \in M$, the (y, x) is in G_M .

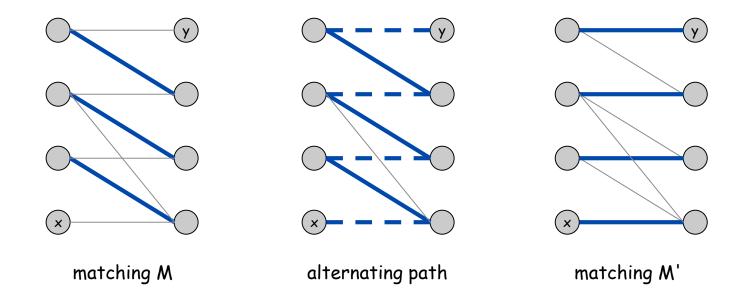


Augmenting path simplifies to:

- Edge from s to an unmatched node $x \in X$.
- Alternating sequence of unmatched and matched edges.
- Edge from unmatched node $y \in Y$ to t.

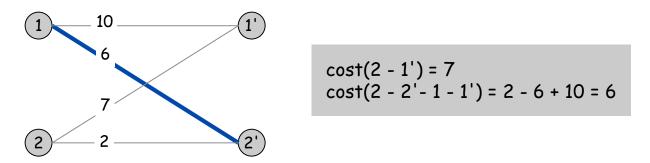
Alternating Path

Alternating path. Alternating sequence of unmatched and matched edges, from unmatched node $x \in X$ to unmatched node $y \in Y$.



Assignment Problem: Successive Shortest Path Algorithm

Cost of an alternating path. Pay c(x, y) to match x-y; receive c(x, y) to unmatch x-y.



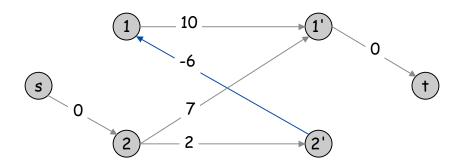
Shortest alternating path. Alternating path from any unmatched node $x \in X$ to any unmatched node $y \in Y$ with smallest cost.

Successive shortest path algorithm.

- Start with empty matching.
- Repeatedly augment along a shortest alternating path.

Finding The Shortest Alternating Path

Shortest alternating path. Corresponds to shortest s-t path in G_M .



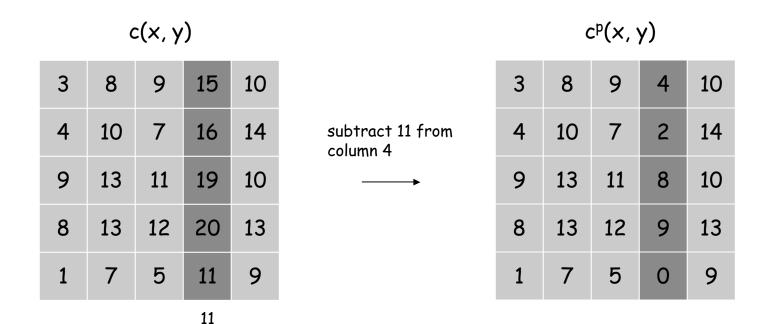
Concern. Edge costs can be negative.

Fact. If always choose shortest alternating path, then G_M contains no negative cycles \Rightarrow compute using Bellman-Ford.

Our plan. Use duality to avoid negative edge costs (and negative cost cycles) \Rightarrow compute using Dijkstra.

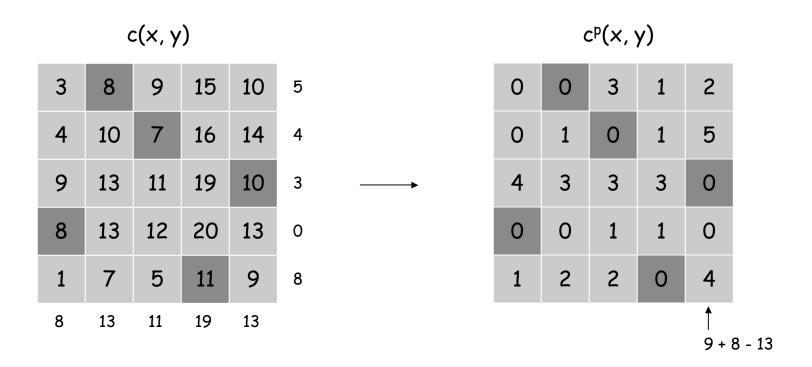
Equivalent Assignment Problem

Duality intuition. Adding (or subtracting) a constant to every entry in row x or column y does not change the min cost perfect matching(s).



Equivalent Assignment Problem

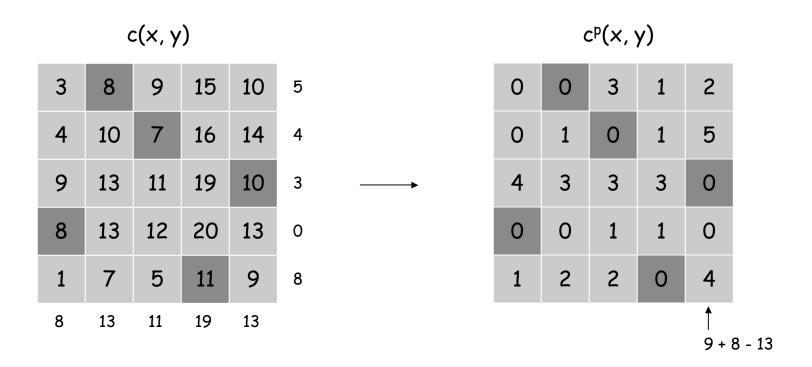
Duality intuition. Adding p(x) to row x and subtracting p(y) from row y does not change the min cost perfect matching(s).



Reduced Costs

Reduced costs. For $x \in X$, $y \in Y$, define $c^p(x, y) = p(x) + c(x, y) - p(y)$.

Observation 1. Finding a min cost perfect matching with reduced costs is equivalent to finding a min cost perfect matching with original costs.



Compatible Prices

Compatible prices. For each node v, maintain prices p(v) such that:

- (i) $c^p(x, y) \ge 0$ for for all $(x, y) \notin M$.
- (ii) $c^p(x, y) = 0$ for for all $(x, y) \in M$.

Observation 2. If p are compatible prices for a perfect matching M, then M is a min cost perfect matching.

| c(x, y) | | | | | $c^p(x, y)$ | | | | | | | | |
|---------|---|----|----|----|-------------|---|--|---|---|---|---|---|---|
| | 3 | 8 | 9 | 15 | 10 | 5 | | | 0 | 0 | 3 | 1 | 2 |
| | 4 | 10 | 7 | 16 | 14 | 4 | | | 0 | 1 | 0 | 1 | 5 |
| | 9 | 13 | 11 | 19 | 10 | 3 | | • | 4 | 3 | 3 | 3 | 0 |
| | 8 | 13 | 12 | 20 | 13 | 0 | | | 0 | 0 | 1 | 1 | 0 |
| | 1 | 7 | 5 | 11 | 9 | 8 | | | 1 | 2 | 2 | 0 | 4 |
| | Q | 12 | 11 | 10 | 12 | | | | | | | | |

$$cost(M) = \Sigma_{(x, y) \in M} c(x, y) = (8+7+10+8+11) = 44$$

 $cost(M) = \Sigma_{y \in Y} p(y) - \Sigma_{x \in X} p(x) = (8+13+11+19+13) - (5+4+3+0+8) = 44$

Successive Shortest Path Algorithm

Successive shortest path.

Maintaining Compatible Prices

Lemma 1. Let p be compatible prices for matching M. Let d be shortest path distances in G_M with costs c^p . All edges (x, y) on shortest path have $c^{p+d}(x, y) = 0$.

Pf. Let (x, y) be some edge on shortest path.

- If $(x, y) \in M$, then (y, x) on shortest path and $d(x) = d(y) c^p(x, y)$. If $(x, y) \notin M$, then (x, y) on shortest path and $d(y) = d(x) + c^p(x, y)$.
- In either case, $d(x) + c^p(x, y) d(y) = 0$.
- By definition, $c^p(x, y) = p(x) + c(x, y) p(y)$.
- Substituting for $c^p(x, y)$ yields: (p(x) + d(x)) + c(x, y) - (p(y) + d(y)) = 0.
- In other words, $c^{p+d}(x, y) = 0$. ■

Reduced costs: $c^p(x, y) = p(x) + c(x, y) - p(y)$.

Maintaining Compatible Prices

Lemma 2. Let p be compatible prices for matching M. Let d be shortest path distances in G_M with costs c^p . Then p' = p + d are also compatible prices for M.

Pf.
$$(x, y) \in M$$

- (y, x) is the only edge entering x in G_M . Thus, (y, x) on shortest path.
- By Lemma 1, $c^{p+d}(x, y) = 0$.

Pf. $(x, y) \notin M$

- (x, y) is an edge in $G_M \Rightarrow d(y) \le d(x) + c^p(x, y)$.
- Substituting $c^{p}(x, y) = p(x) + c(x, y) p(y) \ge 0$ yields $(p(x) + d(x)) + c(x, y) (p(y) + d(y)) \ge 0$.
- In other words, $c^{p+d}(x, y) \ge 0$. ■

Compatible prices. For each node v:

- (i) $c^p(x, y) \ge 0$ for for all $(x, y) \notin M$.
- (ii) $c^p(x, y) = 0$ for for all $(x, y) \in M$.

Maintaining Compatible Prices

Lemma 3. Let M' be matching obtained by augmenting along a min cost path with respect to c^{p+d} . Then p' = p + d is compatible with M'.

Pf.

- By Lemma 2, the prices p + d are compatible for M.
- Since we augment along a min cost path, the only edges (x, y) that swap into or out of the matching are on the shortest path.
- By Lemma 1, these edges satisfy $c^{p+d}(x, y) = 0$.
- Thus, compatibility is maintained.

Compatible prices. For each node v:

- (i) $c^p(x, y) \ge 0$ for for all $(x, y) \notin M$.
- (ii) $c^p(x, y) = 0$ for for all $(x, y) \in M$.

Successive Shortest Path: Analysis

Invariant. The algorithm maintains a matching M and compatible prices p.

Pf. Follows from Lemmas 2 and 3 and initial choice of prices.

Theorem. The algorithm returns a min cost perfect matching. Pf. Upon termination M is a perfect matching, and p are compatible prices. Optimality follows from Observation 2.

Theorem. The algorithm can be implemented in $O(n^3)$ time. Pf.

- Each iteration increases the cardinality of M by $1 \Rightarrow n$ iterations.
- Bottleneck operation is computing shortest path distances d.
 Since all costs are nonnegative, each iteration takes O(n²) time using (dense) Dijkstra.

Weighted Bipartite Matching

Weighted bipartite matching. Given weighted bipartite graph, find maximum cardinality matching of minimum weight.

* m edges, n nodes*

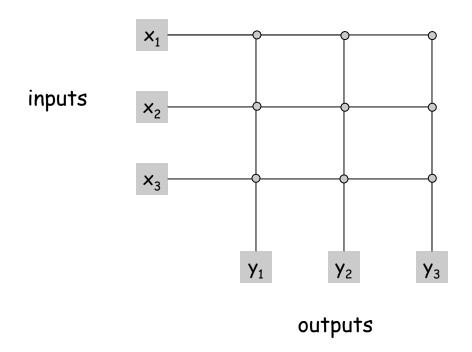
Successive shortest path algorithm. O(mn log n) time using heap-based version of Dijkstra's algorithm.

Best known bounds. $O(mn^{1/2})$ deterministic; $O(n^{2.376})$ randomized.

Planar weighted bipartite matching. $O(n^{3/2} \log^5 n)$.

Input-queued switch.

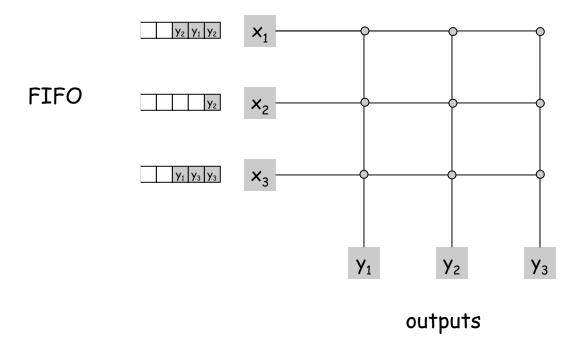
- n inputs and n outputs in an n-by-n crossbar layout.
- At most one cell can depart an input at a time.
- At most one cell can arrive at an output at a time.
- Cell arrives at input x and must be routed to output y.



FIFO queueing. Each input x maintains one queue of cells to be routed.

Head-of-line blocking (HOL).

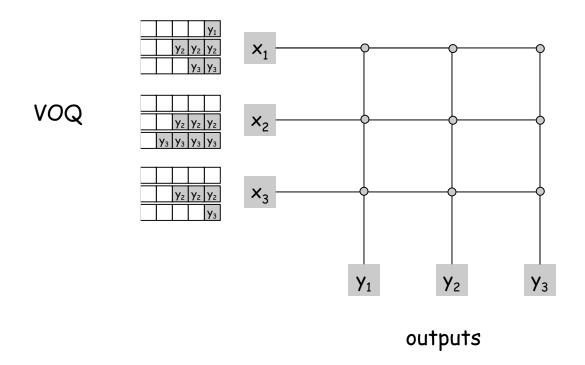
- A cell can be blocked by a cell queued ahead of it that is destined for a different output.
- Can limit throughput to 58%, even when arrivals are uniform.



Virtual output queueing (VOQ). Each input x maintains n queue of cells, one for each output y.

Maximum size matching. Find a max cardinality matching.

- Achieves 100% when arrivals are uniform.
- Can starve input-queues when arrivals are non-uniform.



Max weight matching. Find a min cost perfect matching between inputs x and outputs y, where c(x, y) equals:

- [LQF] The number of cells waiting to go from input x to output y.
- [OCF] The waiting time of the cell at the head of VOQ from x to y.

Theorem. LQF and OCF achieve 100% throughput if arrivals are independent.

Practice.

- Too slow in practice for this application; difficult to implement in hardware. Provides theoretical framework.
- Use maximal (weighted) matching \Rightarrow 2-approximation.