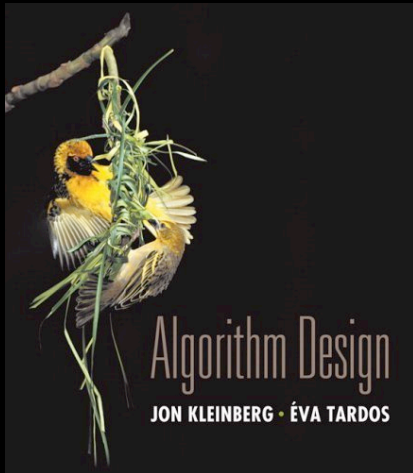


Chapter 7

Network Flow



Slides by Kevin Wayne.
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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 |

Min cost perfect matching

$$M = \{ 1-2', 2-3', 3-5', 4-1', 5-4' \}$$

$$\text{cost}(M) = 8 + 7 + 10 + 8 + 11 = 44$$

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.

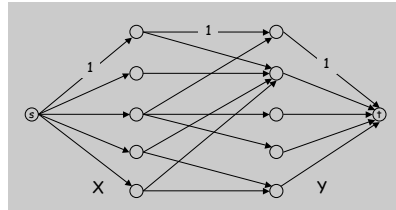
Bipartite Matching

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 .

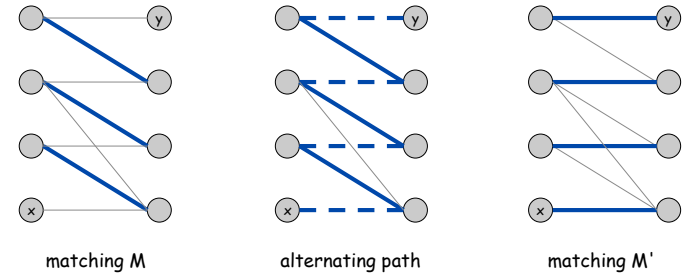


Augmenting path simplifies to:

- 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 $y \in Y$.

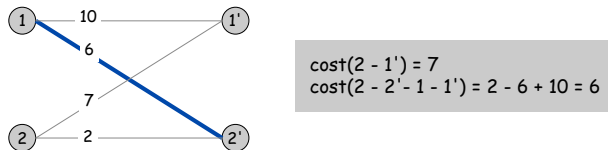


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Assignment Problem: Successive Shortest Path Algorithm

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



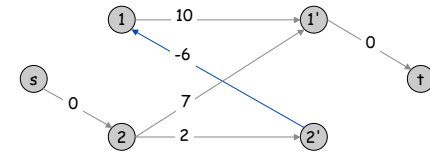
Shortest alternating path. Alternating path from any unmatched node $x \in X$ to any unmatched node $y \in Y$ with smallest cost.

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.

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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).

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

11

→ subtract 11 from column 4

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

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).

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

8 13 11 19 13

→

| $c^p(x, y)$ | | | | |
|-------------|---|---|---|---|
| 0 | 0 | 3 | 1 | 2 |
| 0 | 1 | 0 | 1 | 5 |
| 4 | 3 | 3 | 3 | 0 |
| 0 | 0 | 1 | 1 | 0 |
| 1 | 2 | 2 | 0 | 4 |

↑
9 + 8 - 13

9

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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.

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

8 13 11 19 13

→

| $c^p(x, y)$ | | | | |
|-------------|---|---|---|---|
| 0 | 0 | 3 | 1 | 2 |
| 0 | 1 | 0 | 1 | 5 |
| 4 | 3 | 3 | 3 | 0 |
| 0 | 0 | 1 | 1 | 0 |
| 1 | 2 | 2 | 0 | 4 |

↑
9 + 8 - 13

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Compatible Prices

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

- (i) $c^p(x, y) \geq 0$ 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.

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

8 13 11 19 13

→

| $c^p(x, y)$ | | | | |
|-------------|---|---|---|---|
| 0 | 0 | 3 | 1 | 2 |
| 0 | 1 | 0 | 1 | 5 |
| 4 | 3 | 3 | 3 | 0 |
| 0 | 0 | 1 | 1 | 0 |
| 1 | 2 | 2 | 0 | 4 |

↑
9 + 8 - 13

$\text{cost}(M) = \sum_{(x,y) \in M} c(x, y) = (8+7+10+8+11) = 44$
 $\text{cost}(M) = \sum_{y \in Y} p(y) - \sum_{x \in X} p(x) = (8+13+11+19+13) - (5+4+3+0+8) = 44$

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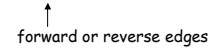
Successive shortest path.

```

Successive-Shortest-Path(X, Y, c) {
  M ← ∅
  foreach x ∈ X: p(x) ← 0
  foreach y ∈ Y: p(y) ← mine into y c(e)      p is compatible
                                             with M = ∅

  while (M is not a perfect matching) {
    Compute shortest path distances d
    P ← shortest alternating path using costs cP
    M ← updated matching after augmenting along P
    foreach v ∈ X ∪ Y: p(v) ← p(v) + d(v)
  }
  return M
}
    
```

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

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 $G_M \Rightarrow d(y) \leq d(x) + c^P(x, y)$.
 - Substituting $c^P(x, y) = p(x) + c(x, y) - p(y) \geq 0$ yields
 $(p(x) + d(x)) + c(x, y) - (p(y) + d(y)) \geq 0$.
 - In other words, $c^{P+d}(x, y) \geq 0$. ■

Compatible prices. For each node v :

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

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) $c^P(x, y) \geq 0$ for for all $(x, y) \notin M$.
- (ii) $c^P(x, y) = 0$ for for all $(x, y) \in M$.

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 $O(n^2)$ time using (dense) Dijkstra. ■

Input Queued Switching

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.

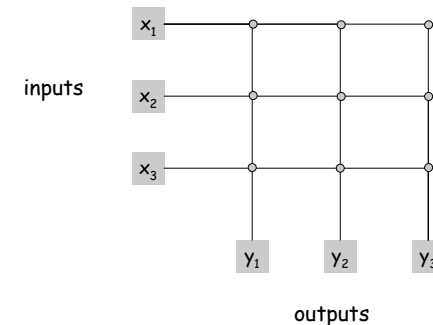
Best known bounds. $O(mn^{1/2})$ deterministic; $O(n^{2.376})$ randomized.

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

Input-Queued Switching

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 .

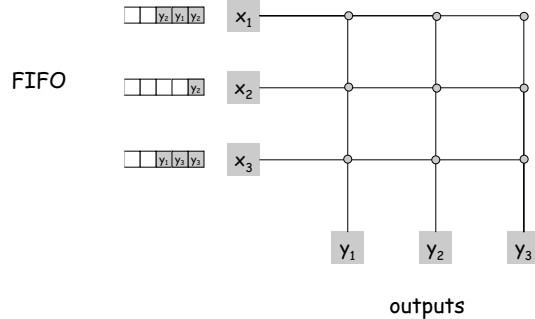


Input-Queued Switching

FIFO queuing. 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.



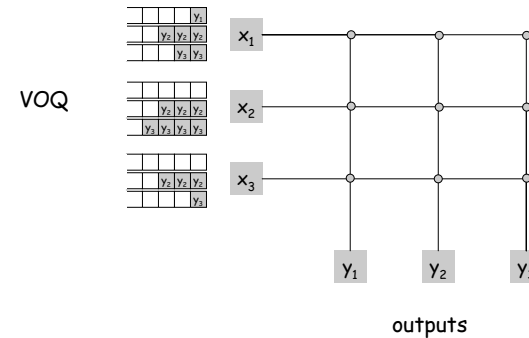
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Input-Queued Switching

Virtual output queuing (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.



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Input-Queued Switching

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.