

## 7.13 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.

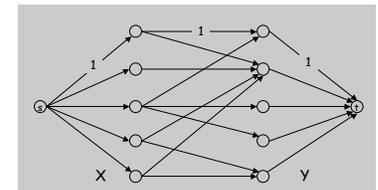
### Bipartite Matching

**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  $G_M$ .

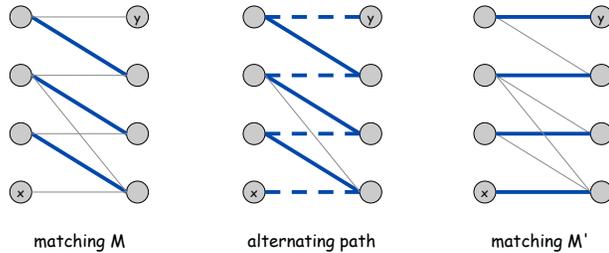


#### Augmenting path simplifies to:

- 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  $t$ .

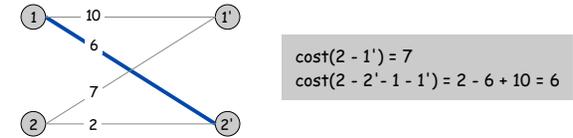
## Alternating Path

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



## 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.**

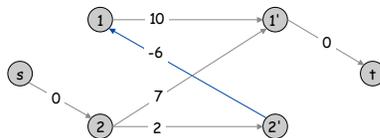
- Start with empty matching.
- Repeatedly augment along a **shortest** alternating path.

5

6

## 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).

$c(x, y)$						$c^p(x, y)$				
3	8	9	15	10	subtract 11 from column 4 $\longrightarrow$	3	8	9	4	10
4	10	7	16	14		4	10	7	2	14
9	13	11	19	10		9	13	11	8	10
8	13	12	20	13		8	13	12	9	13
1	7	5	11	9		1	7	5	0	9

11

7

8

## 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

## 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	
					↑ $9 + 8 - 13$

10

## 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	

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

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

11

## 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  $c^{p+d}(x, y) = 0$ .

↑  
forward or reverse edges

**Pf.**

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

12

## 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  $G_M \Rightarrow d(y) \leq d(x) + c^p(x, y)$ .
- 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  $(x, y) \in M$ .

13

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

14

## Successive Shortest Path Algorithm

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
}
    
```

15

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

16

## 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.**  $O(n^{3/2} \log^5 n)$ .

17

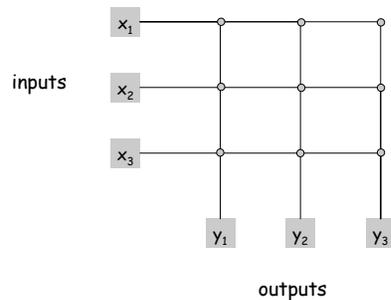
Algorithm Design by Éva Tardos and Jon Kleinberg · Copyright © 2005 Addison Wesley · Slides by Kevin Wayne

## Input Queued Switching

### 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.

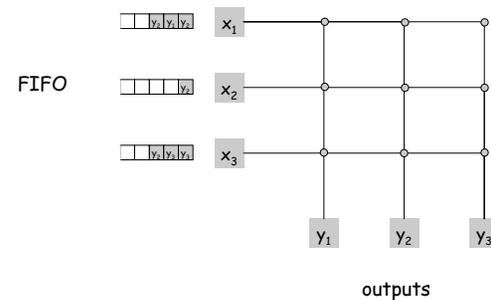


### Input-Queued Switching

**FIFO 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.



19

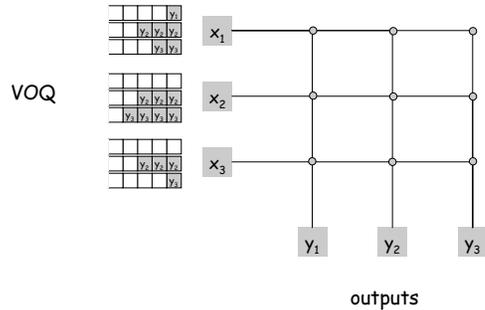
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## Input-Queued Switching

**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.



## 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.

Reference: <http://robotics.eecs.berkeley.edu/~wlr/Papers/AMMW.pdf>