### Algorithm II

# 7. Network Flow III

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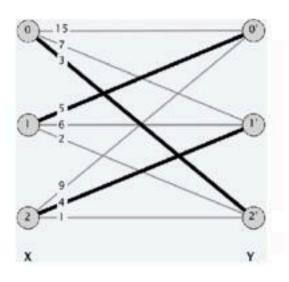


# Assignment problem

# Assignment problem

**Input**. Weighted, complete bipartite graph  $G = (X \cup Y, E)$  with |X| = |Y|.

Goal. Find a perfect matching of min weight.



min-cost perfect matching M = { 0-2', 1-0', 2-1' }

$$cost(M) = 3 + 5 + 4 = 12$$



# Seminar assignment

**Goal**. Given m seminars and n = 12m students who rank their top 8 choices, assign each student to one seminar so that:

- Each seminar is assigned exactly 12 students.
- Students tend to be "happy" with their assigned seminar.

#### Solution.

- Create one node for each student i and 12 nodes for each seminar j.
- Solve assignment problem where  $c_{ij}$  is some function f of the ranks:

$$c_{ij} = \left\{ egin{array}{ll} f(rank(i,j)) & ext{if } i ext{ ranks } j \ \infty & ext{otherwise} \end{array} 
ight.$$



# Applications

### Natural applications.

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

#### Non-obvious applications.

- · Vehicle routing.
- Signal processing.
- Earth-mover's distance.
- Multiple object tracking.
- Virtual output queueing.
- Handwriting recognition.
- Locating objects in space.
- Approximate string matching.
- Enhance accuracy of solving linear systems of equations.

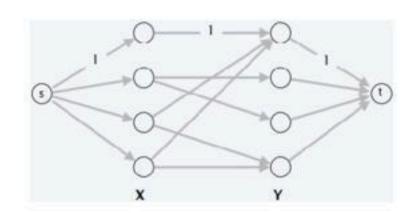
# Bipartite matching

Bipartite matching. Can solve via reduction to maximum flow.

**Flow**. During Ford-Fulkerson, all residual 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  $(x,y) \in M$ , then (y,x) is in  $G_M$ .



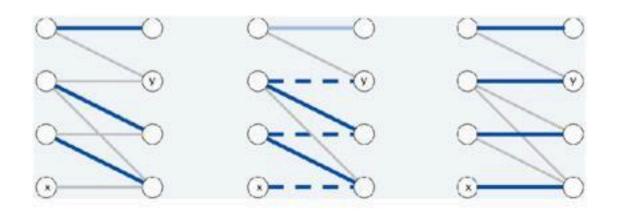
#### Augmenting path simplifies to:

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

# Alternating path

**Def**. An **alternating path** P with respect to a matching M is an alternating sequence of unmatched and matched edges, starting from an unmatched node  $x \in X$  and going to an unmatched node  $y \in Y$ .

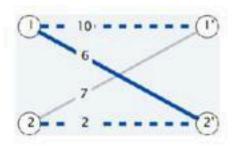
**Key property**. Can use P to increase by one the cardinality of the matching. **Pf**. Set  $M' = M \ominus P$ .





# Successive shortest path

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



$$P=2
ightarrow 2'
ightarrow 1
ightarrow 1'$$

$$cost(P) = 2 - 6 + 10 = 6$$

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

#### Successive shortest path algorithm.

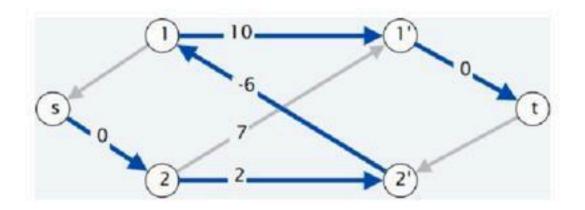
- Start with empty matching.
- Repeatedly augment along a shortest alternating path.

# Demo: Successive shortest path algorithm



# Finding the shortest alternating path

**Shortest alternating path**. Corresponds to minimum cost  $s \sim t$  path in  $G_M$ .



Concern. Edge costs can be negative.

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

Our plan. Use duality to avoid negative edge costs (and negative cycles) ⇒ can compute using Dijkstra.

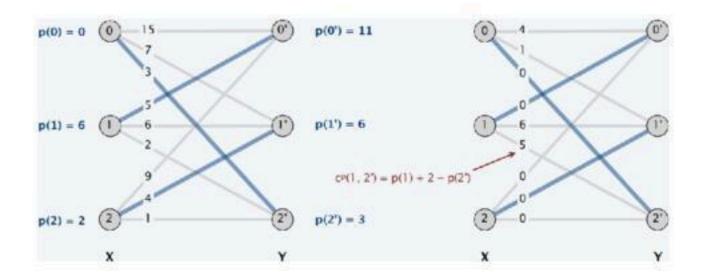
# Equivalent assignment problem

**Duality intuition**. Adding a constant p(x) to the cost of every edge incident to node  $x \in X$  does not change the min-cost perfect matching(s).

**Pf**. Every perfect matching uses exactly one edge incident to node x.

**Duality intuition**. Adding a constant p(y) to the cost of every edge incident to node  $y \in Y$  does not change the min-cost perfect matching(s).

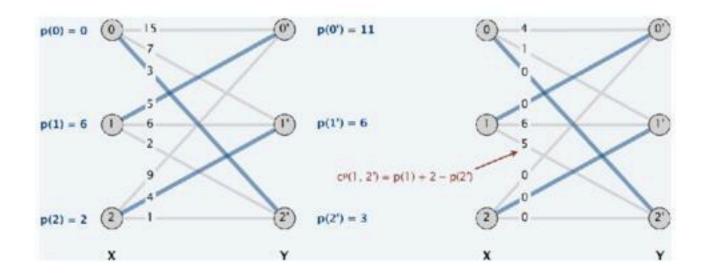
**Pf**. Every perfect matching uses exactly one edge incident to node y.



### 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.



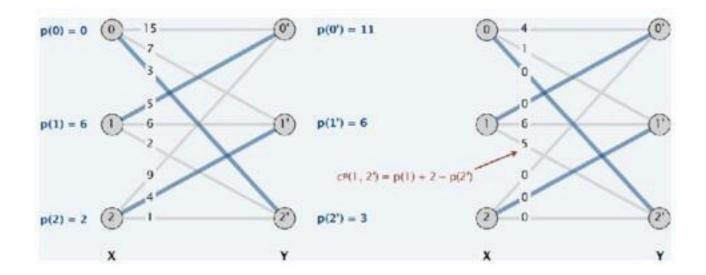
# Compatible prices

**Compatible prices**. For each node  $v \in X \cup Y$ , maintain prices p(v) such that:

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

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

**Pf**. Matching *M* has 0 cost.



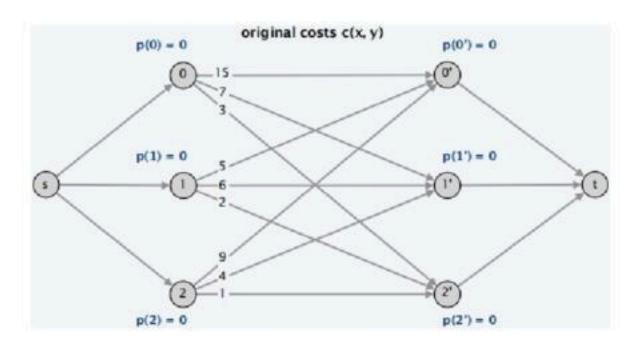
# Successive shortest path: algorithm

SUCCESSIVE-SHORTEST-PATH (X,Y,c)

- 1.  $M = \emptyset$ ;
- 2. FOREACH  $v \in X \cup Y$ : p(v) = 0;
- 3. WHILE (M is not a perfect matching)
  - 1. d = shortest path distances using costs  $c^p$ ;
  - P = shortest alternating path using costs c^p;
  - 3. M = updated matching after augmenting along P;
  - 4. FOREACH  $v \in X \cup Y$ : p(v) = p(v) + d(v); RETURN M;



# Successive shortest path: demo



### Maintaining compatible prices 1

**Lemma 1**. Let p be compatible prices for 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,  $c^{p+d}(x,y) = 0$ .

# Maintaining compatible prices 2

**Lemma 2**. Let p be compatible prices for 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,  $c^{p+d}(x,y) \geq 0$ .

### Maintaining compatible prices 3

**Lemma 3**. Let p be compatible prices for M and let M' be matching obtained by augmenting along a min cost path with respect to  $c^{p+d}$ . Then p'=p+d are compatible prices for 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 min-cost path.
- By LEMMA 1, these edges satisfy  $c^{p+d}(x,y)=0$ .
- Thus, compatibility is maintained.

# Successive shortest path: analysis

**Invariant**. The algorithm maintains a matching M and compatible prices p. **Pf**. Follows from LEMMA 2 and LEMMA 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.



# Weighted bipartite matching

Weighted bipartite matching. Given a weighted bipartite graph with n nodes and m edges, find a maximum cardinality matching of minimum weight.

**Theorem**. [Fredman-Tarjan 1987] The successive shortest path algorithm solves the problem in  $O(n^2 + mn \log n)$  time using Fibonacci heaps.

**Theorem**. [Gabow-Tarjan 1989] There exists an  $O(mn^{1/2}\log(nC))$  time algorithm for the problem when the costs are integers between 0 and C.

# Input-queued switching

# Input-queued switching Problem

### Input-queued switch.

- n input ports and n output ports 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.

**Application**. High-bandwidth switches.

