

# **Data Structures and Algorithms**

**Session 19. April 6, 2009**

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# Announcements

- \* Homework 4 due by midnight tonight
- \* Homework 5 assigned

# Review

- \* Topological Sort
- \* Shortest Path
  - \* Unweighted version: Breadth-first search
  - \* Weighted version: Dijkstra's Algorithm

# Today's Plan

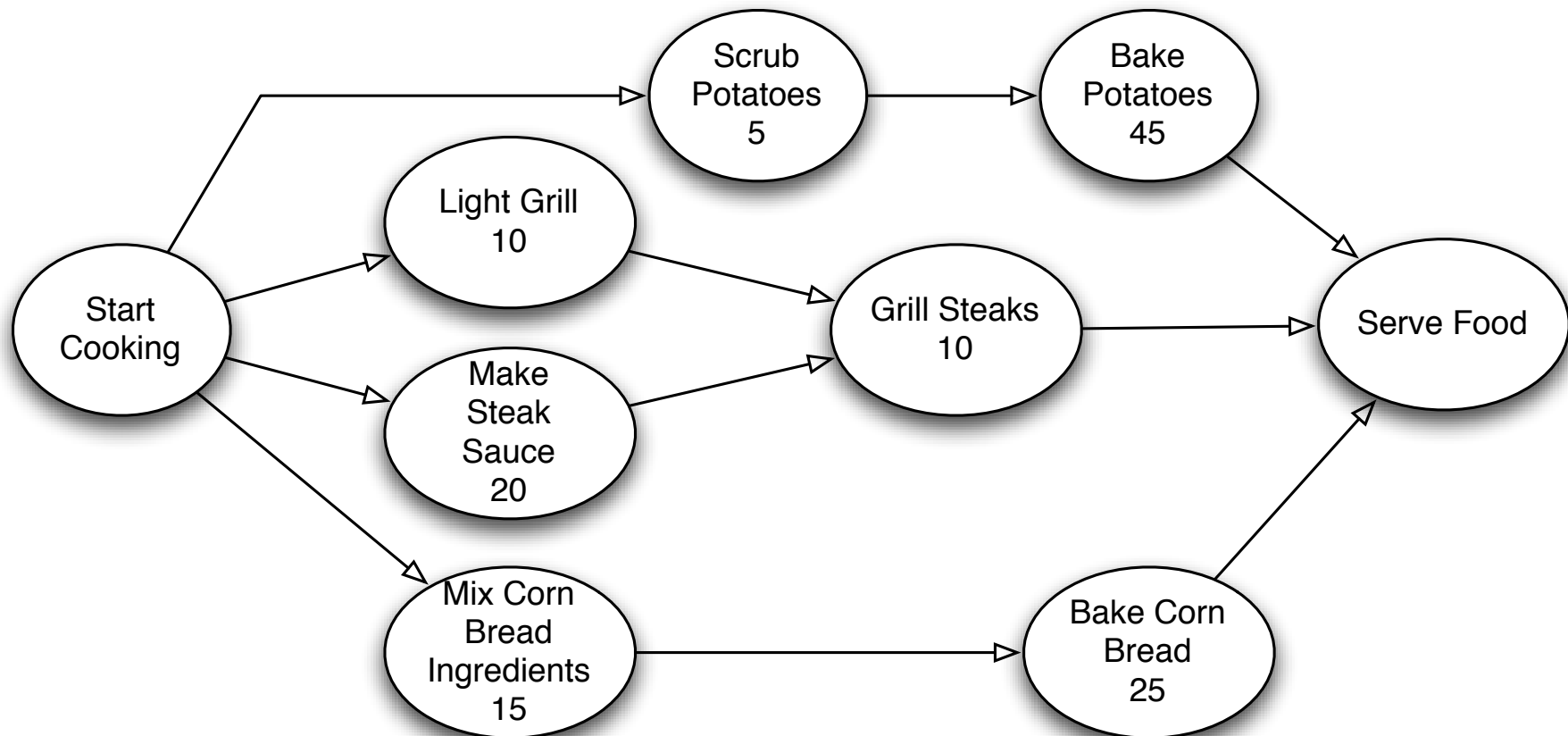
- \* Extensions of Dijkstra's Algorithm
  - \* Critical Path Analysis
  - \* All Pairs Shortest Path (Floyd-Warshall)
- \* Maximum Flow
  - \* Floyd-Fulkerson Algorithm

# Critical Path Analysis

- \* Recall motivational example for topological sort: edges represent dependencies between tasks
- \* Consider a similar **event-node graph** in which nodes represent events and edges represent dependencies and costs
- \* We want to find the fastest time we can complete all tasks if we can run job in parallel

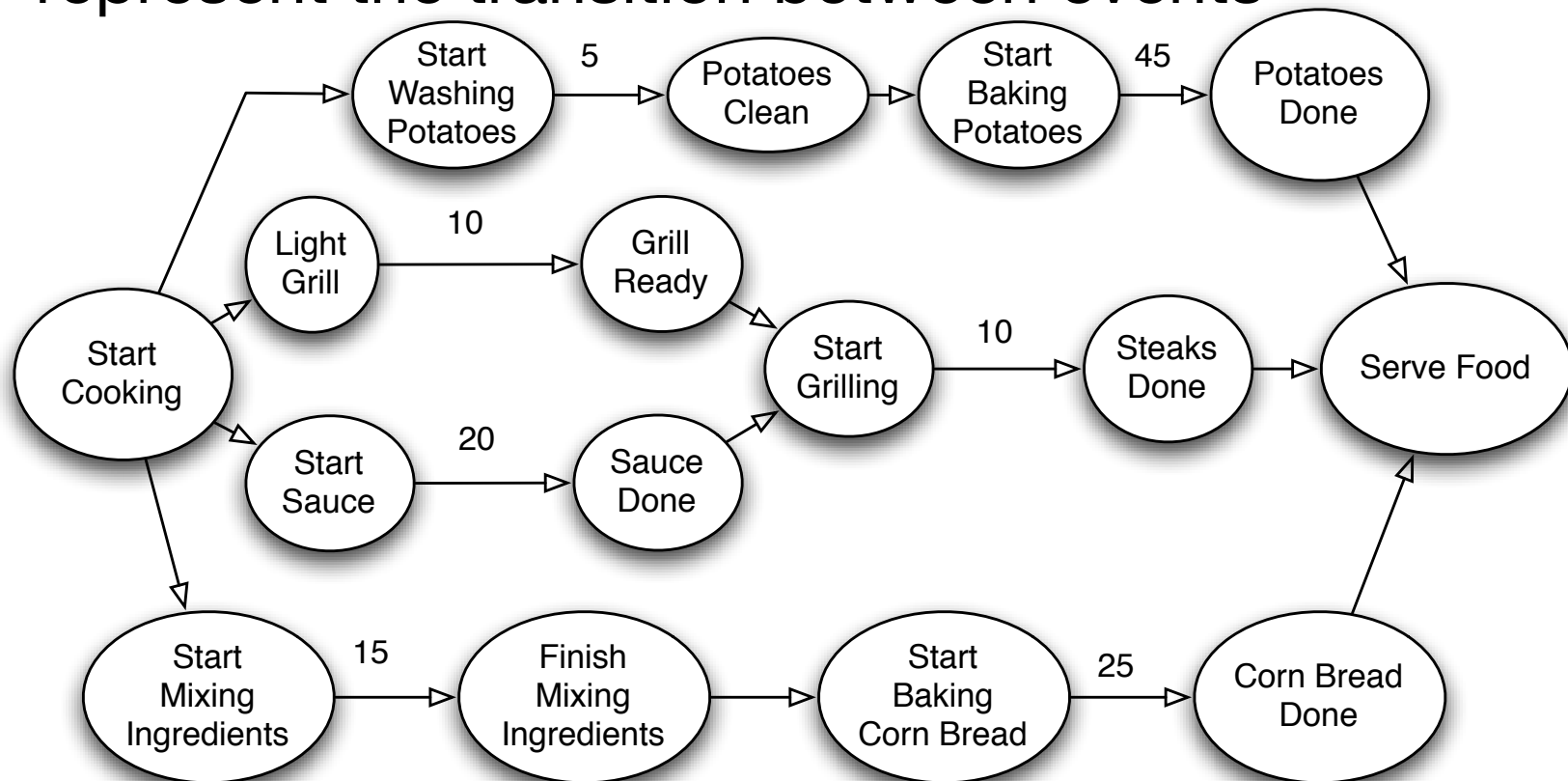
# Critical Path Example

- ✦ We start with the activity graph, which includes time for each activity



# Critical Path Example

- Convert it to an event-node graph, where edges represent the transition between events

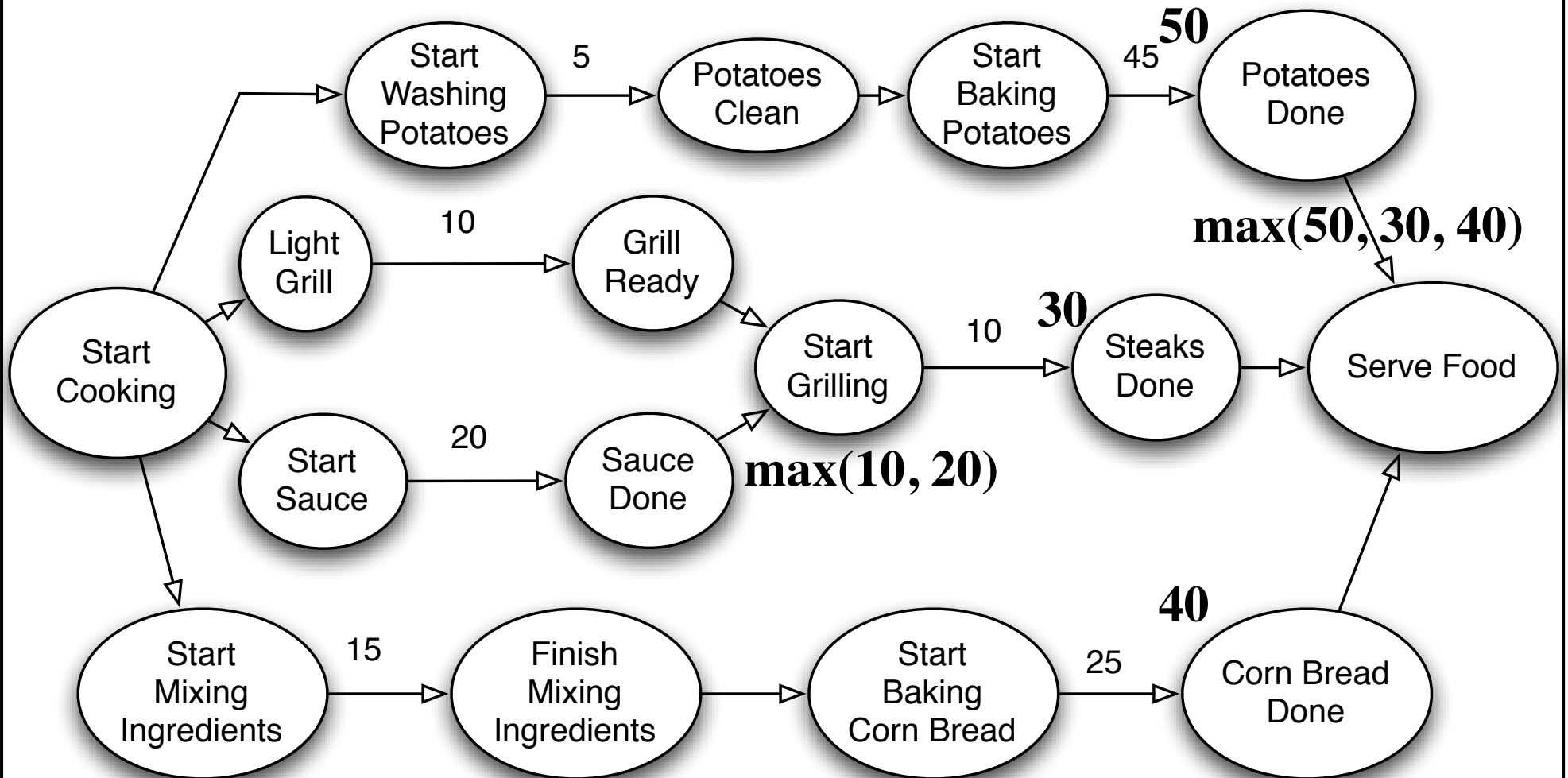


# Longest Path in a DAG

- \* Store “longest known path” for each node
- \* Start node = 0
- \* Max of incoming nodes’ longest known path + incoming edge cost
- \* Longest path from start to end is critical path



# Critical Path for BBQ



# Latest Completion Time

- \* If you want to procrastinate, compute the latest you can finish each job without delaying total time
- \* Set time of end node to critical path time
- \* Set nodes' latest completion time to:  
min of (outgoing node time) - (outgoing edge cost)
- \* (Similar to finding the shortest path to end following edges backwards)

# All Pairs Shortest Path

- \* Dijkstra's Algorithm finds shortest paths from one node to all other nodes
- \* What about computing shortest paths for all pairs of nodes?
- \* We can run Dijkstra's  $|V|$  times. Total cost:  $O(|V|^3)$
- \* Floyd-Warshall algorithm is often faster in practice (though same asymptotic time)

# Recursive Motivation

- \* Consider the set of numbered nodes **1** through **k**
- \* The shortest path between any node **i** and **j** using only nodes in the set **{1, ..., k}** is the minimum of
  - \* shortest path from **i** to **j** using nodes **{1, ..., k-1}**
  - \* shortest path from **i** to **j** using node **k**
- \*  $\text{path}(i,j,k) = \min( \text{path}(i,j,k-1), \text{path}(i,k,k-1) + \text{path}(k,j,k-1) )$

# Dynamic Programming

- \* Instead of repeatedly computing recursive calls, store lookup table
- \* To compute  $\text{path}(i,j,k)$  for any  $i,j$ , we only need to look up  $\text{path}(-,-, k-1)$ 
  - \* but never  $k-2, k-3, \text{etc.}$
- \* We can incrementally compute the path matrix for  $k=0$ , then use it to compute for  $k=1$ , then  $k=2\dots$

# Floyd-Warshall Code

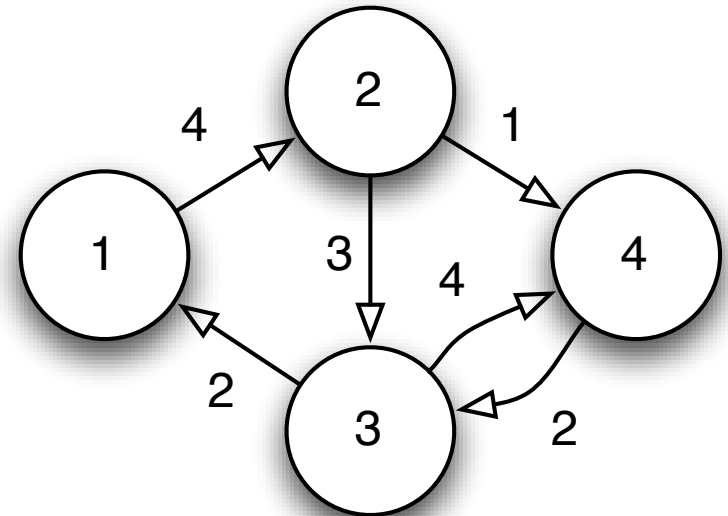
- \* Initialize  $d$  = weight matrix
- \* 

```
for (k=0; k<N; k++)  
  for (i=0; i<N; i++)  
    for (j=0; j<N; j++)  
      if (d[i][j] > d[i][k]+d[k][j])  
        d[i][j] = d[i][k] + d[k][j];
```
- \* Additionally, we can store the actual path by keeping a “midpoint” matrix

# All Pairs Shortest Path Example

**K=0**

	1	2	3	4
1	-	4	-	-
2	-	-	3	1
3	2	-	-	4
4	-	-	2	-



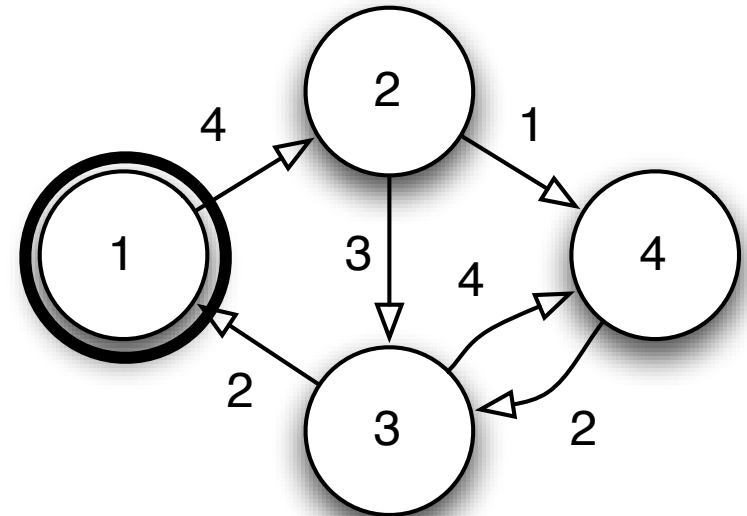
# All Pairs Shortest Path Example

**K=0**

	1	2	3	4
1	-	4	-	-
2	-	-	3	1
3	2	-	-	4
4	-	-	2	-

**K=1**

	1	2	3	4
1	-	4	-	-
2	-	-	3	1
3	2	6	-	4
4	-	-	2	-





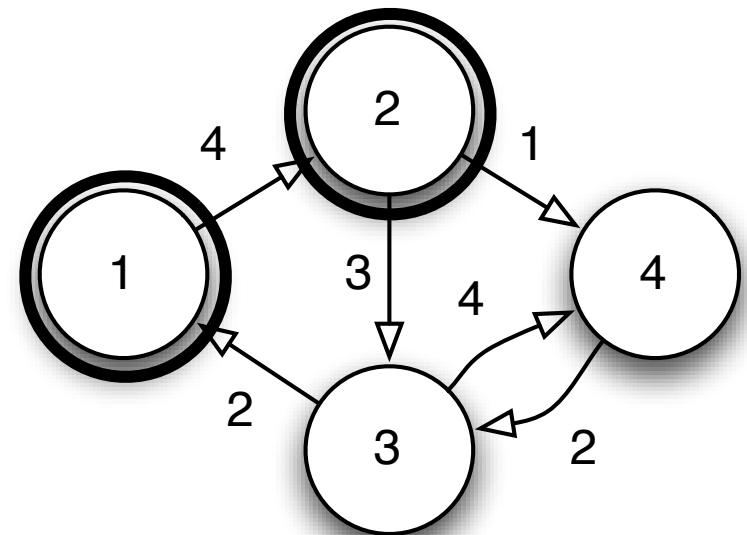
# All Pairs Shortest Path Example

**K=1**

	1	2	3	4
1	-	4	-	-
2	-	-	3	1
3	2	6	-	4
4	-	-	2	-

**K=2**

	1	2	3	4
1	-	4	7	5
2	-	-	3	1
3	2	6	9	4
4	-	-	2	-



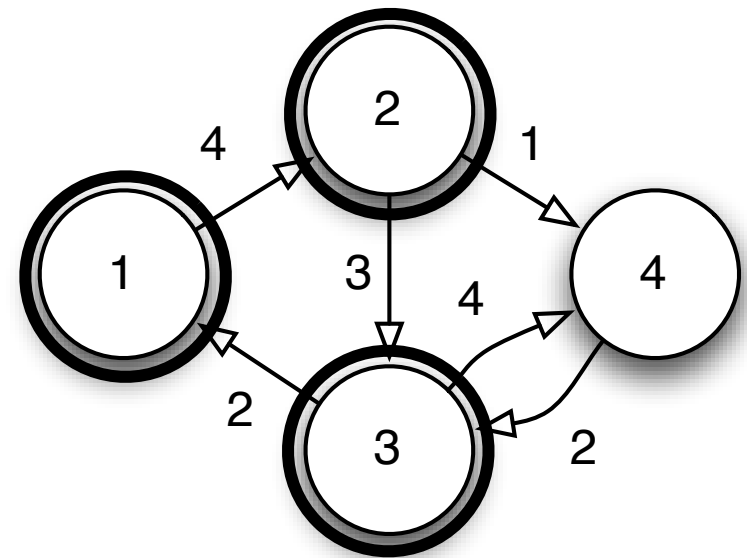
# All Pairs Shortest Path Example

**K=2**

	1	2	3	4
1	-	4	7	5
2	-	-	3	1
3	2	6	9	4
4	-	-	2	-

**K=3**

	1	2	3	4
1	9	4	7	5
2	5	9	3	1
3	2	6	9	4
4	4	8	2	6



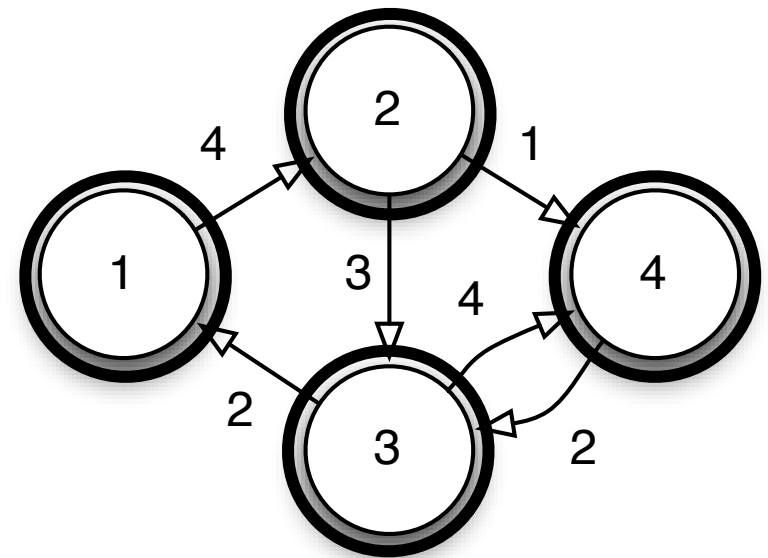
# All Pairs Shortest Path Example

**K=3**

	1	2	3	4
1	9	4	7	5
2	5	9	3	1
3	2	6	9	4
4	4	8	2	6

**K=4**

	1	2	3	4
1	9	4	7	5
2	5	9	3	1
3	2	6	6	4
4	4	8	2	6



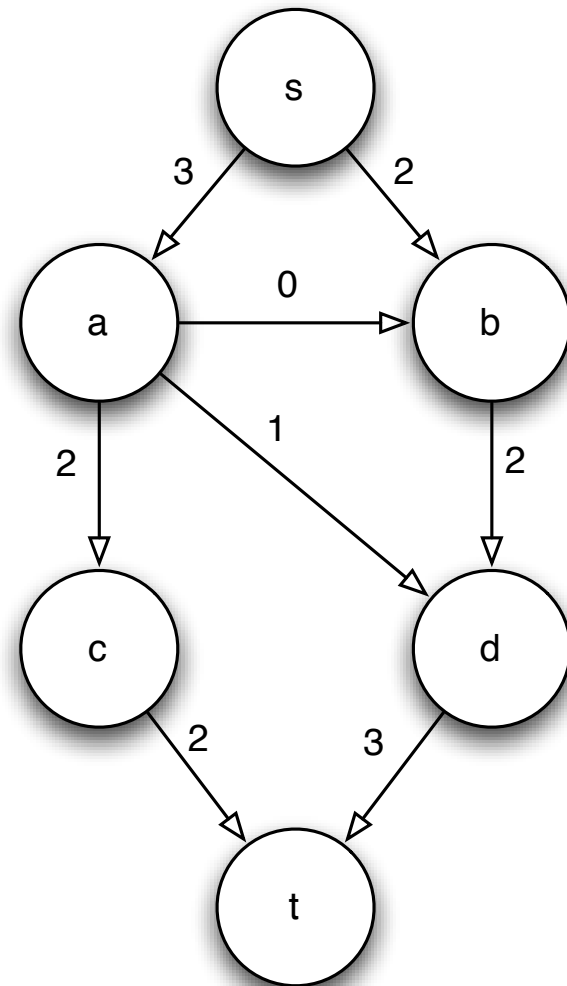
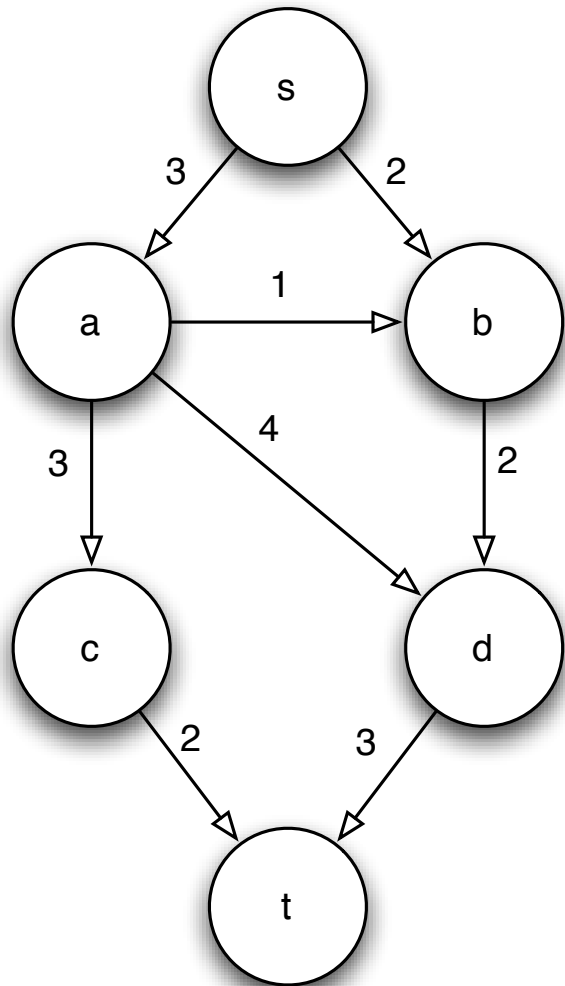
# Transitive Closure

- \* For any nodes  $i, j$ , is there a path from  $i$  to  $j$ ?
- \* Instead of computing shortest paths, just compute Boolean if a path exists
- \*  $\text{path}(i,j,k) = \text{path}(i,j,k-1) \text{ OR}$   
 $\text{path}(i,k,k-1) \text{ AND path}(k,j,k-1)$

# Maximum Flow

- \* Consider a graph representing flow capacity
- \* Directed graph with **source** and **sink** nodes
- \* Physical analogy: water pipes
  - \* Each edge weight represents the **capacity**: how much “water” can run through the pipe from source to sink?

# Capacity Example



**MAXIMUM FLOW SOLUTION**

# Max Flow Algorithm

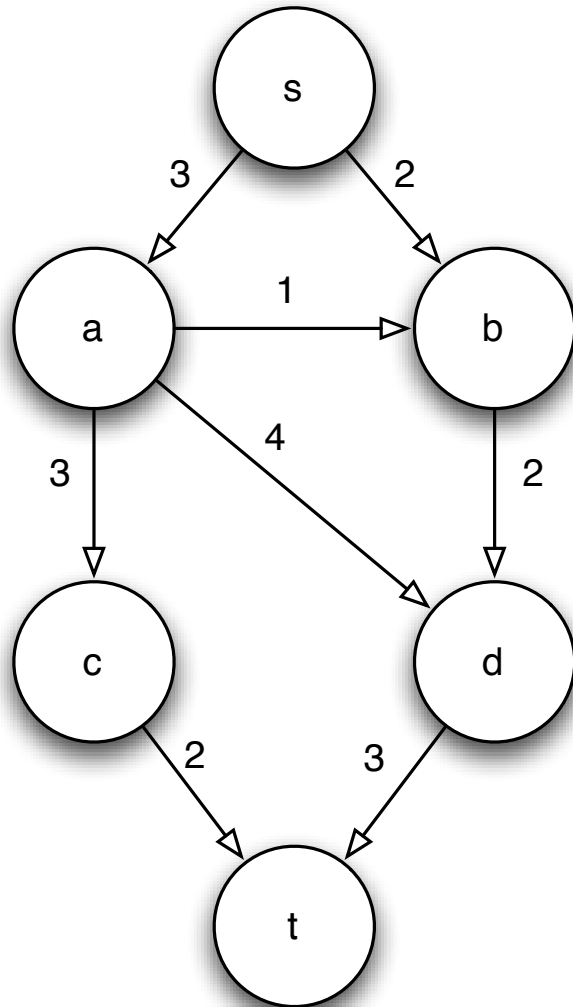
- \* Create 2 copies of original graph: **flow graph** and **residual graph**
- \* The flow graph tells us how much flow we have currently on each edge
- \* The residual graph tells us how much flow is available on each edge
- \* Initially, the residual graph is the original graph

# Augmenting Path

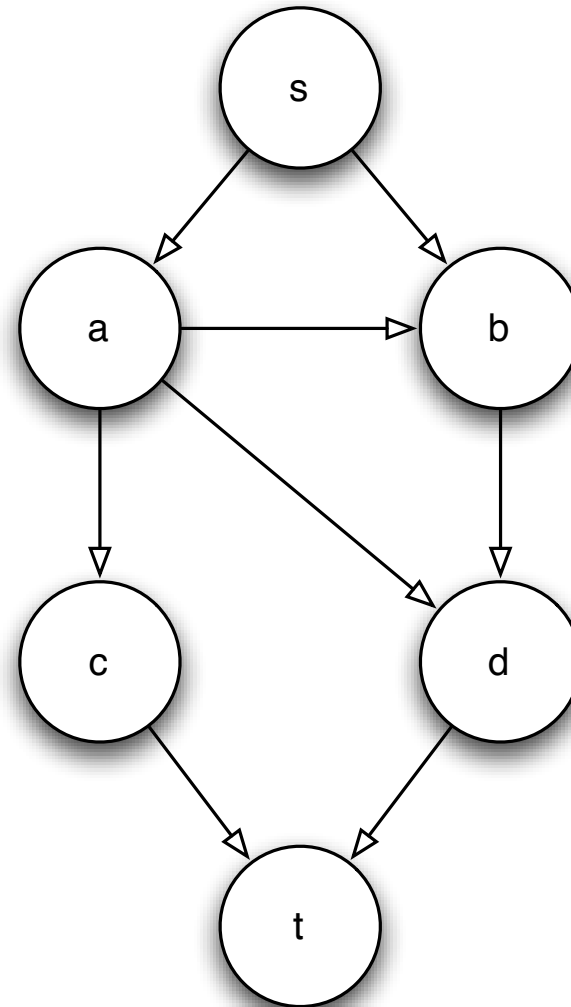
- \* Find any path in residual graph from source to sink
  - \* called an **augmenting path**.
- \* The minimum weight along path can be added as flow to the flow graph
- \* But we don't want to commit to this flow; add a reverse-direction undo edge to the residual graph



# Example

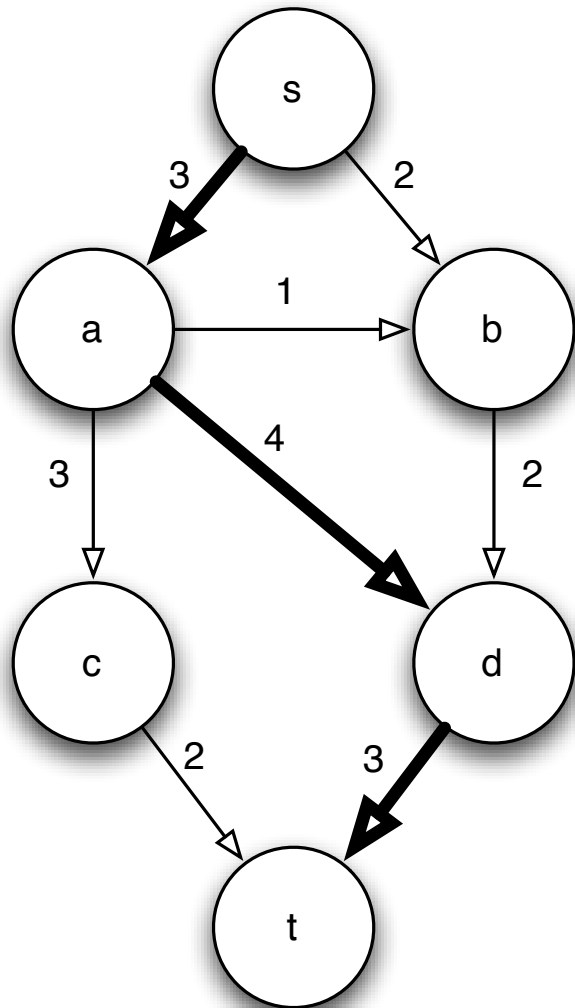


**RESIDUAL**

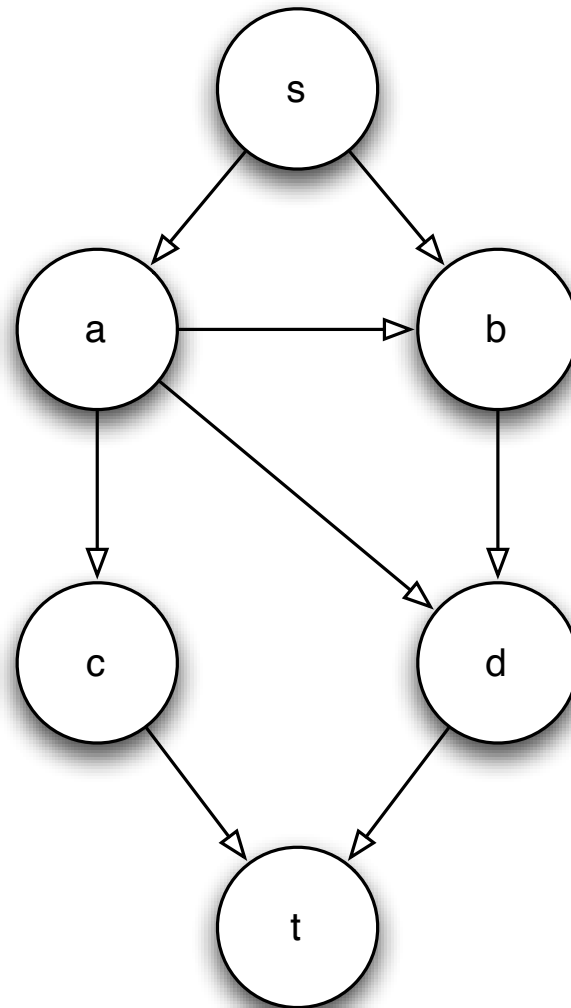


**FLOW**

# Example

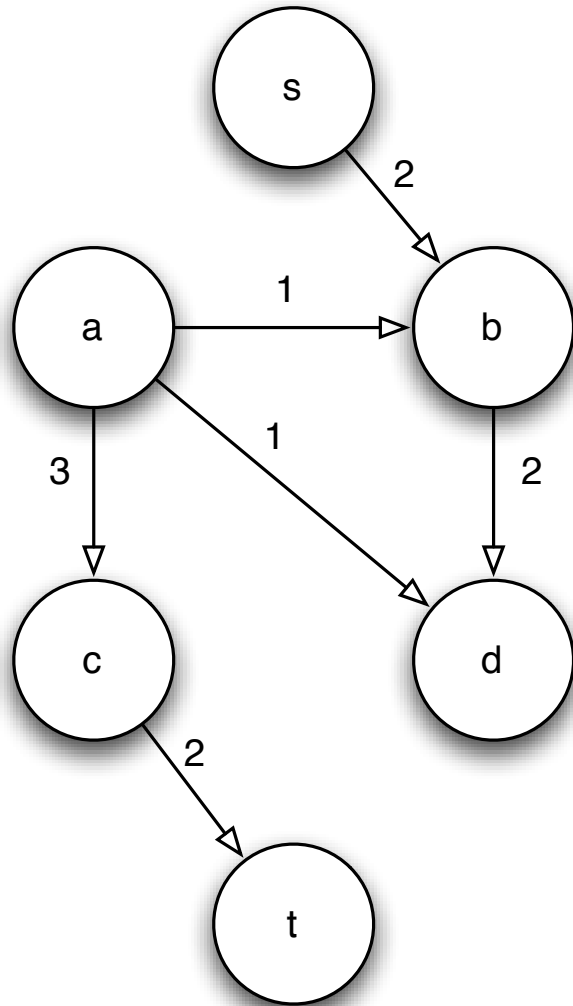


**RESIDUAL**

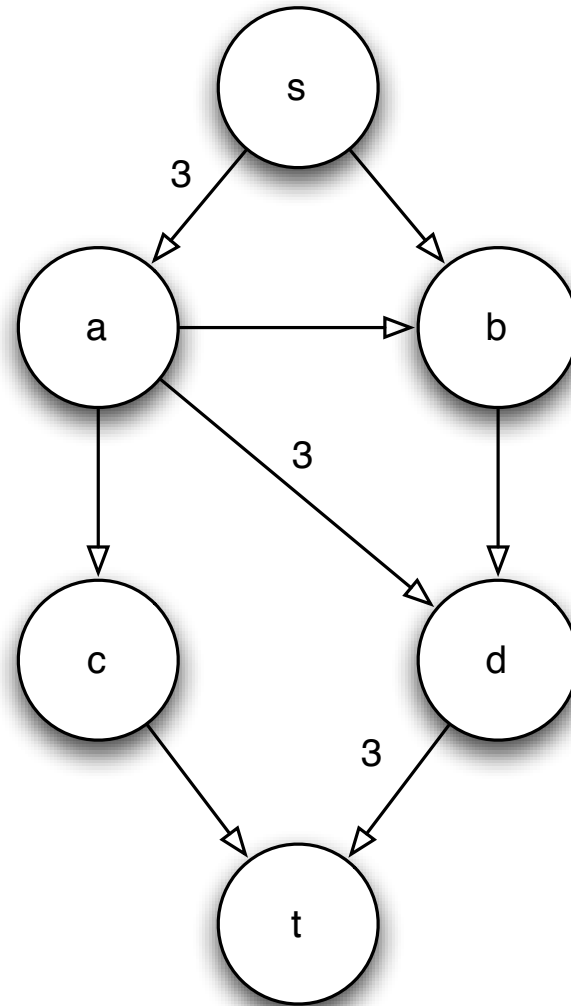


**FLOW**

# Example

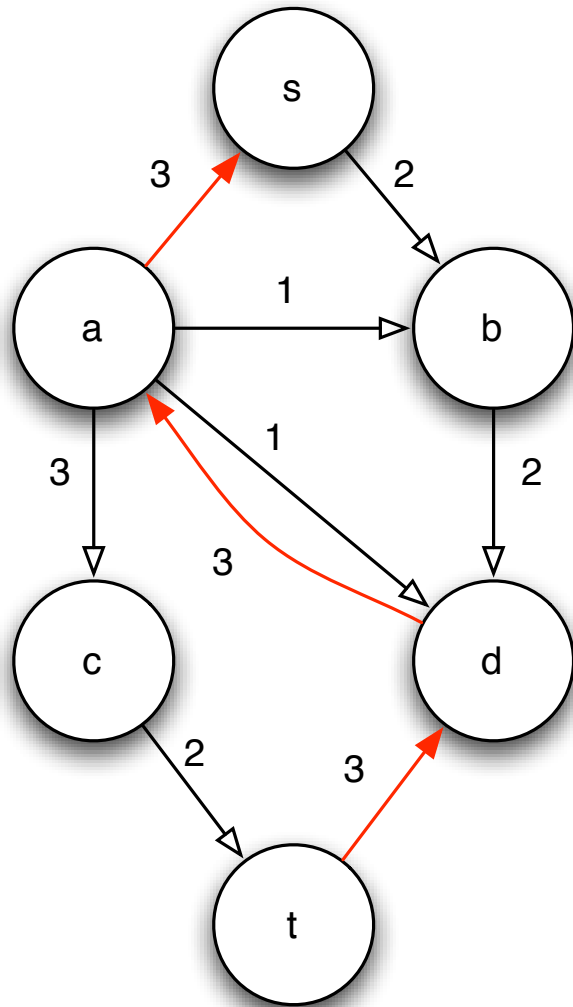


**RESIDUAL**

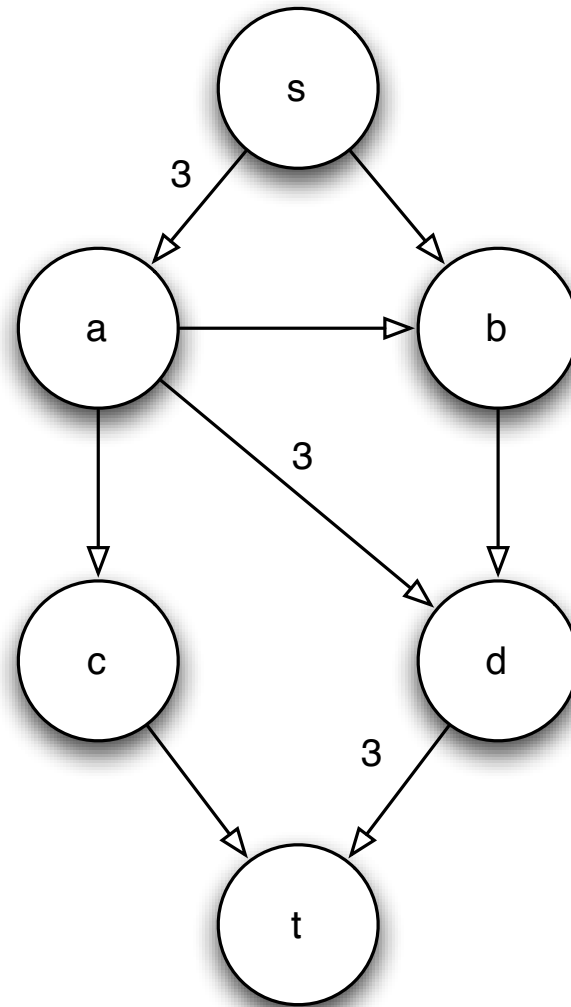


**FLOW**

# Example

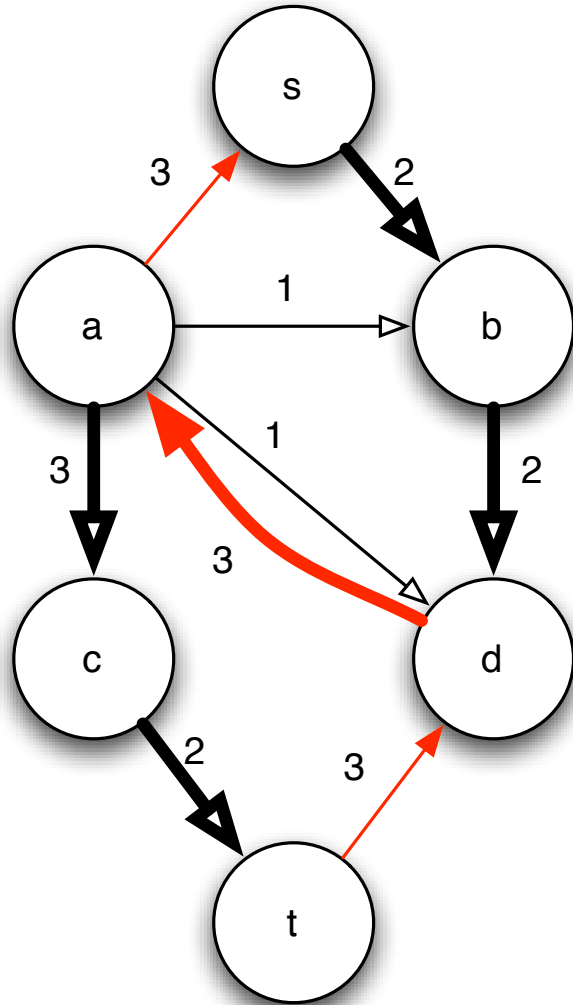


**RESIDUAL**

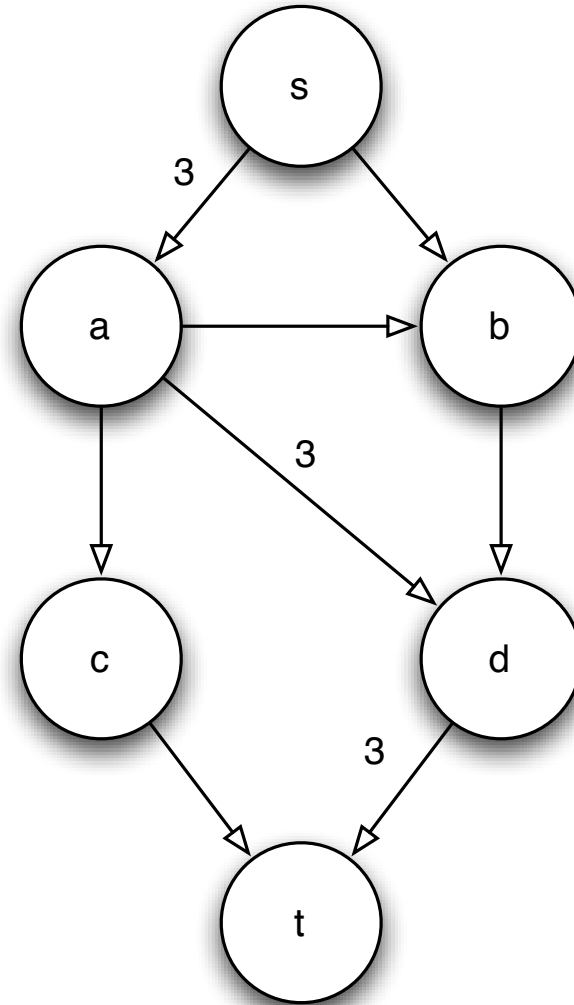


**FLOW**

# Example

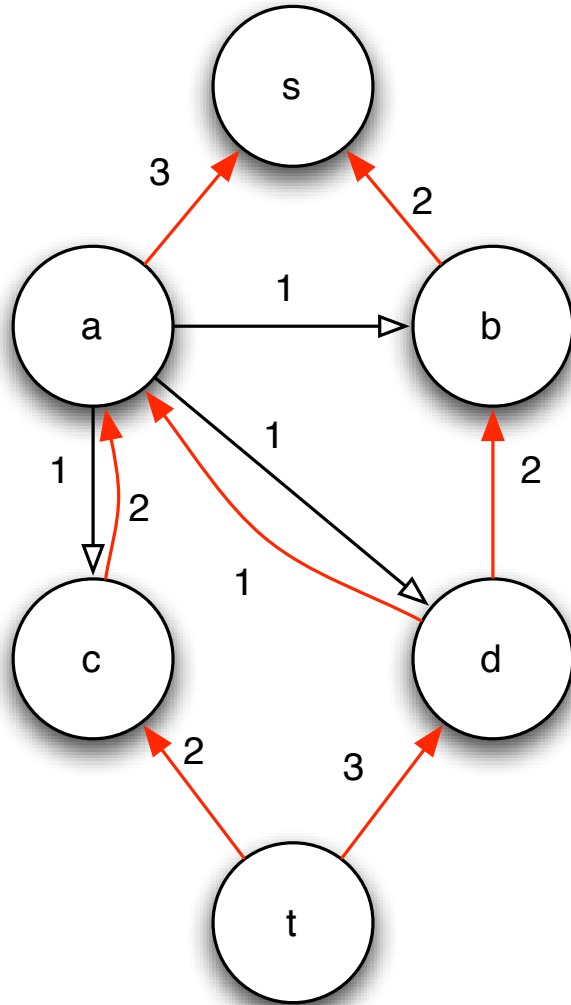


**RESIDUAL**

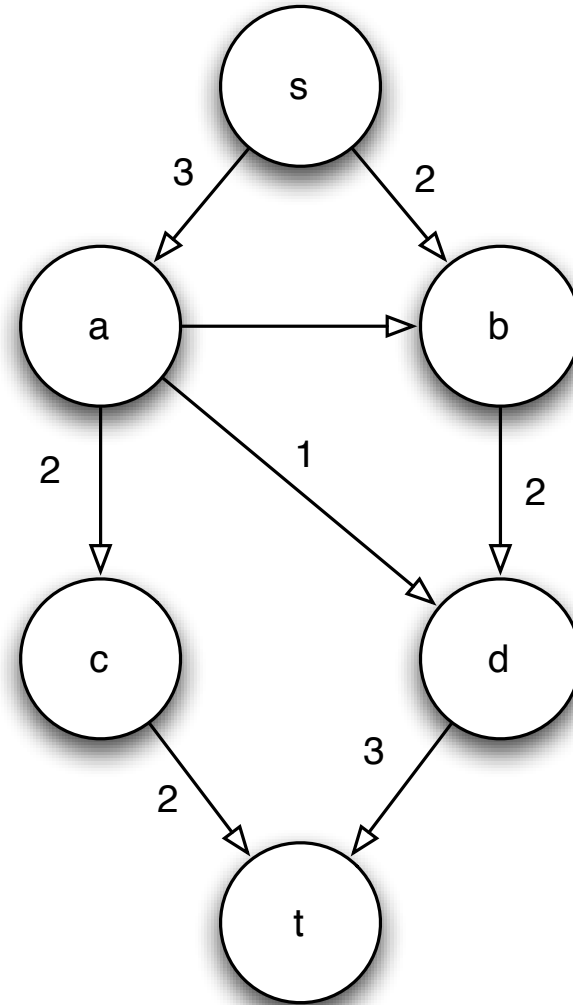


**FLOW**

# Example



**RESIDUAL**



**FLOW**

# Running Times

- \* If integer weights, each augmenting path increases flow by at least 1
- \* Costs  $O(|E|)$  to find an augmenting path
- \* For max flow  $f$ , finding max flow (Floyd-Fulkerson) costs  $O(f|E|)$
- \* Choosing shortest unweighted path (Edmonds-Karp),  $O(|V||E|^2)$

# Sports Elimination

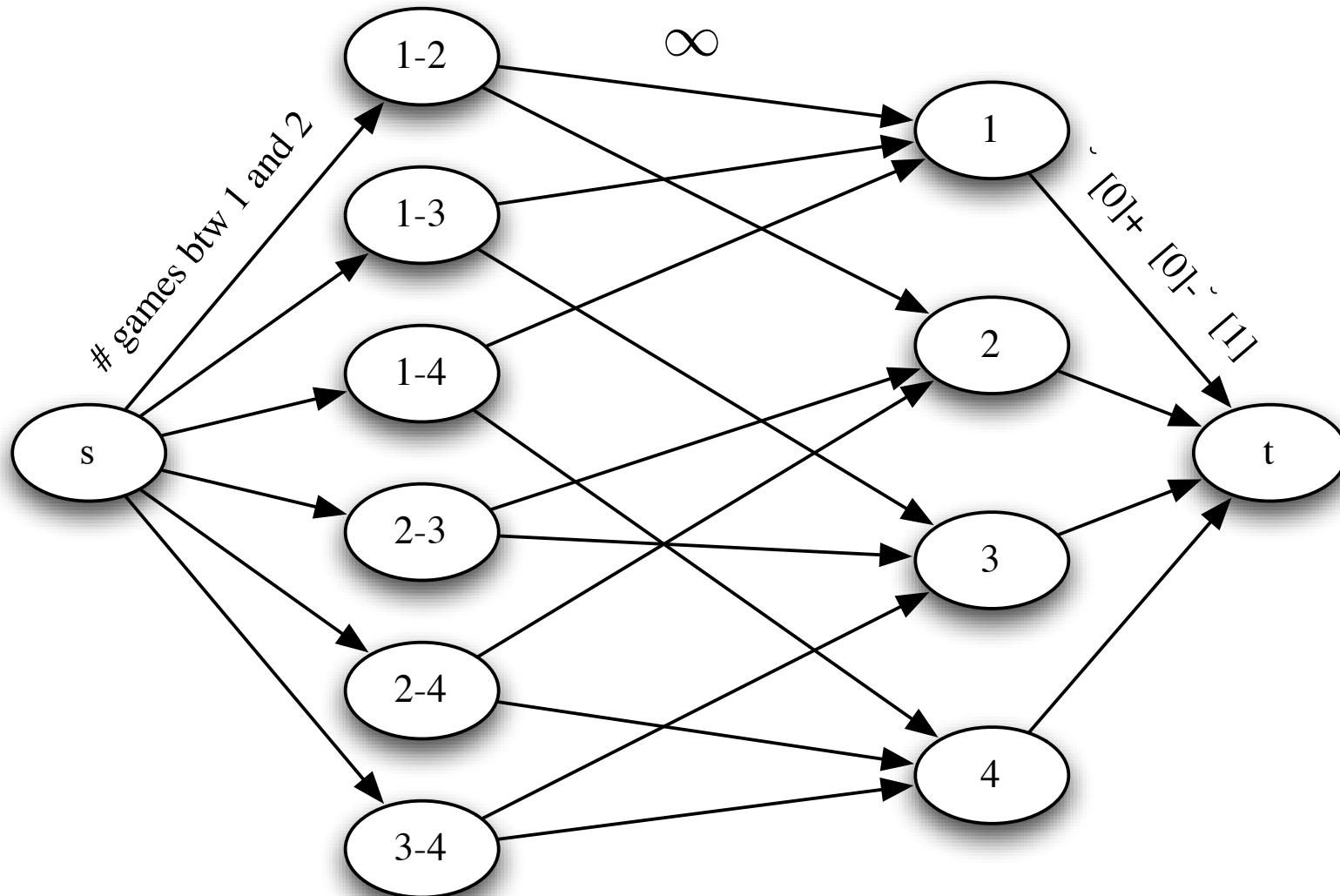
- \* In many organized sports, teams are split into divisions
  - \* the team in a division with the most wins at end of season earns a divisional title
- \* Fans and writers like to talk about whether a team is mathematically eliminated from the division race
- \* The standard formula is often wrong, instead, compute a max flow



# Standard Formula

- \* If team  $i$  has  $W[i]$  wins, and  $R[i]$  remaining games, pretend  $i$  wins all of its  $R[i]$  games.  $W[i]+R[i]$
- \* Pretend all other teams in division win no more games. If  $W[i]+R[i] > W[j]$ , for all  $j$ ,  $i$  can still win
- \* The problem is the other teams may have games against each other; both teams can't lose

# Max Flow Graph



# Max Flow Solution: team $i$

- \* Connect source to all game nodes (team  $j$ , team  $k$ )
  - \* Capacity of edge to game node is # of games btw  $j$  and  $k$
- \* Connect game nodes to participating team nodes with infinite capacity
- \* Connect team nodes to sink,  
capacity = # of games before team  $j$  overtakes team  $i$
- \* Team  $i$  can win only if max flow saturates outgoing edges from source

# Reading

\* Weiss Section 9.5