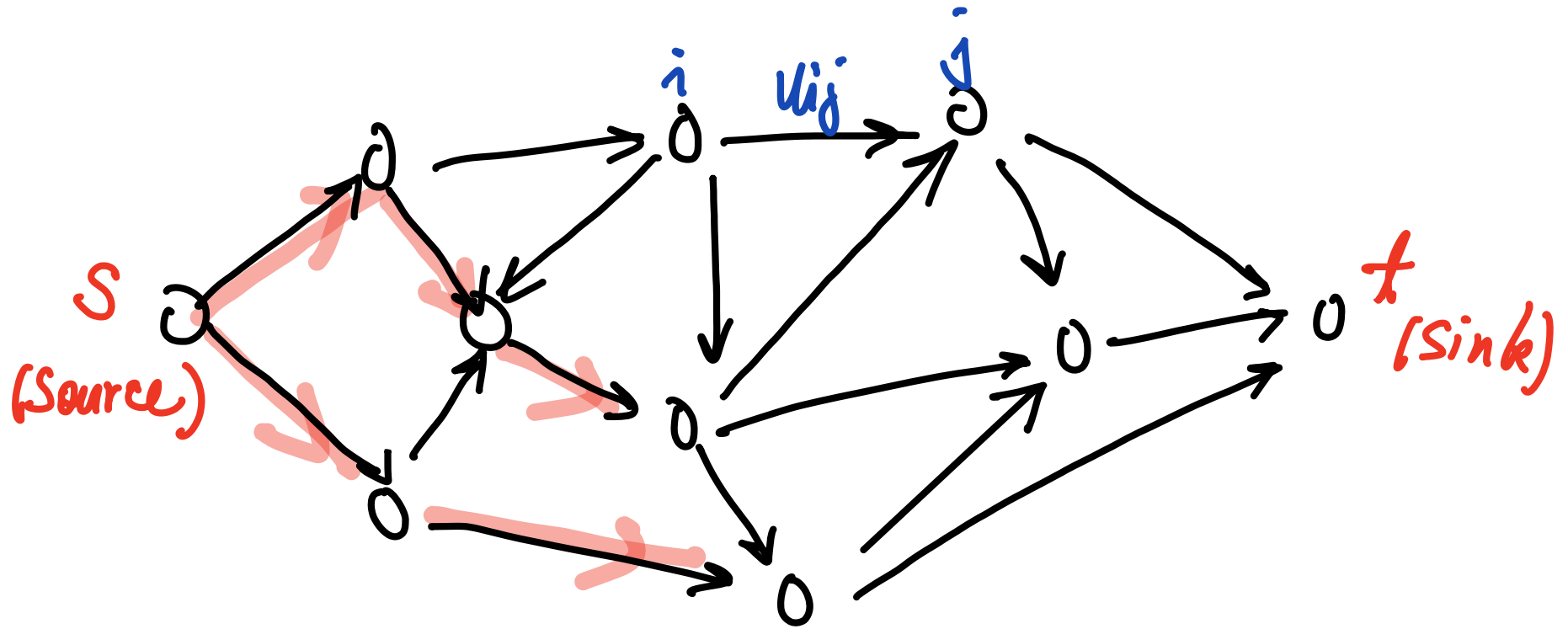


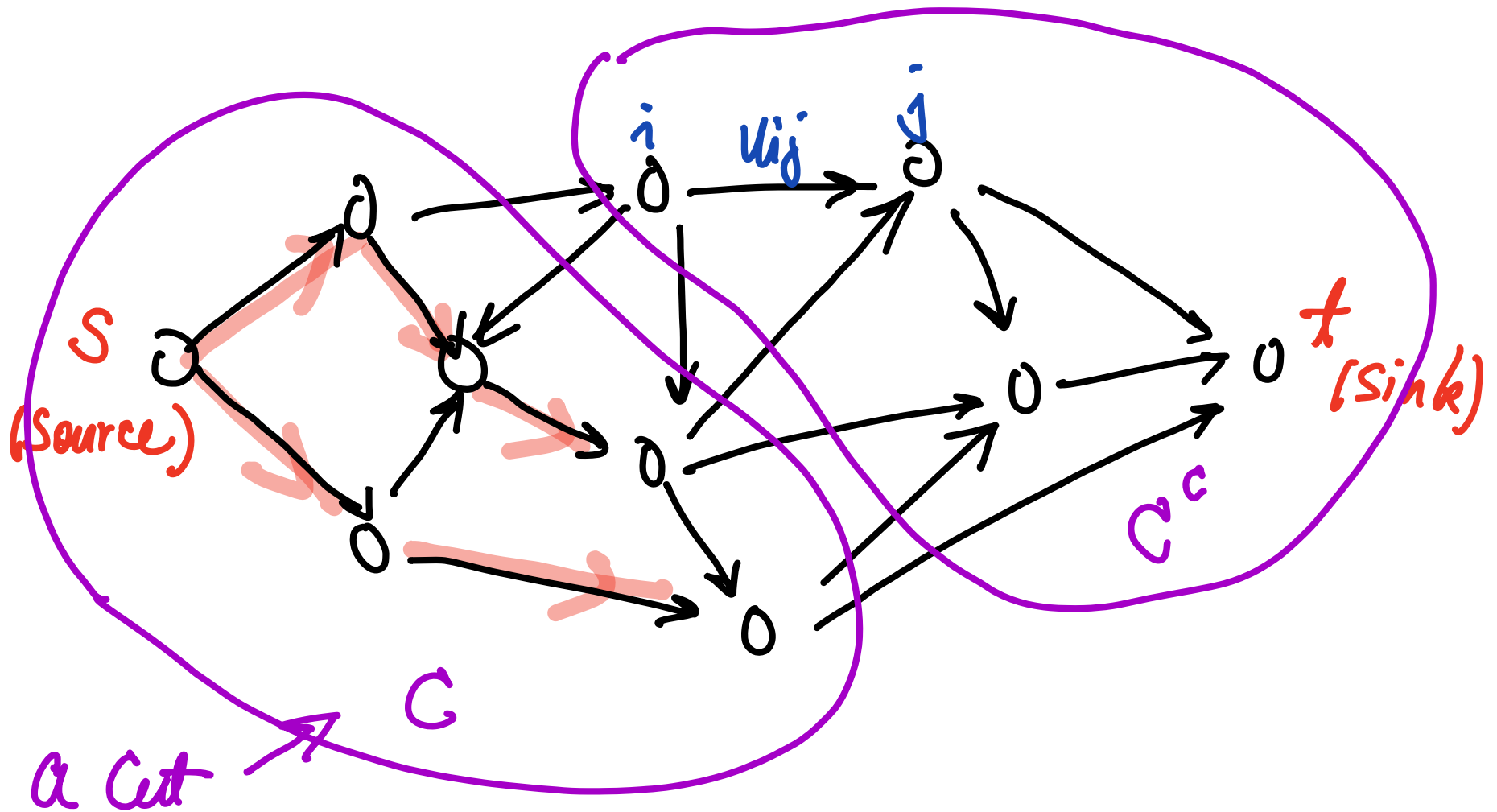
# Max Flow Min Cut



What is the max. flow from s to t subject to

$$0 \leq x_{ij} \leq u_{ij} \quad ?$$

# Max Flow Min Cut

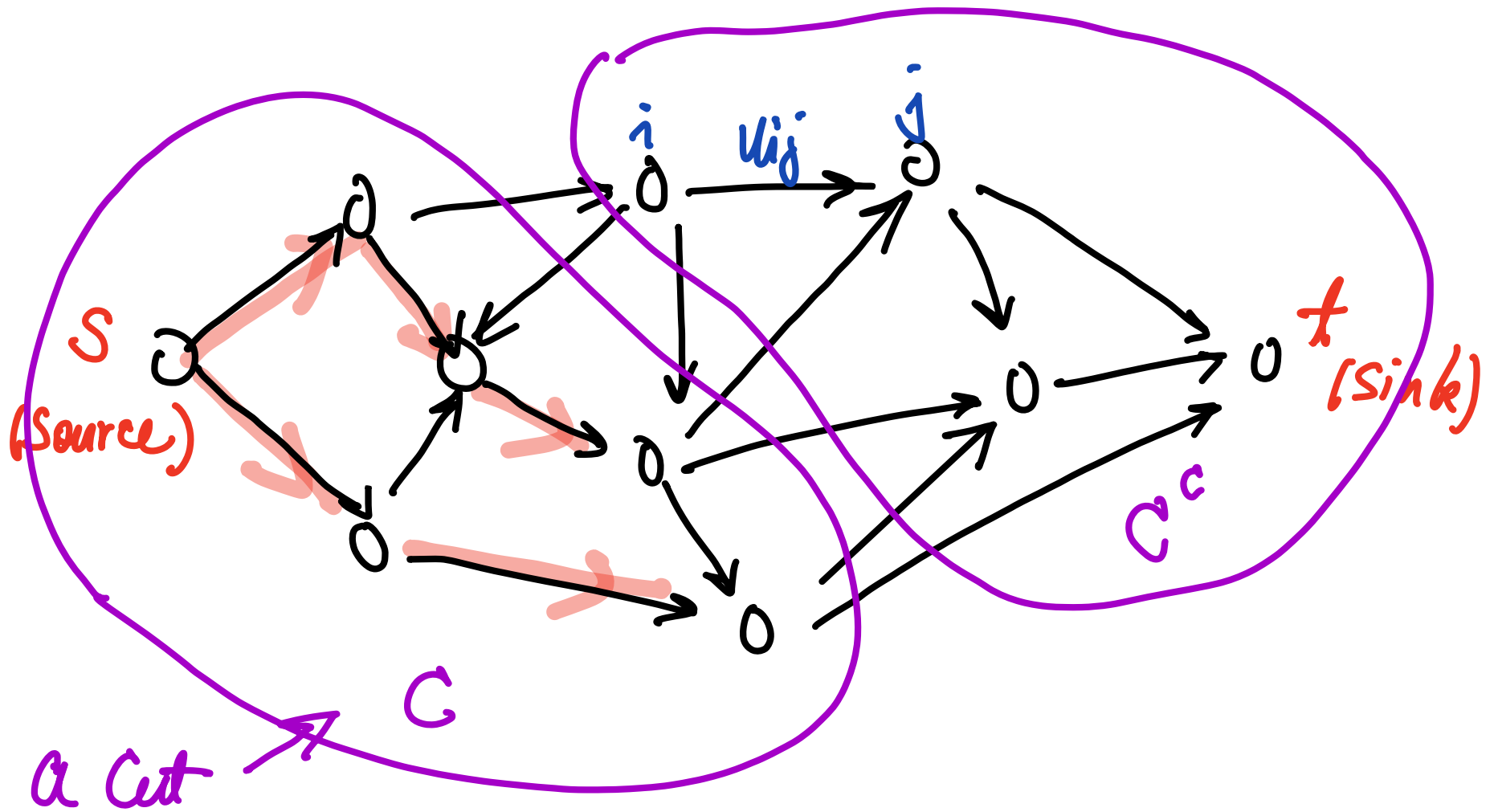


$$N = C \cup C^c, \quad s \in C \quad (t \in C^c)$$

$$K(C) = \sum_{i \in C, j \notin C} u_{ij}$$

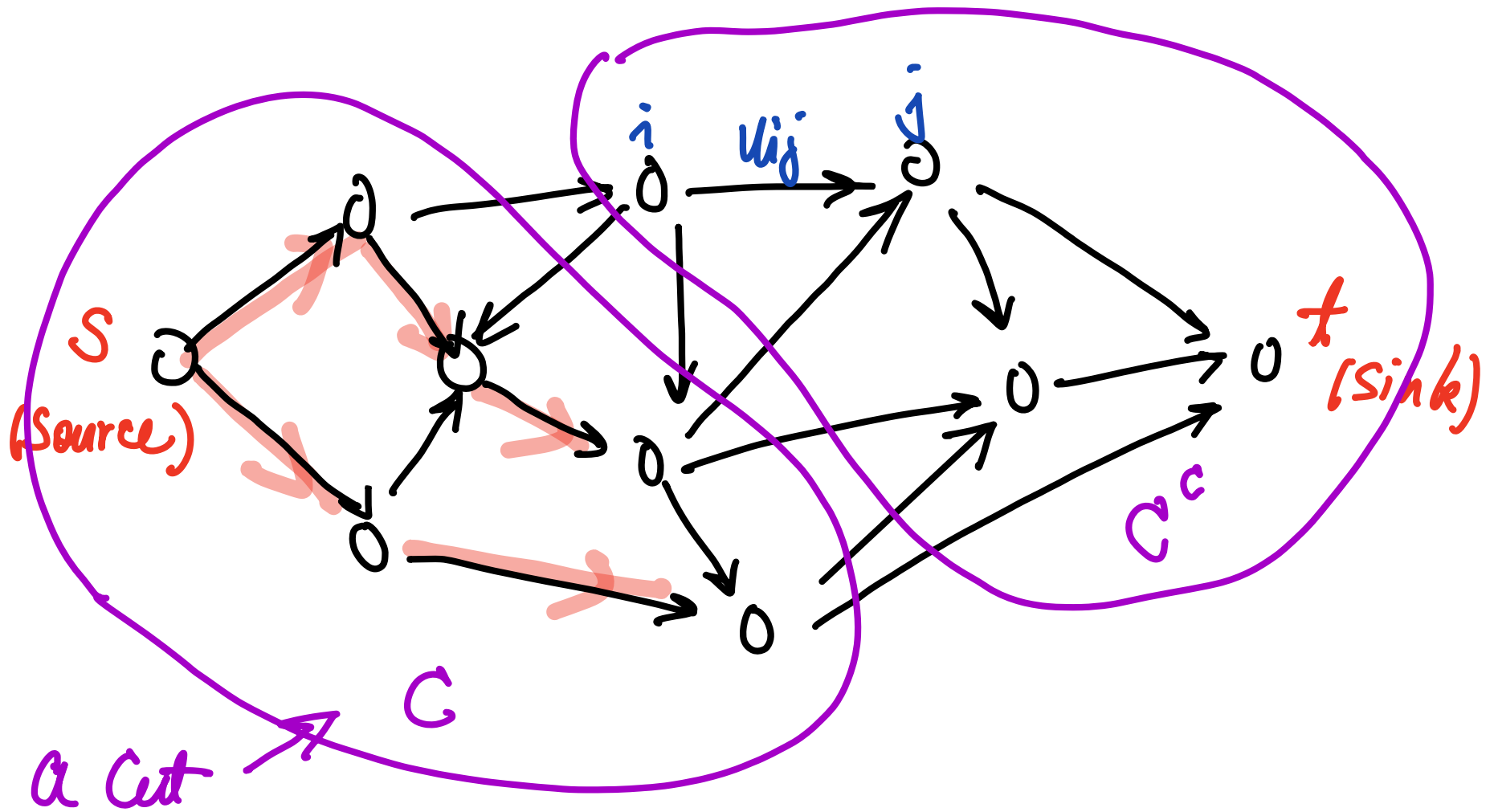
capacity of the cut  $C$ .

# Max Flow Min Cut



$$(\text{any}) \text{ Flow } (s \rightarrow t) \leq (\text{any}) K(C)$$

# Max Flow Min Cut



$$\text{max} \text{ Flow}(s \rightarrow t) \leq \text{min} K(C)$$

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Table 22.1 The Travelers' Example:  
Seat Availability

From	To	Number of seats
San Francisco	Denver	5
San Francisco	Houston	6
Denver	Atlanta	4
Denver	Chicago	2
Houston	Atlanta	5
Atlanta	New York	7
Chicago	New York	4

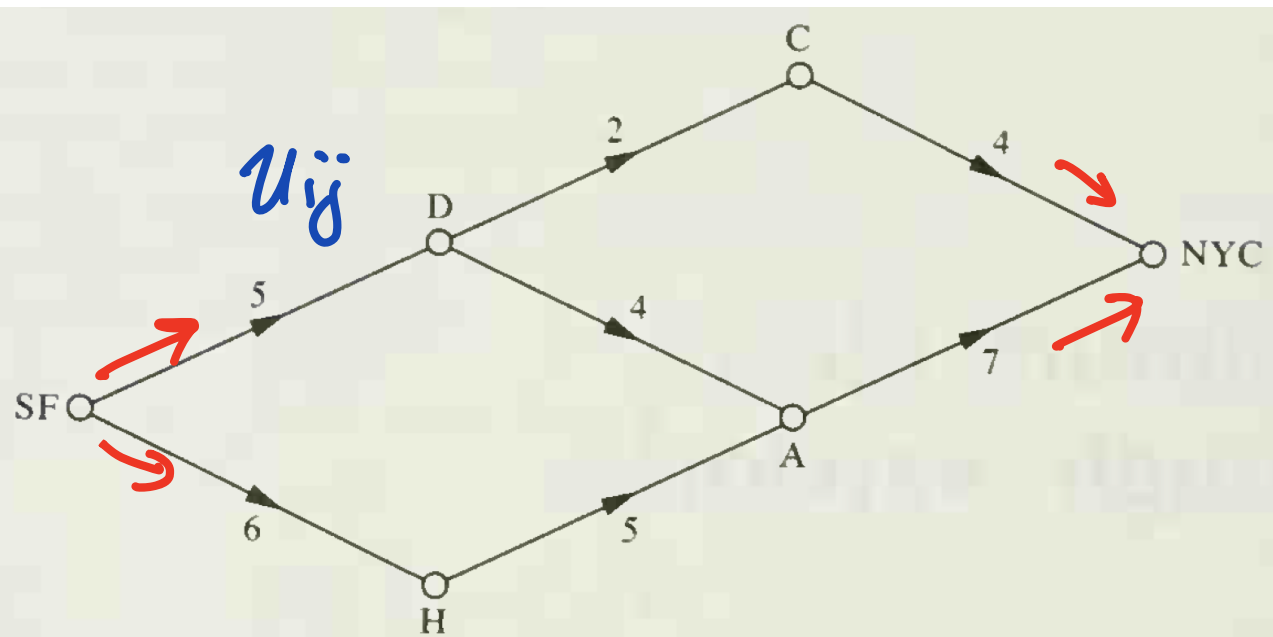


Figure 22.1

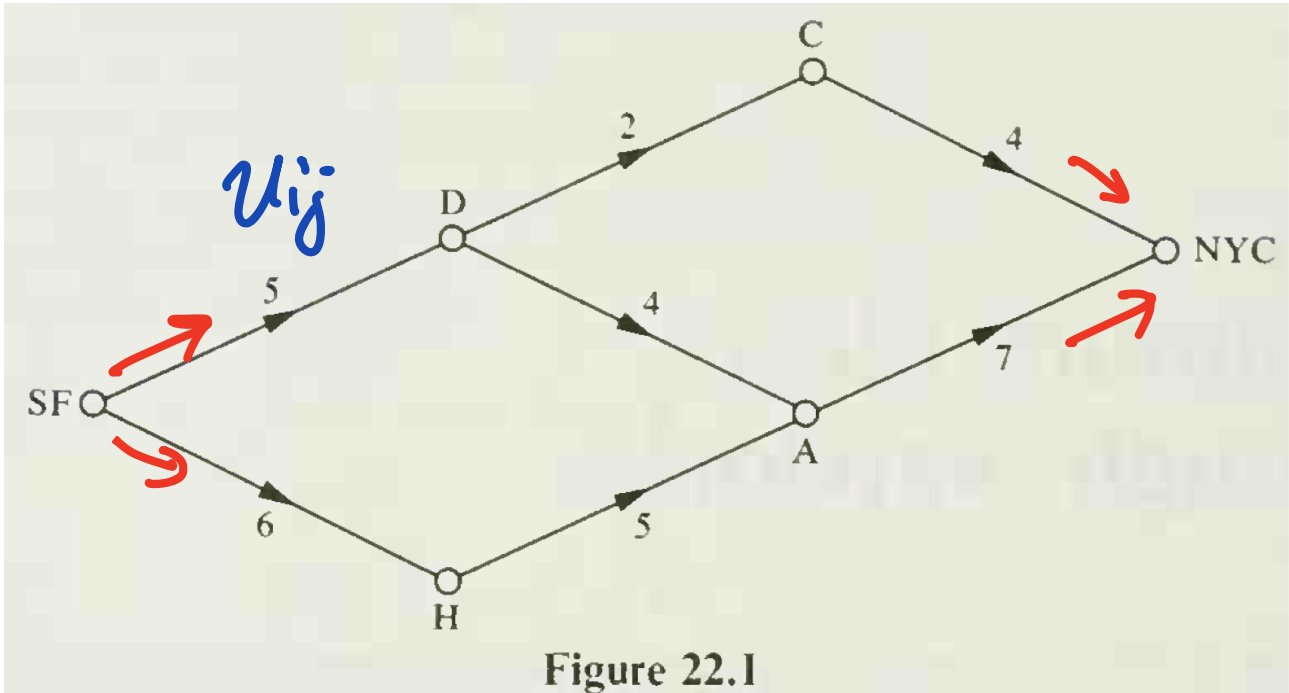


Figure 22.1

Flow = 8

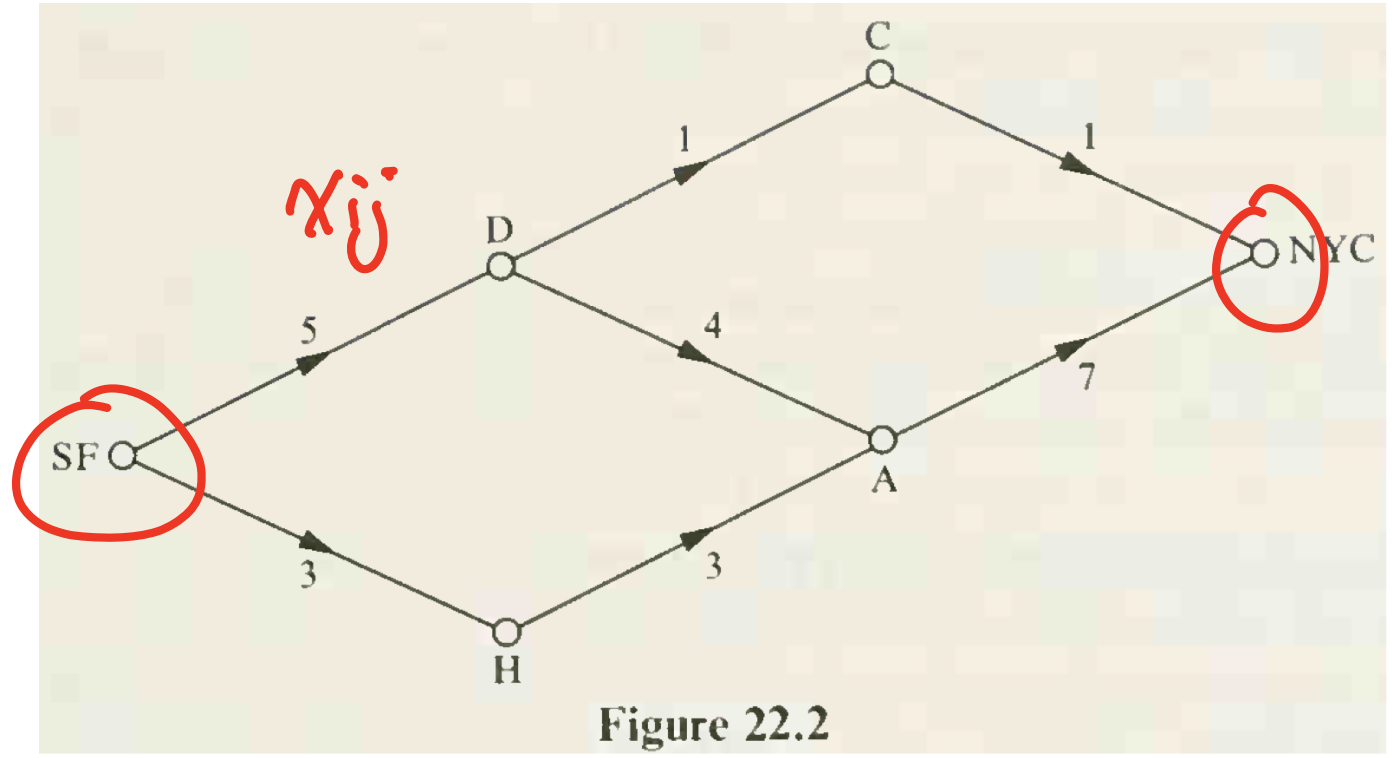


Figure 22.2

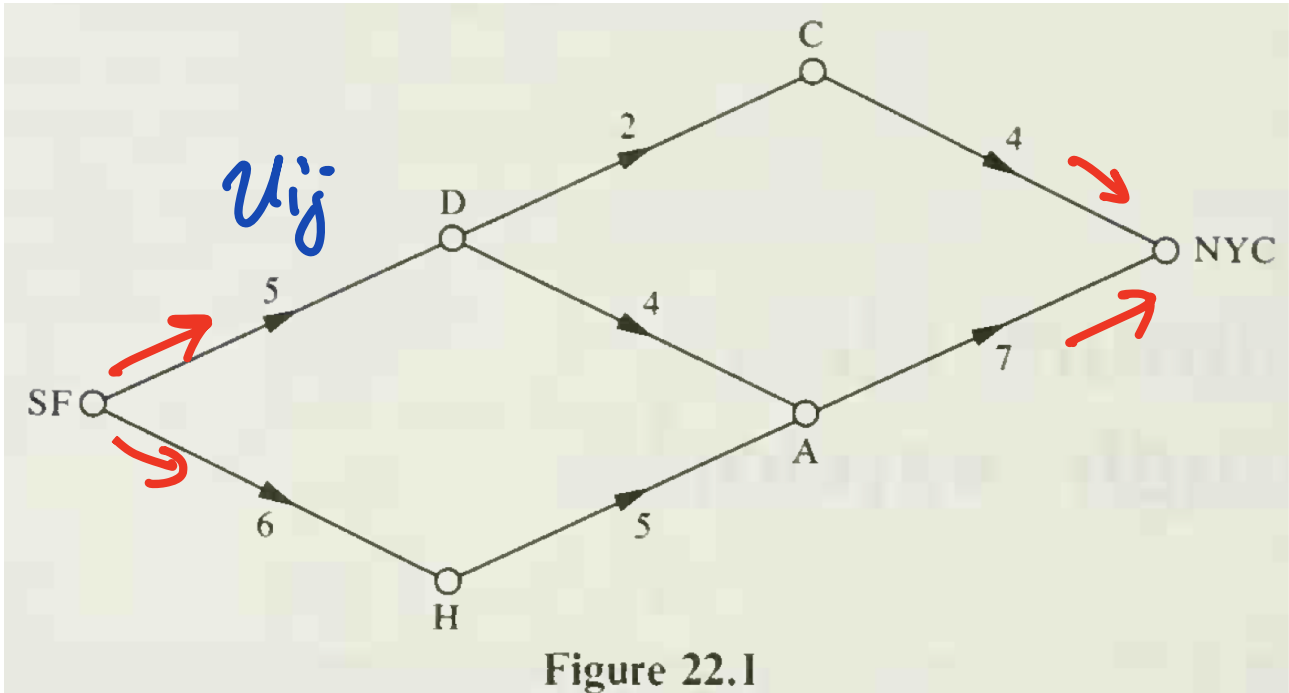


Figure 22.1

Flow = 9

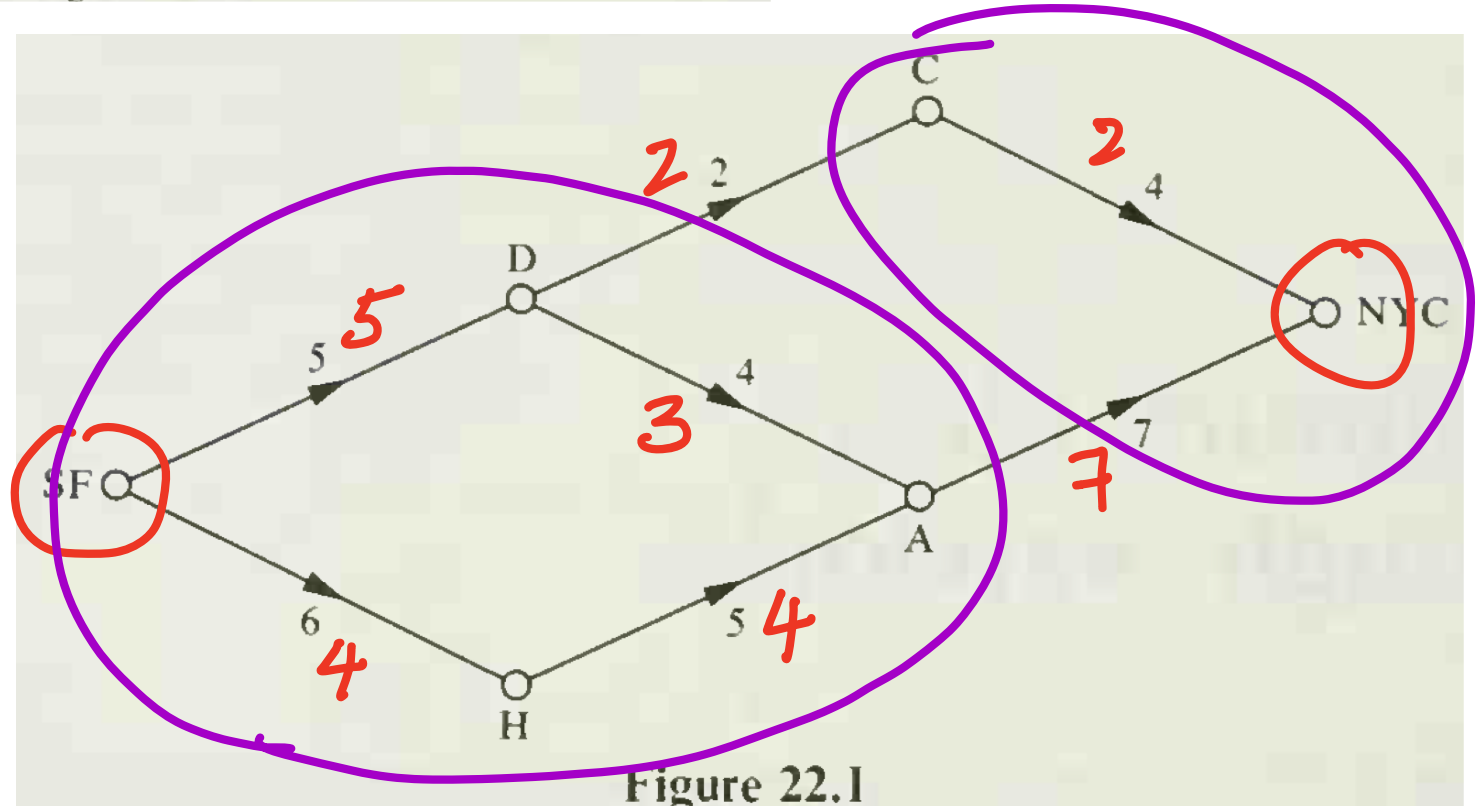


Figure 22.1

# Proof of Max Flow = Min Cut

- Let  $\{x_{ij}^*\}$  be a (optimal) max flow
- Let  $C^*$  be those nodes that some more flow can be pushed from  $s$ .
- $t \notin C^*$  (By default,  $s \in C^*$ )

Claim

$$\text{Flow}(C^* \text{ to } C^{*c}) = K(C^*)$$

$$\sum_{i \in C^*, j \notin C^*} x_{ij}^* = \sum_{i \in C^*, j \notin C^*} u_{ij}$$



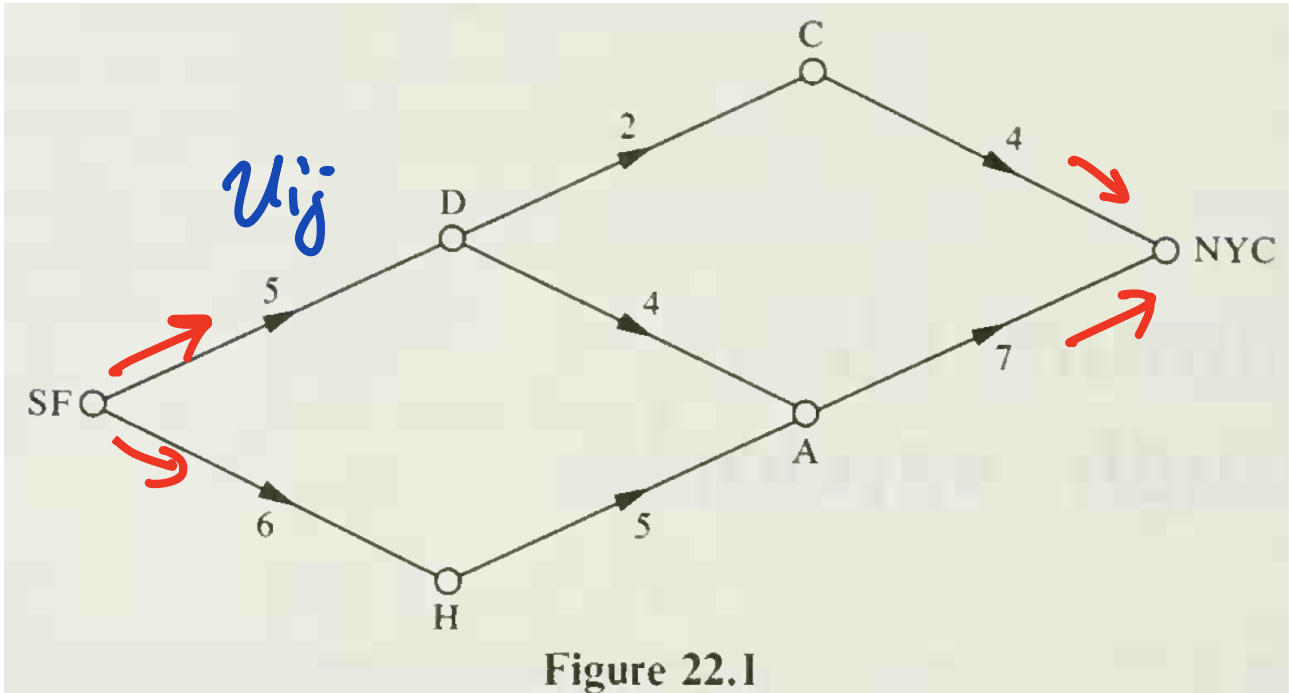


Figure 22.1

Flow = 9

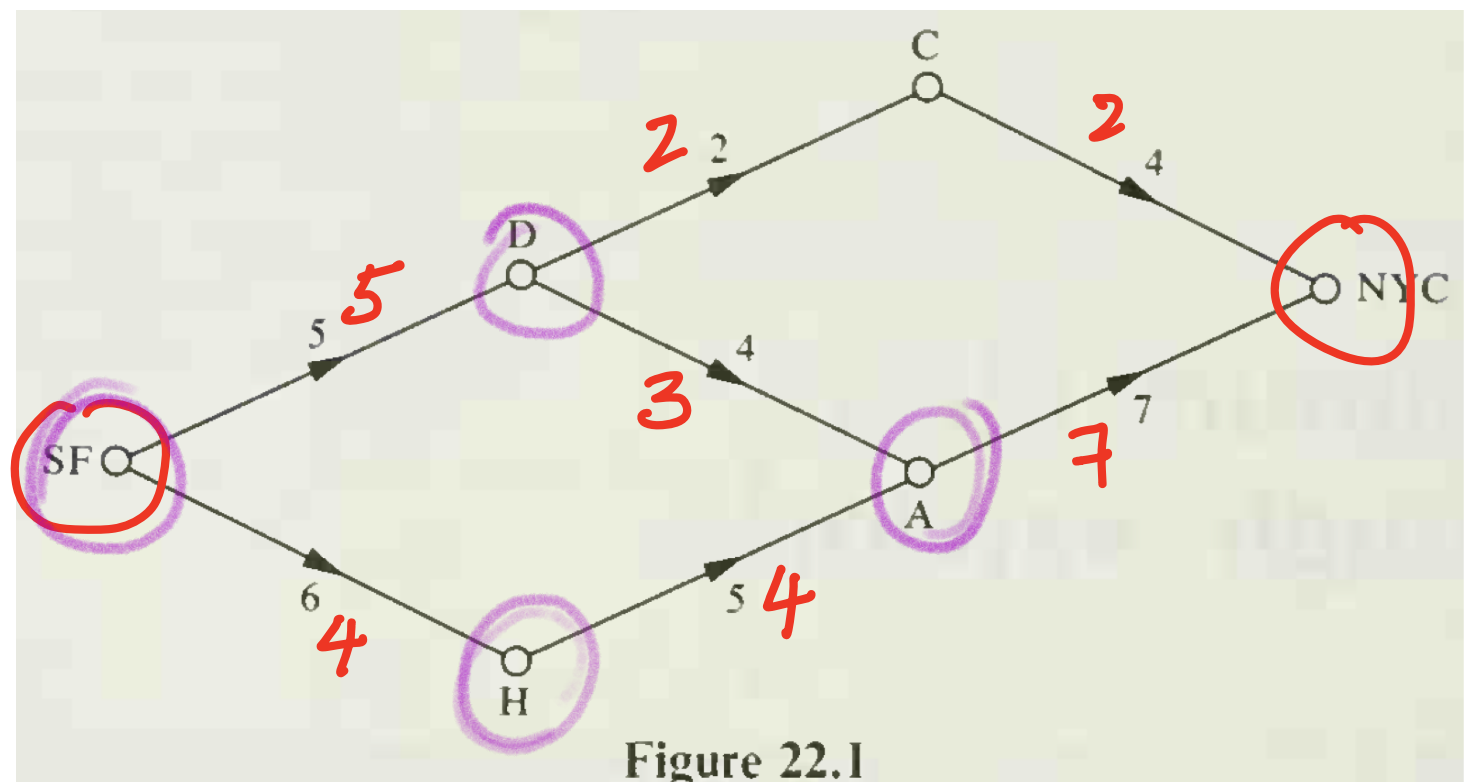


Figure 22.1

# Ford-Fulkerson Algorithm (Augmented Path)

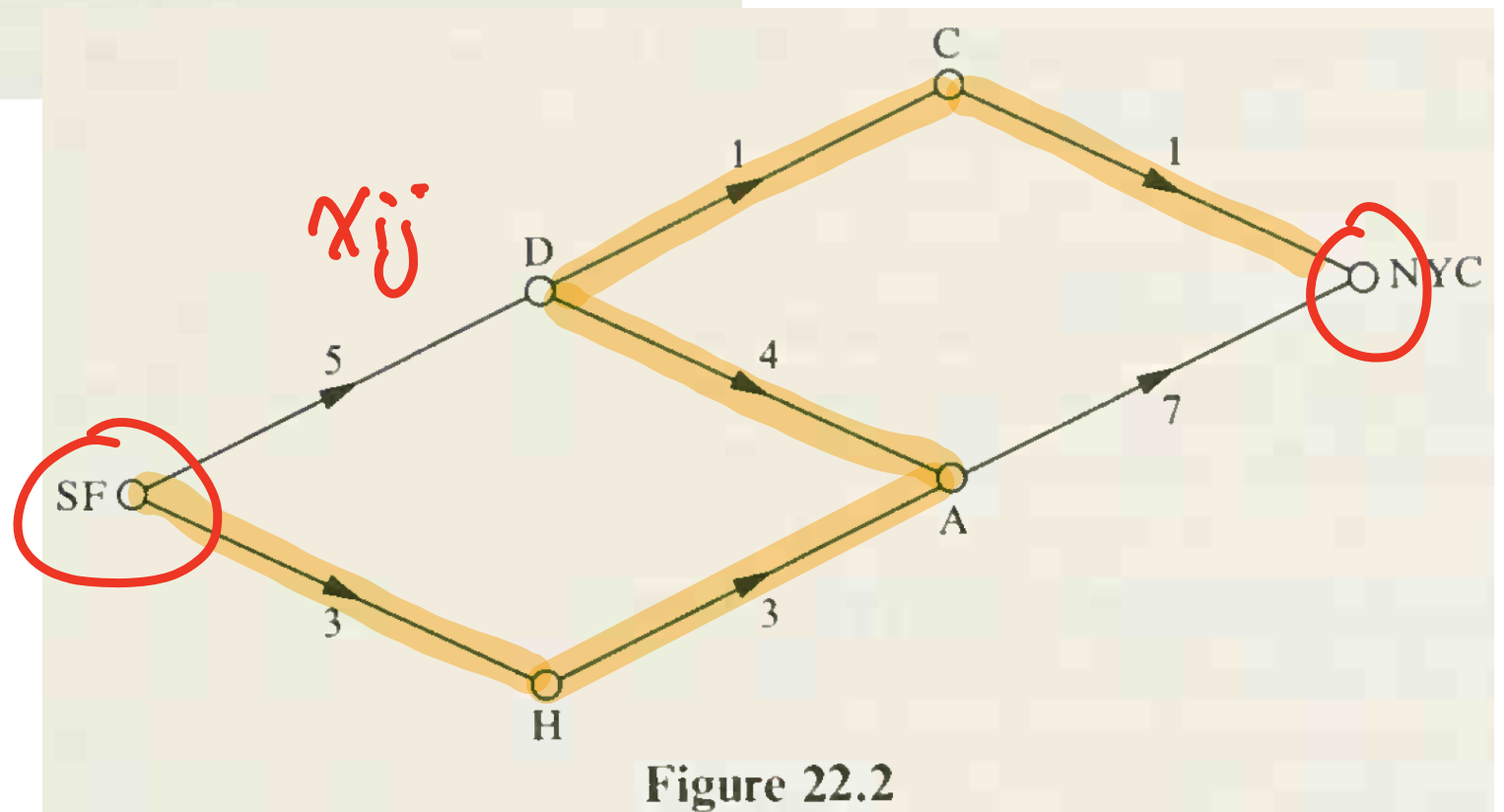
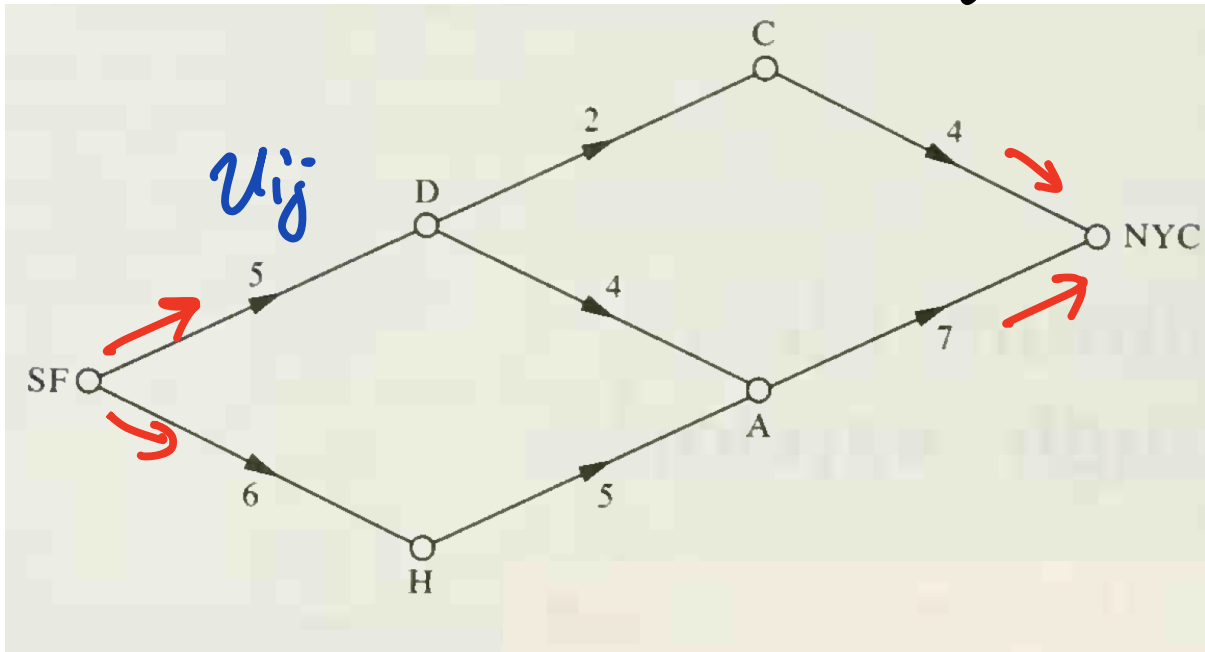


Figure 22.2

# Ford-Fulkerson Algorithm (Augmented Path)

[C] p. 374

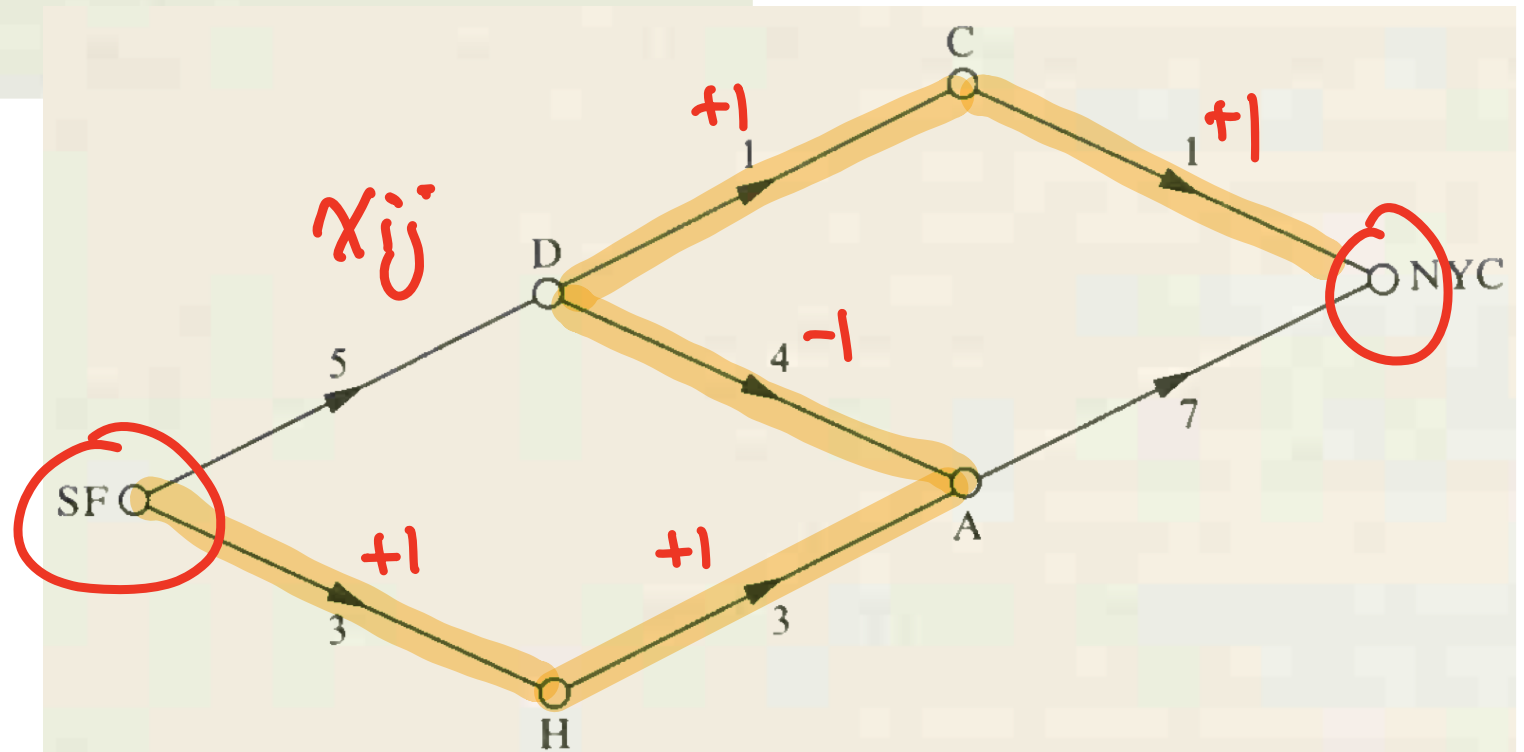
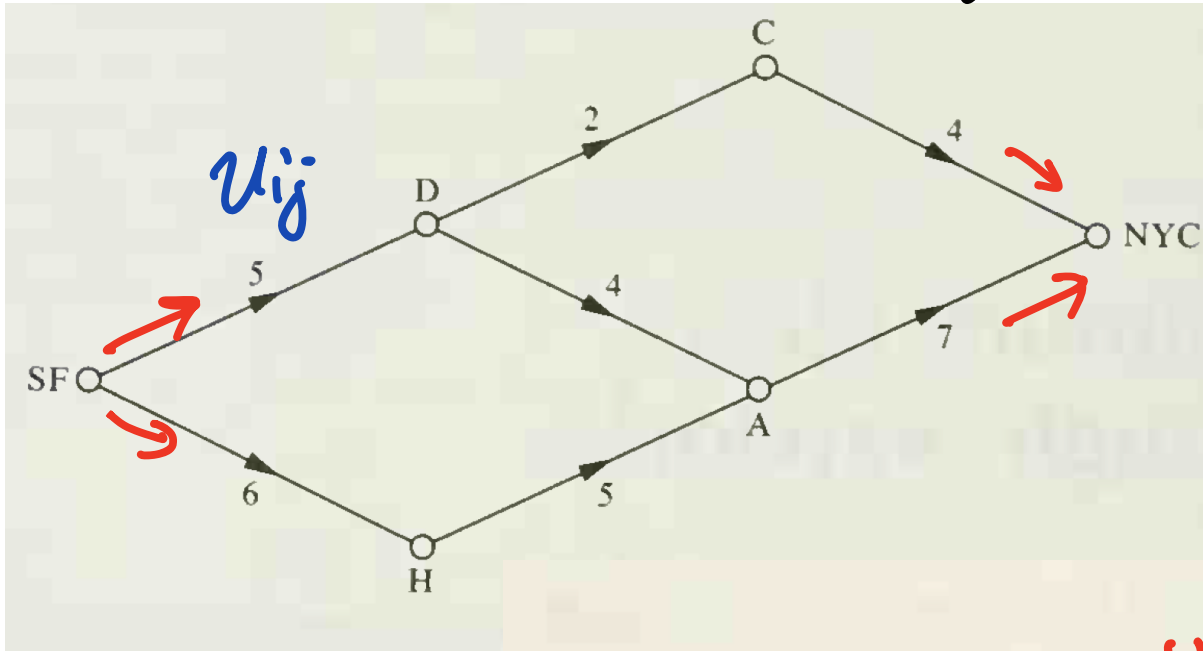


Figure 22.2

# Ford-Fulkerson Algorithm (Augmented Path)

## Implementations

Our description of the augmenting path method does not specify a way of searching for the arcs  $ij$  such that  $i \in C, j \notin C, x_{ij} < u_{ij}$ , and the arcs  $ji$  such that  $j \notin C, i \in C, x_{ji} > 0$ . Ford and Fulkerson did specify a way of doing that. In their terminology, nodes in  $C$  are called labeled and nodes outside  $C$  are called unlabeled; the labeled nodes are divided further into scanned and unscanned. Initially, the source  $s$  is labeled but unscanned and all the remaining nodes are unlabeled. Scanning a labeled node  $i$  means examining all the arcs  $ij$  and, whenever such an arc satisfies  $x_{ij} < u_{ij}, j \notin C$ , setting  $j \in C, p(j) = ij$  and examining all the arcs  $ji$  and, whenever such an arc satisfies  $x_{ji} > 0, j \notin C$ , setting  $j \in C, p(j) = ji$ . The search may be described as in Box 22.1.

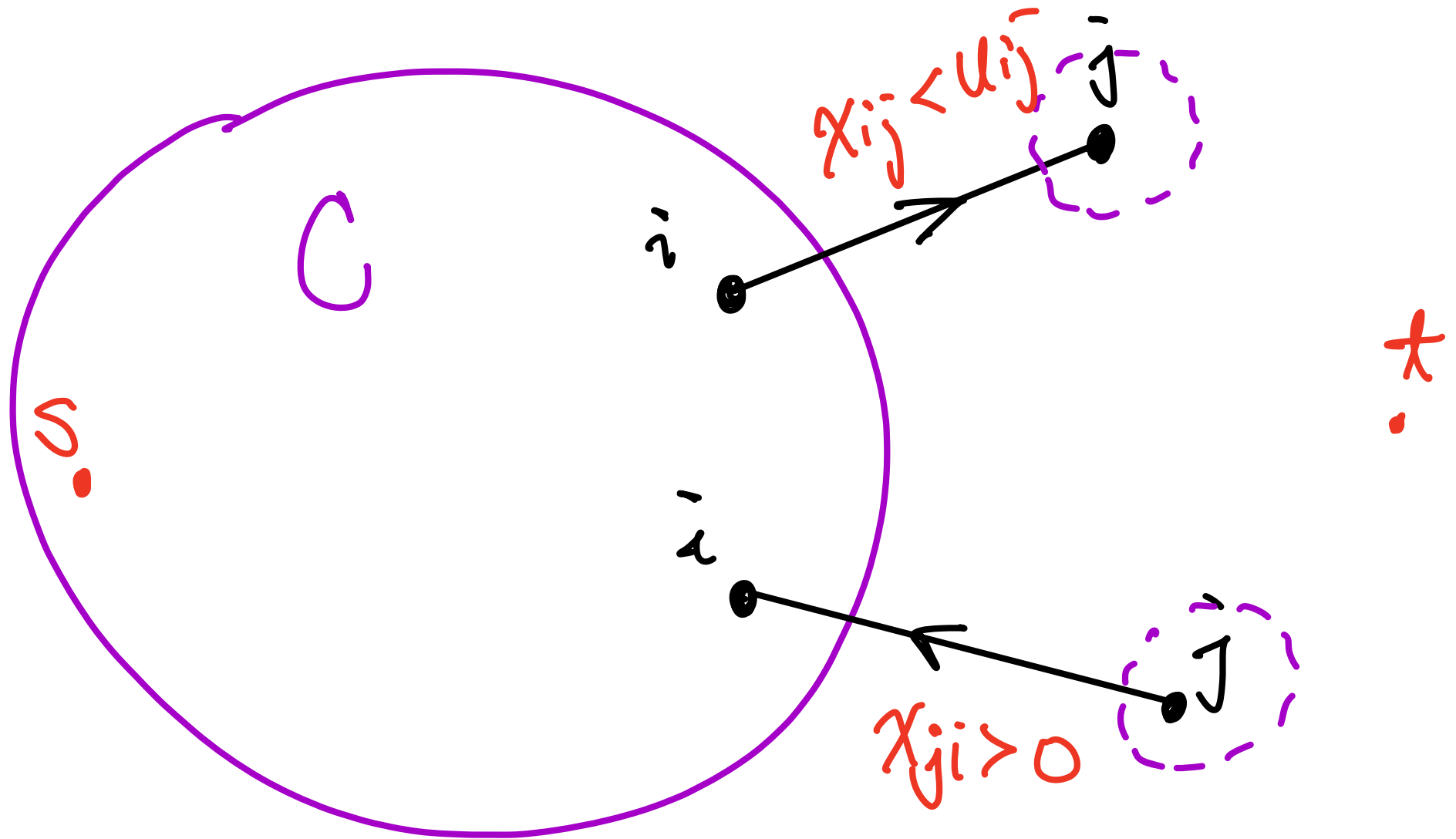
### BOX 22.1 Search for an Augmenting Path

*Step 0.* Mark  $s$  as labeled unscanned; mark the remaining nodes as unlabeled.

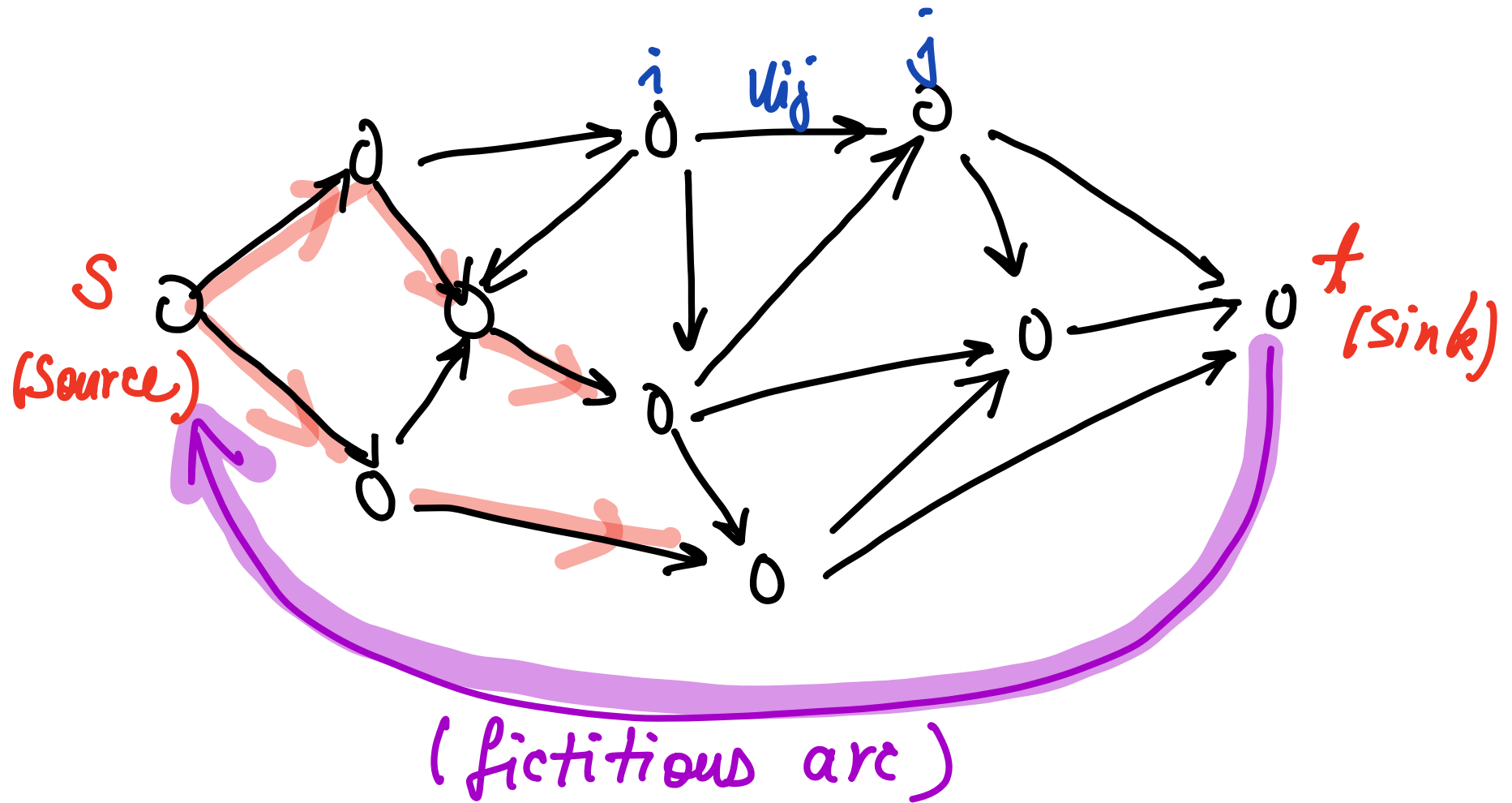
*Step 1.* If all the labeled nodes are scanned then stop [the set  $C$  of labeled nodes satisfies (22.7) and (22.8)]; otherwise, choose a labeled unscanned node  $i$ .

*Step 2* Scan  $i$ . If  $t$  has become labeled then stop (an  $x$ -augmenting path has been found); otherwise return to step 1.

# Ford-Fulkerson Algorithm (Augmented Path)



# Max Flow Min Cut (LP Formulation)



$$x_{ts}, \quad c_{ts} = -1, \quad u_{ts} = +\infty$$

# Max Flow Min Cut (LP Formulation)

$$\underline{\min} \quad -X_{ts}$$

$$\text{st.} \quad X_{ts} = \sum_j X_{sj}$$

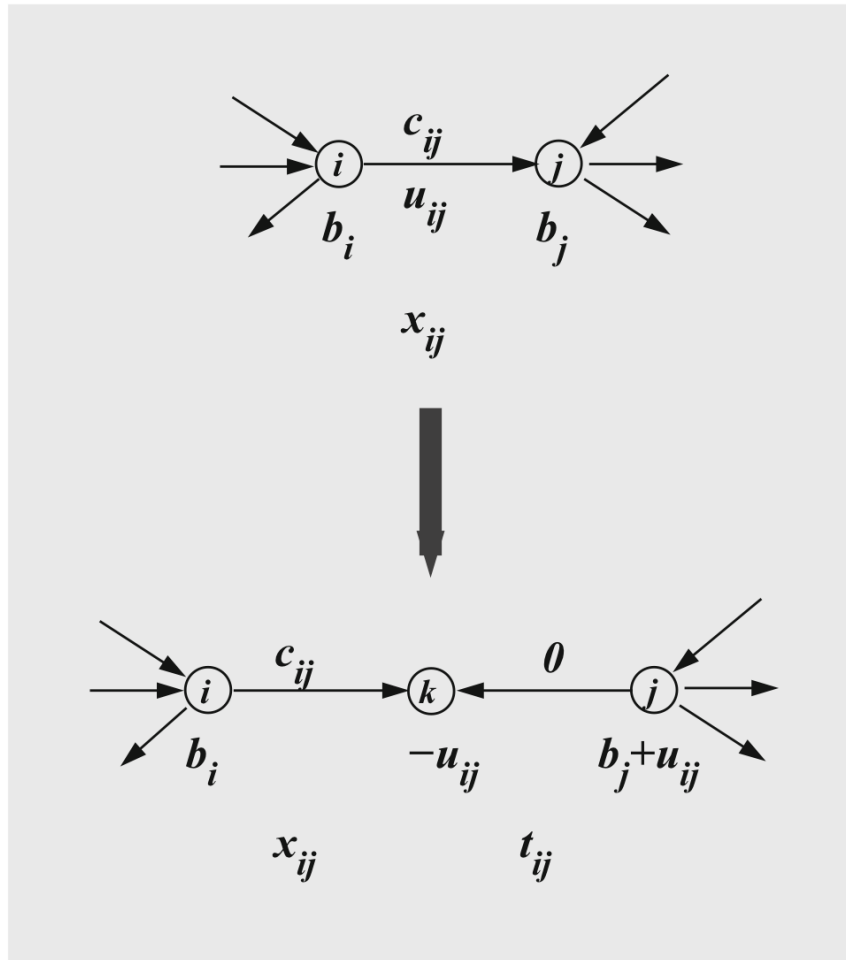
$$\sum_i X_{it} = X_{ts}$$

$$\sum_i X_{ik} = \sum_j X_{kj} \quad k \neq s, t$$

$$0 \leq X_{ij} \leq U_{ij}$$

$$(U_{ts} = +\infty)$$

# Proof of Max Flow = Min Cut (LP)



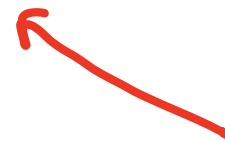
$$0 \leq x_{ij} \leq u_{ij}$$



$$x_{ij} = 0 \implies y_i + c_{ij} \geq y_j$$

$$x_{ij} = u_{ij} \implies y_i + c_{ij} \leq y_j$$

$$0 < x_{ij} < u_{ij} \implies y_i + c_{ij} = y_j$$



$$x_{ij} + t_{ij} = u_{ij}$$

FIGURE 15.4. Adding a new node,  $k$ , to accommodate an arc  $(i, j)$  having an upper bound  $u_{ij}$  on its flow capacity.



# Proof of Max Flow = Min Cut (LP)

Let  $x_{ij}^*$ ,  $(i, j) \in \mathcal{A}$ , denote the optimal values of the primal variables, and let  $y_i^*$ ,  $i \in \mathcal{N}$ , denote the optimal values of the dual variables. Then the complementarity conditions (15.6) imply that

(15.9)

$$x_{ij}^* = 0 \text{ whenever } y_i^* + c_{ij} > y_j^*$$

(15.10)

$$x_{ij}^* = u_{ij} \text{ whenever } y_i^* + c_{ij} < y_j^*.$$

In particular,

$$y_t^* - 1 \geq y_s^*$$

(since  $u_{ts} = \infty$ ). Put  $C^* = \{k : y_k^* \leq y_s^*\}$ . Clearly,  $C^*$  is a cut.

Consider an arc having its tail in  $C^*$  and its head in the complement of  $C^*$ . It follows from the definition of  $C^*$  that  $y_i^* \leq y_s^* < y_j^*$ . Since  $c_{ij}$  is zero, we see from (15.10) that  $x_{ij}^* = u_{ij}$ .

Now consider an original arc having its tail in the complement of  $C^*$  and its head in  $C^*$  (i.e., bridging the two sets in the opposite direction). It follows then that  $y_j^* \leq y_s^* < y_i^*$ . Hence, we see from (15.9) that  $x_{ij}^* = 0$ .