

**15.082J, 6.855J, and ESD.78J**  
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**Eulerian Walks**  
**Flow Decomposition and**  
**Transformations**

# Eulerian Walks in Directed Graphs in $O(m)$ time.

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**Step 1.** Create a breadth first search tree into node

1. For  $j$  not equal to 1, put the arc out of  $j$  in  $T$  last on the arc list  $A(j)$ .

**Step 2.** Create an Eulerian cycle by starting a walk at node 1 and selecting arcs in the order they appear on the arc lists.

# Proof of Correctness

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**Relies on the following observation and invariant:**

**Observation:** The walk will terminate at node 1.

Whenever the walk visits node  $j$  for  $j \neq 1$ , the walk has traversed one more arc entering node  $j$  than leaving node  $j$ .

**Invariant:** If the walk has not traversed the tree arc for node  $j$ , then there is a path from node  $j$  to node 1 consisting of nontraversed tree arcs.

Eulerian Cycle  
Animation

# Eulerian Cycles in undirected graphs

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**Strategy:** reduce to the directed graph problem as follows:

**Step 1.** Use dfs to partition the arcs into disjoint cycles

**Step 2.** Orient each arc along its directed cycle.

Afterwards, for all  $i$ , the number of arcs entering node  $i$  is the same as the number of arcs leaving node  $i$ .

**Step 3.** Run the algorithm for finding Eulerian Cycles in directed graphs

# Flow Decomposition and Transformations

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- ◆ Flow Decomposition
  - ◆ Removing Lower Bounds
  - ◆ Removing Upper Bounds
  - ◆ Node splitting
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- ◆ Arc flows: an arc flow  $x$  is a vector  $x$  satisfying:

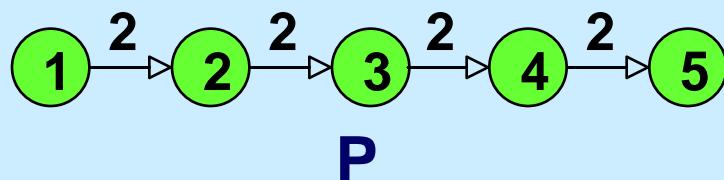
$$\text{Let } b(i) = \sum_j x_{ij} - \sum_i x_{ji}$$

We are not focused on upper and lower bounds  
on  $x$  for now.

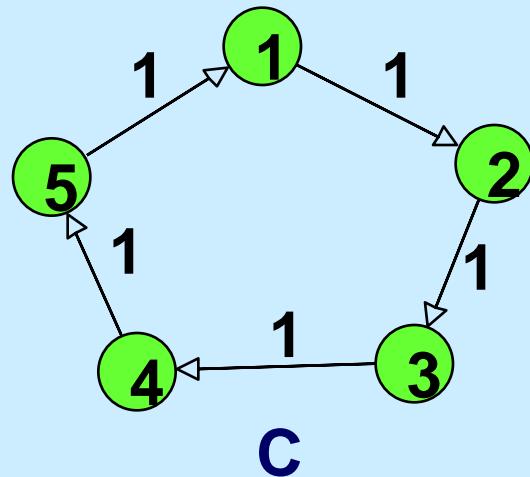
# Flows along Paths

**Usual:** represent flows in terms of flows in arcs.

**Alternative:** represent a flow as the sum of flows in paths and cycles.



Two units of flow  
in the path P



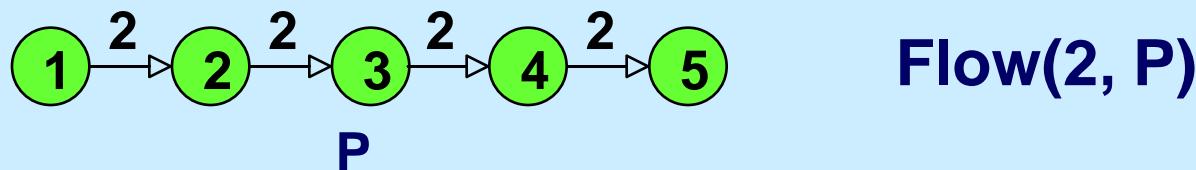
One unit of flow  
around the cycle C

# Properties of Path Flows

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Let  $P$  be a directed path.

Let  $\text{Flow}(\text{TM}, P)$  be a flow of  $\text{TM}$  units in each arc of the path  $P$ .

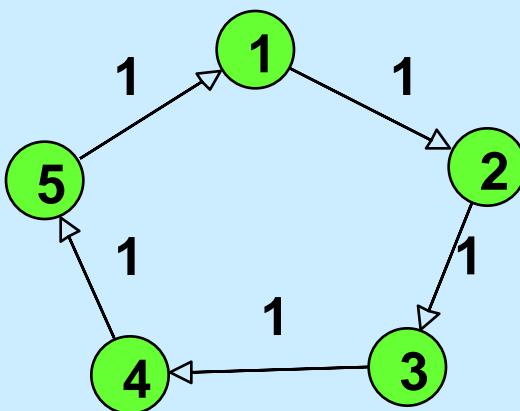


**Observation.** If  $P$  is a path from  $s$  to  $t$ , then  $\text{Flow}(\text{TM}, P)$  sends units of  $\delta$  flow from  $s$  to  $t$ , and has conservation of flow at other nodes.

# Property of Cycle Flows

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- ◆ If  $p$  is a cycle, then sending one unit of flow along  $p$  satisfies conservation of flow everywhere.



# Representations as Flows along Paths and Cycles

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Let  $\mathcal{P}$  be a collection of Paths; let  $f(P)$  denote the flow in path  $P$

Let  $\mathcal{C}$  be a collection of cycles; let  $f(C)$  denote the flow in cycle  $C$ .

One can convert the path and cycle flows into an arc flow  $x$  as follows: for each arc  $(i,j) \in A$

$$x_{ij} = \sum_{P \ni (i,j)} f(P) + \sum_{C \ni (i,j)} f(C)$$

# Flow Decomposition

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**x:** Initial flow

**y:** updated flow

**G(y):** subgraph with arcs  $(i, j)$  with  $y_{ij} > 0$  and incident nodes

**f(P)** Flow around path P (during the algorithm)

**P:** paths with flow in the decomposition

**C:** cycles with flow in the decomposition

**INVARIANT**

$$x_{ij} = y_{ij} + \sum_{P \ni (i,j)} f(P) + \sum_{C \ni (i,j)} f(C)$$

Initially,  $x = y$  and  $f = 0$ .

At end,  $y = 0$ , and  $f$  gives the flow decomposition.

# Deficit and Excess Nodes

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Let  $x$  be a flow (not necessarily feasible)

If the flow out of node  $i$  exceeds the flow into node  $i$ , then node  $i$  is a **deficit** node.

Its deficit is  $\sum_j x_{ij} - \sum_k x_{ki}$ .

If the flow out of node  $i$  is less than the flow into node  $i$ , then node  $i$  is an **excess** node.

Its excess is  $-\sum_j x_{ij} + \sum_k x_{ki}$ .

If the flow out of node  $i$  equals the flow into node  $i$ , then node  $i$  is a **balanced** node.

# Flow Decomposition Algorithm

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**Step 0.** Initialize:  $y := x$ ;  $f := 0$ ;  $\mathcal{P} := \emptyset$ ;  $C := \emptyset$ ;

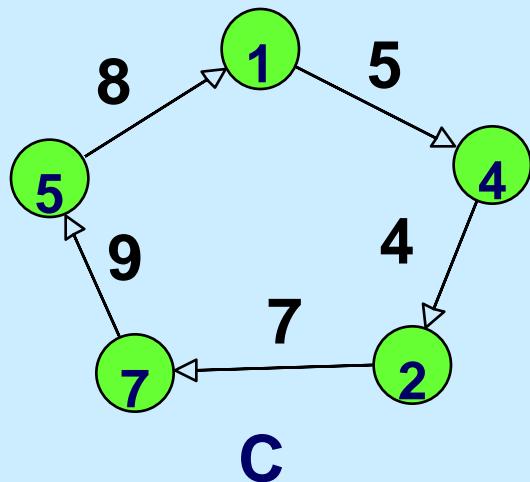
**Step 1.** Select a deficit node  $j$  in  $G(y)$ . If no deficit node exists, select a node  $j$  with an incident arc in  $G(y)$ ;

**Step 2.** Carry out depth first search from  $j$  in  $G(y)$  until finding a directed cycle  $W$  in  $G(y)$  or a path  $W$  in  $G(y)$  from  $s$  to a node  $t$  with excess in  $G(y)$ .

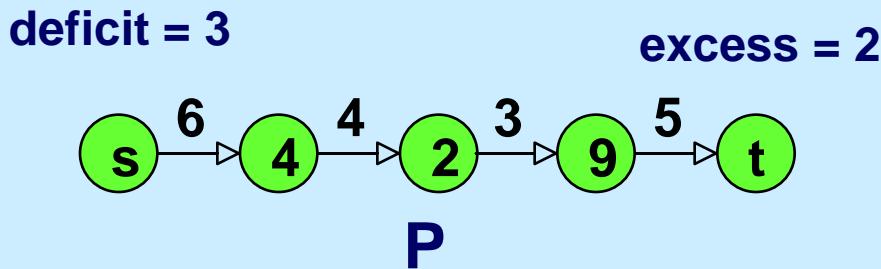
**Step 3.**

1. Let  $\Delta = \text{capacity of } W \text{ in } G(y)$ . (See next slide)
2. Add  $W$  to the decomposition with  $f(W) = \Delta$ .
3. Update  $y$  (subtract flow in  $W$ ) and excesses and deficits
4. If  $y \neq 0$ , then go to Step 1

# Capacities of Paths and Cycles



The capacity of  $C$  is  
= min arc flow on  $C$   
wrt flow  $y$ .  
**capacity = 4**



The capacity of  $P$  is  
denoted as  $\Delta(P, y) =$   
 $\min[ \text{def}(s), \text{excess}(t),$   
 $\min (x_{ij} : (i,j) \in P) ]$

$\chi\alpha\pi\alpha\chi\tau\psi = 2$

**Flow Decomposition**  
**Animation**

# Complexity Analysis

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- ◆ **Select initial node:**
  - $O(1)$  per path or cycle, assuming that we maintain a set of supply nodes and a set of balanced nodes incident to a positive flow arc
- ◆ **Find cycle or path**
  - $O(n)$  per path or cycle since finding the next arc in depth first search takes  $O(1)$  steps.
- ◆ **Update step**
  - $O(n)$  per path or cycle

## Complexity Analysis (continued)

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**Lemma.** The number of paths and cycles found in the flow decomposition is at most  $m + n - 1$ .

**Proof.** In the update step for a cycle, at least one of the arcs has its capacity reduced to 0, and the arc is eliminated.

In an update step for a path, either an arc is eliminated, or a deficit node has its deficit reduced to 0, or an excess node has its excess reduced to 0.

(Also, there is never a situation with exactly one node whose excess or deficit is non-zero).

# Conclusion

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***Flow Decomposition Theorem.*** Any non-negative feasible flow  $x$  can be decomposed into the following:

- i. the sum of flows in paths directed from deficit nodes to excess nodes, plus
- ii. the sum of flows around directed cycles.

It will always have at most  $n + m$  paths and cycles.

**Remark.** The decomposition usually is not unique.

## Corollary

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A **circulation** is a flow with the property that the flow in is the flow out for each node.

***Flow Decomposition Theorem for circulations.*** Any non-negative feasible flow  $x$  can be decomposed into the sum of flows around directed cycles.

It will always have at most  $m$  cycles.

# An application of Flow Decomposition

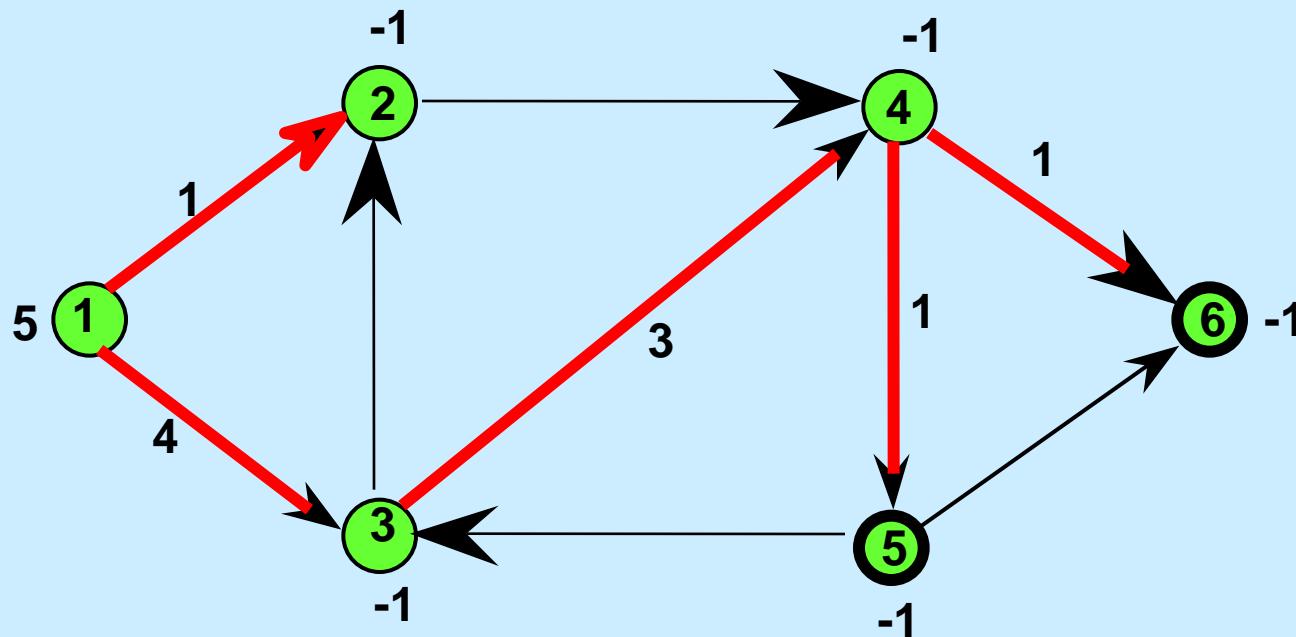
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Consider a feasible flow where the supply of node 1 is  $n-1$ , and the supply of every other node is  $-1$ .

$$\sum_j x_{ij} - \sum_j x_{ji} = \begin{cases} n-1 & \text{if } i=1 \\ -1 & \text{if } i \neq 1 \end{cases}$$

Suppose the arcs with positive flow have no cycle.  
Then the flow can be decomposed into unit flows  
along paths from node 1 to node  $j$  for each  $j \neq 1$ .

# A flow and its decomposition



The decomposition of flows yields the paths:

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

1-3-4-5 and 1-3-4-6.

There are no cycles in the decomposition.

## Application to shortest paths

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To find a shortest path from node 1 to each other node in a network, find a minimum cost flow in which  $b(1) = n-1$  and  $b(j) = -1$  for  $j \neq 1$ .

The flow decomposition gives the shortest paths.

# Other Applications of Flow Decomposition

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- ◆ **Reformulations of Problems.**
  - There are network flow models that use path and cycle based formulations.
  - Multicommodity Flows
- ◆ **Used in proving theorems**
- ◆ **Can be used in developing algorithms**

# The min cost flow problem (again)

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**The minimum cost flow problem**

$u_{ij}$  = capacity of arc  $(i,j)$ .

$c_{ij}$  = unit cost of flow sent on  $(i,j)$ .

$x_{ij}$  = amount shipped on arc  $(i,j)$

**Minimize**       $\sum c_{ij}x_{ij}$   
 $\sum_j x_{ij} - \sum_k x_{ki} = b_i \quad \text{for all } i \in N.$   
and  $0 \leq x_{ij} \leq u_{ij} \quad \text{for all } (i,j) \in A.$

# The model seems very limiting

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- The lower bounds are 0.
- The supply/demand constraints must be satisfied exactly
- There are no constraints on the flow entering or leaving a node.

We can model each of these constraints using transformations.

- In addition, we can transform a min cost flow problem into an equivalent problem with no upper bounds.

# Eliminating Lower Bound on Arc Flows

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Suppose that there is a lower bound  $l_{ij}$  on the arc flow in  $(i,j)$

**Minimize**  $\sum c_{ij}x_{ij}$

$$\sum_j x_{ij} - \sum_k x_{ki} = b_i \quad \text{for all } i \in N.$$

$$\text{and } l_{ij} \leq x_{ij} \leq u_{ij} \quad \text{for all } (i,j) \in A.$$

Then let  $y_{ij} = x_{ij} - l_{ij}$ . Then  $x_{ij} = y_{ij} + l_{ij}$

**Minimize**  $\sum c_{ij}(y_{ij} + l_{ij})$

$$\sum_j (y_{ij} + l_{ij}) - \sum_k (y_{ij} + l_{ij}) = b_i \quad \text{for all } i \in N.$$

$$\text{and } l_{ij} \leq (y_{ij} + l_{ij}) \leq u_{ij} \quad \text{for all } (i,j) \in A.$$

Then simplify the expressions.

# Allowing inequality constraints

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Minimize  $\sum c_{ij}x_{ij}$

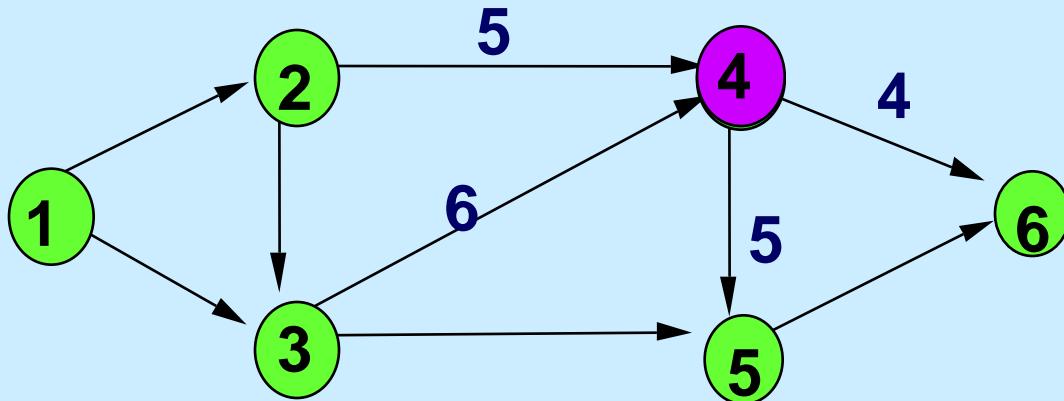
$\sum_j x_{ij} - \sum_k x_{ki} \leq b_i \quad \text{for all } i \in N.$

and  $l_{ij} \leq x_{ij} \leq u_{ij} \quad \text{for all } (i,j) \in A.$

Let  $B = \sum_i b_i$ . For feasibility, we need  $B \geq 0$

Create a “dummy node”  $n+1$ , with  $b_{n+1} = -B$ . Add arcs  $(i, n+1)$  for  $i = 1$  to  $n$ , with  $c_{i,n+1} = 0$ . Any feasible solution for the original problem can be transformed into a feasible solution for the new problem by sending excess flow to node  $n+1$ .

# Node Splitting

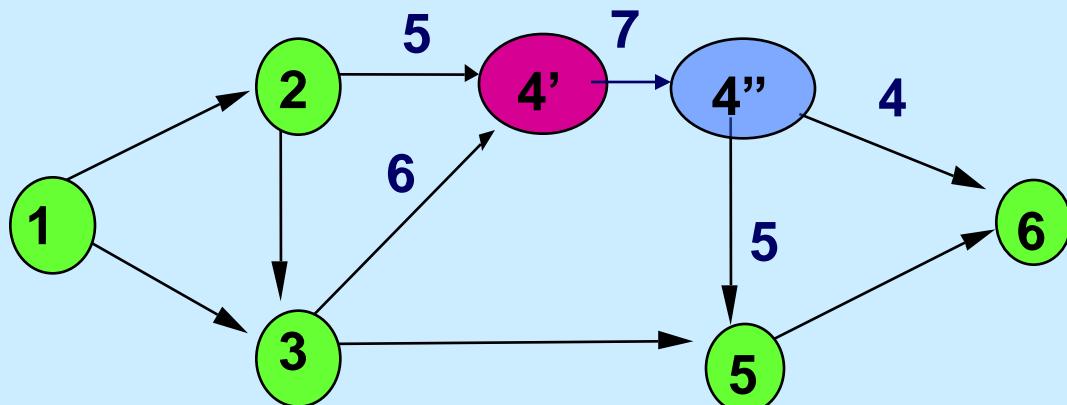


Flow  $x$

Arc numbers  
are capacities

Suppose that we want to add the constraint that the flow into node 4 is at most 7.

Method: split node 4 into two nodes, say  $4'$  and  $4''$



Flow  $x'$  can be obtained from flow  $x$ , and vice versa.

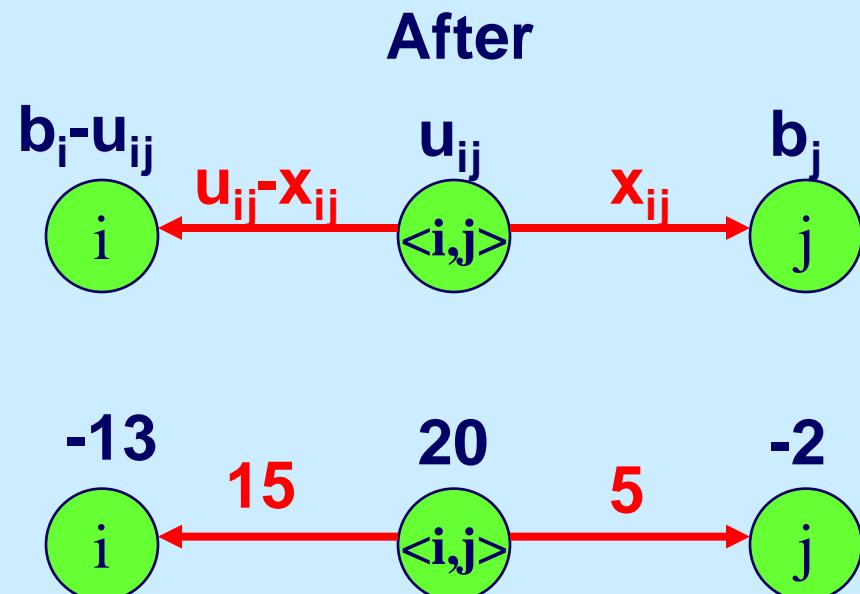
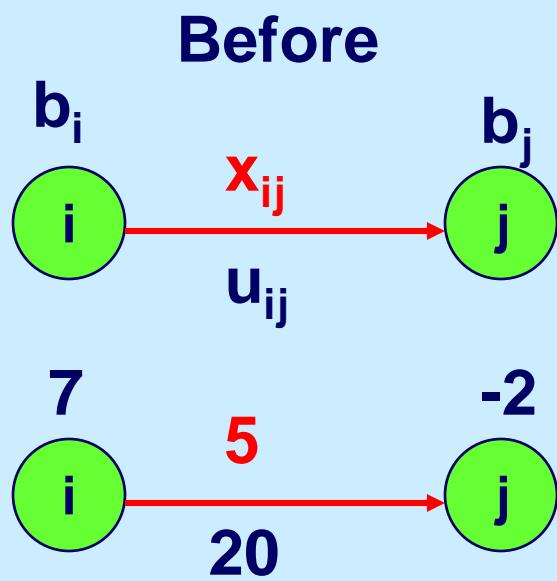
# Eliminating Upper Bounds on Arc Flows

The minimum cost flow problem

$$\text{Min } \sum c_{ij}x_{ij}$$

$$\text{s.t. } \sum_j x_i - \sum_k x_{ki} = b_i \text{ for all } i \in N.$$

$$\text{and } 0 \leq x_{ij} \leq u_{ij} \text{ for all } (i,j) \in A.$$



# Summary

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1. Efficient implementation of finding an eulerian cycle.
2. Flow decomposition theorem
3. Transformations that can be used to incorporate constraints into minimum cost flow problems.

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