

Integer Programming

Nonlinear Objective Function [V] p.394

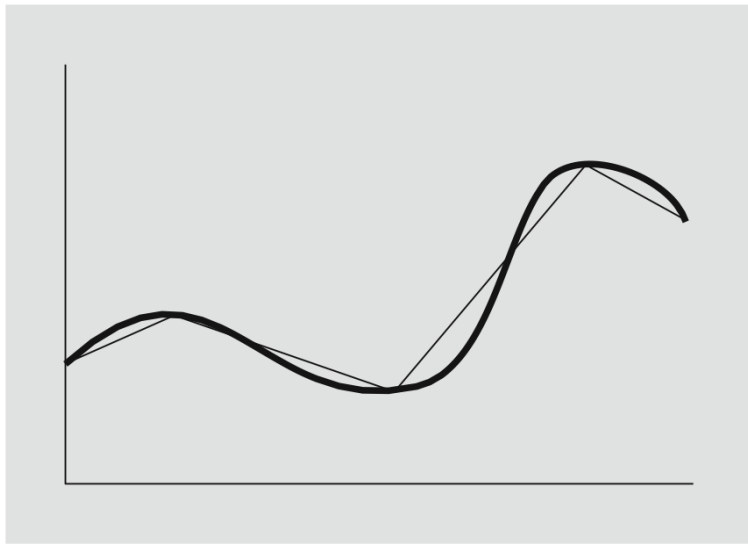


FIGURE 23.3. A nonlinear function and a piecewise linear approximation to it.

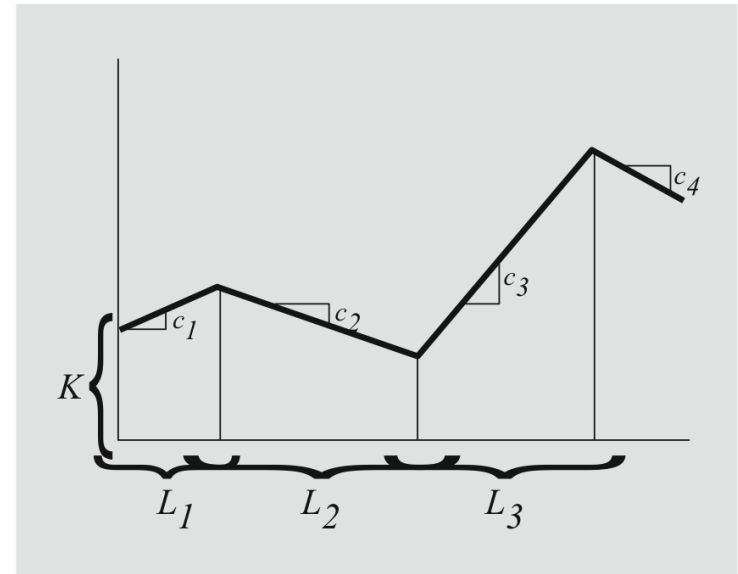
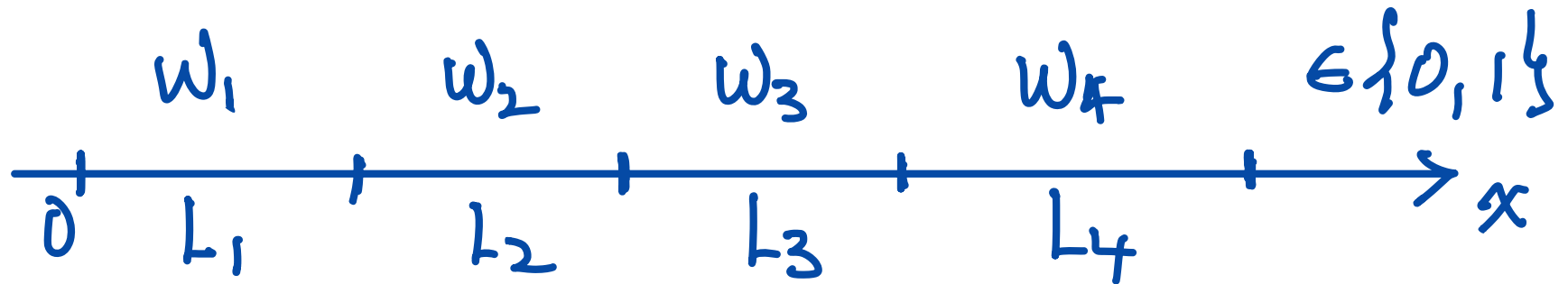


FIGURE 23.4. A piecewise linear function.

Nonlinear Objective Function [V] p.394



$$L_j w_j \leq x_j \leq L_j w_{j-1} \quad j = 1, 2, \dots, k$$

$$w_0 = 1$$

$$w_j \in \{0, 1\} \quad j = 1, 2, \dots, k$$

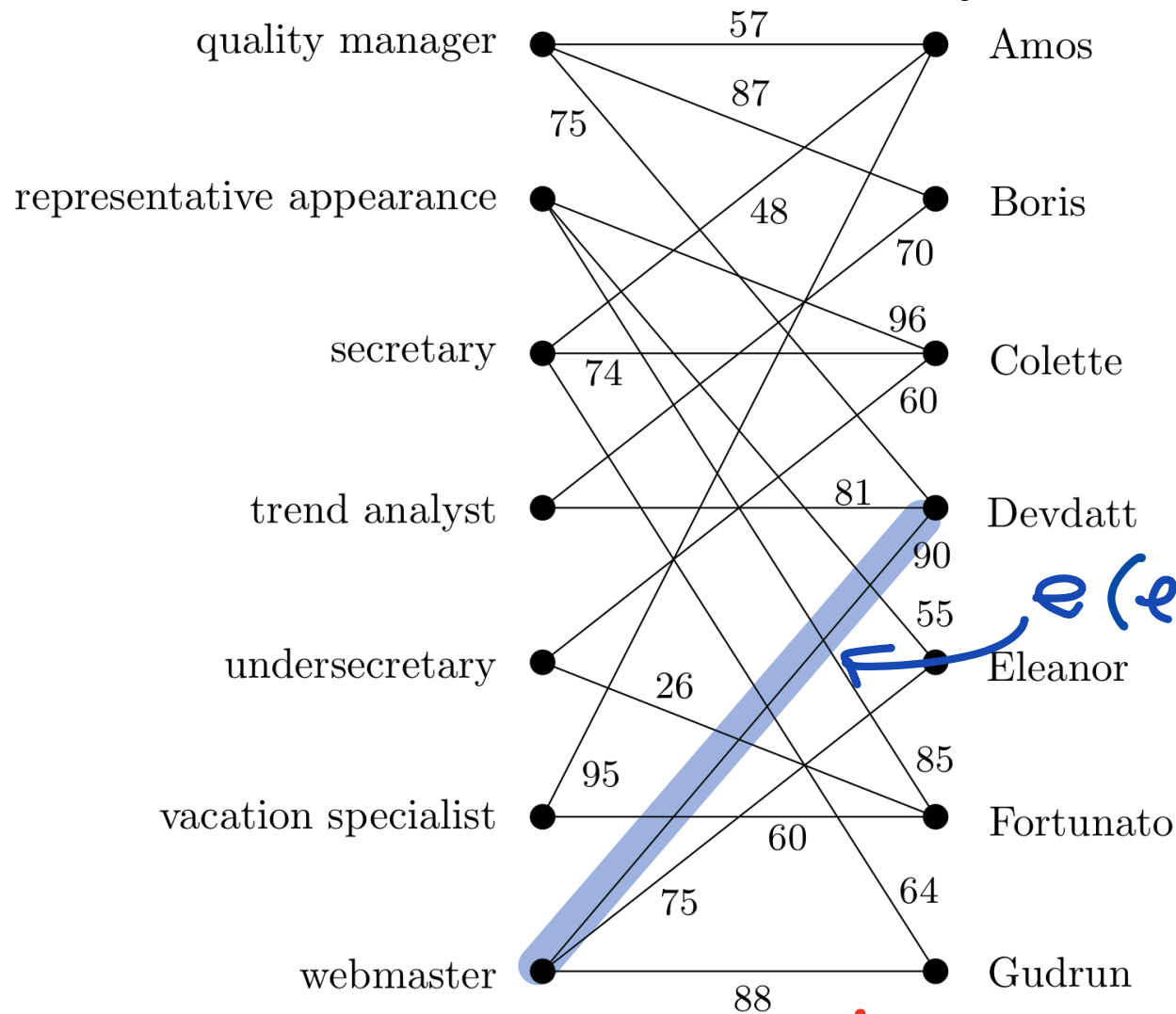
$$x_j \geq 0 \quad j = 1, 2, \dots, k.$$

Indeed, it follows from these constraints that $w_j \leq w_{j-1}$ for $j = 1, 2, \dots, k$. This inequality implies that once one of the w_j 's is zero, then all the subsequent ones must be zero. If $w_j = w_{j-1} = 1$, the two-sided inequality on x_j reduces to $L_j \leq x_j \leq L_j$. That is, $x_j = L_j$. Similarly, if $w_j = w_{j-1} = 0$, then the two-sided inequality reduces to $x_j = 0$. The only other case is when $w_j = 0$ but $w_{j-1} = 1$. In this case, the two-sided inequality becomes $0 \leq x_j \leq L_j$. Therefore, in all cases, we get what we want. Now with this decomposition we can write the piecewise linear function as

$$K + c_1 x_1 + c_2 x_2 + \dots + c_k x_k.$$

Integer Programming

[MG] p. 3, Maximum Weight Matching

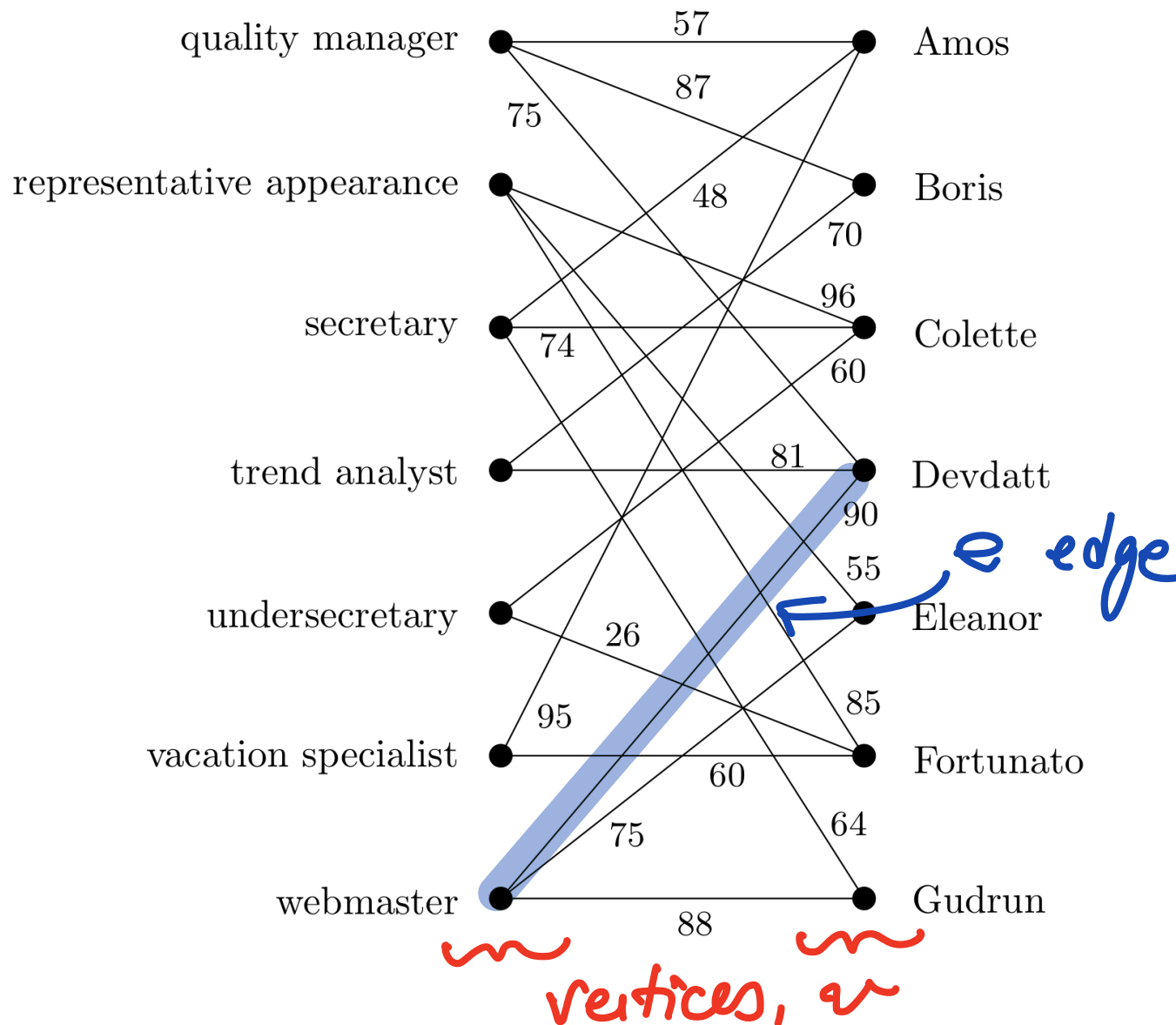


e (edge)

vertices, v

Integer Programming

[MG] p. 3, Maximum Weight Matching



$$\begin{aligned} \max \quad & \sum_e w_e x_e \\ \text{s.t.} \quad & \text{for any } v, \\ & \sum_{e, v \in e} x_e = 1 \\ & x_e \in \{0, 1\} \end{aligned}$$

2. The Assignment Problem

Given a set $\mathcal{S}$ of m people, a set $\mathcal{D}$ of m tasks, and for each $i \in \mathcal{S}$, $j \in \mathcal{D}$ a cost c_{ij} associated with assigning person i to task j , the *assignment problem* is to assign each person to one and only one task in such a manner that each task gets covered by someone and the total cost of the assignments is minimized. If we let

$$x_{ij} = \begin{cases} 1 & \text{if person } i \text{ is assigned task } j, \\ 0 & \text{otherwise,} \end{cases}$$

then the objective function can be written as

$$\text{minimize } \sum_{i \in \mathcal{S}} \sum_{j \in \mathcal{D}} c_{ij} x_{ij}.$$

The constraint that each person is assigned exactly one task can be expressed simply as

$$\sum_{j \in \mathcal{D}} x_{ij} = 1, \quad \text{for all } i \in \mathcal{S}.$$

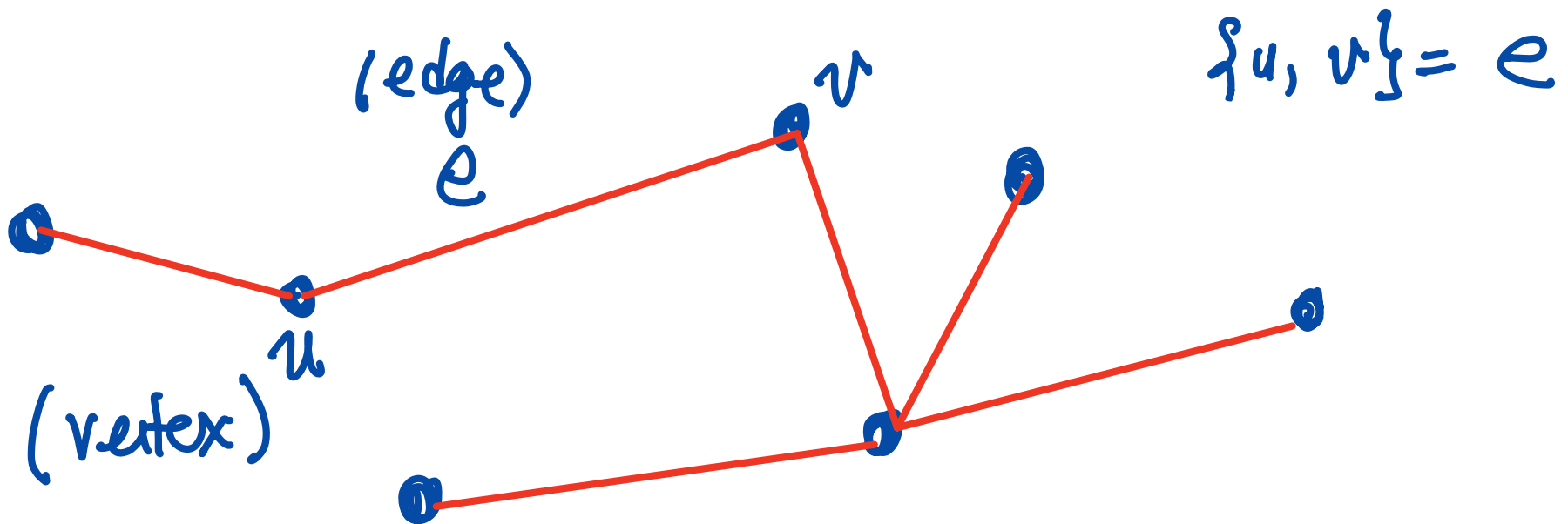
Also, the constraint that every task gets covered by someone is just

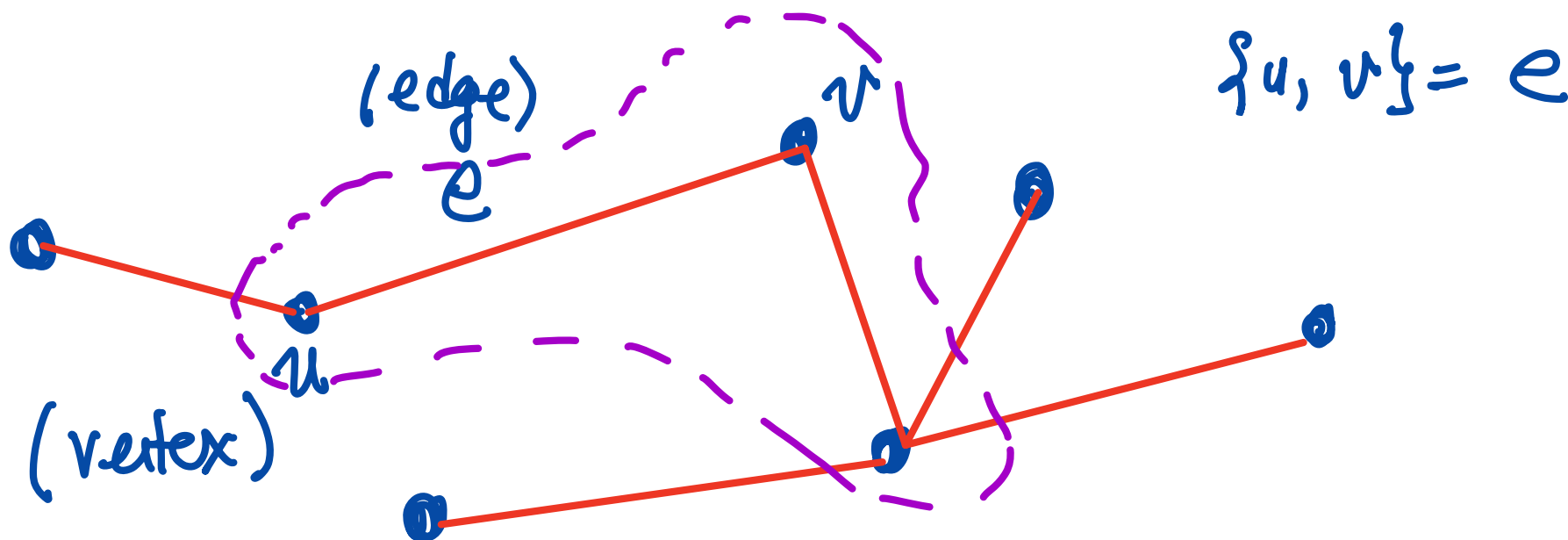
$$\sum_{i \in \mathcal{S}} x_{ij} = 1, \quad \text{for all } j \in \mathcal{D}.$$

3.3 Minimum Vertex Cover

[MG, p. 37]

The Internet had been expanding rapidly in the Free Republic of West Mor-dor, and the government issued a regulation, purely in the interest of im-proved security of the citizens, that every data link connecting two computers must be equipped with a special device for gathering statistical data about the traffic. An operator of a part of the network has to attach the govern-ment's monitoring boxes to some of his computers so that each link has a monitored computer on at least one end. Which computers should get boxes so that the total price is minimum? Let us assume that there is a flat rate per box.





This problem can be written as an integer program:

$$\begin{array}{ll}
 \text{Minimize} & \sum_{v \in V} x_v \\
 \text{subject to} & x_u + x_v \geq 1 \quad \text{for every edge } \{u, v\} \in E \\
 & x_v \in \{0, 1\} \quad \text{for all } v \in V.
 \end{array} \tag{3.2}$$

For every vertex v we have a variable x_v , which can attain values 0 or 1. The meaning of $x_v = 1$ is $v \in S$, and $x_v = 0$ means $v \notin S$. The constraint $x_u + x_v \geq 1$ guarantees that the edge $\{u, v\}$ has at least one vertex in S . The objective function is the size of S .

Knapsack Problem

23.1 Knapsack Problem. Consider a picnicker who will be carrying a knapsack that holds a maximum amount b of “stuff.” Suppose that our picnicker must decide what to take and what to leave behind. The j th thing that might be taken occupies a_j units of space in the knapsack and will bring c_j amount of “enjoyment.” The knapsack problem then is to maximize enjoyment subject to the constraint that the stuff brought must fit into the knapsack:

$$\begin{aligned} &\text{maximize} && \sum_{j=1}^n c_j x_j \\ &\text{subject to} && \sum_{j=1}^n a_j x_j \leq b \\ &&& x_j \in \{0, 1\} \quad j = 1, 2, \dots, n. \end{aligned}$$

In the usual notation, the knapsack problem is recorded as

$$\begin{aligned} &\text{maximize} && \sum_{i=1}^m c_i x_i \\ &\text{subject to} && \sum_{i=1}^m a_i x_i \leq b \\ &&& x_i = \text{nonnegative integer} \quad (i = 1, 2, \dots, m). \end{aligned} \tag{13.5}$$

[v]

p.413

[c]

p.201

Integer Programming

[MG] p. 148 Machine Scheduling

	Single B&W	Duplex B&W	Duplex Color
Master's thesis, 90 pages two-sided, 10 B&W copies	—	45 min	60 min
<i>All the Best Deals</i> flyer, 1 page one-sided, 10,000 B&W copies	2h 45 min	4h 10 min	5h 30 min
<i>Buyer's Paradise</i> flyer, 1 page one-sided, 10,000 B&W copies	2h 45 min	4h 10 min	5h 30 min
Obituary, 2 pages two-sided, 100 B&W copies	—	2 min	3 min
Party platform, 10 pages two-sided, 5,000 color copies	—	—	3h 30 min

Integer Programming

[MG] p. 148 Machine Scheduling

$M = \{1, 2, \dots, m\}$ m - Machines

$J = \{1, 2, \dots, n\}$ n - Jobs

d_{ij} = running time of Job j in Machine i

min t (total run time)

s.t. $\sum_{i \in M} x_{ij} = 1 \quad j \in J$

$\sum_{j \in J} d_{ij} x_{ij} \leq t$

$x_{ij} \in \{0, 1\}$

Cutting Stock Problem [ME p. 26]

A paper mill manufactures rolls of paper of a standard width 3 meters. But customers want to buy paper rolls of shorter width, and the mill has to cut such rolls from the 3 m rolls. One 3 m roll can be cut, for instance, into two rolls 93 cm wide, one roll of width 108 cm, and a rest of 6 cm (which goes to waste).

Let us consider an order of

- 97 rolls of width 135 cm,
- 610 rolls of width 108 cm,
- 395 rolls of width 93 cm, and
- 211 rolls of width 42 cm.



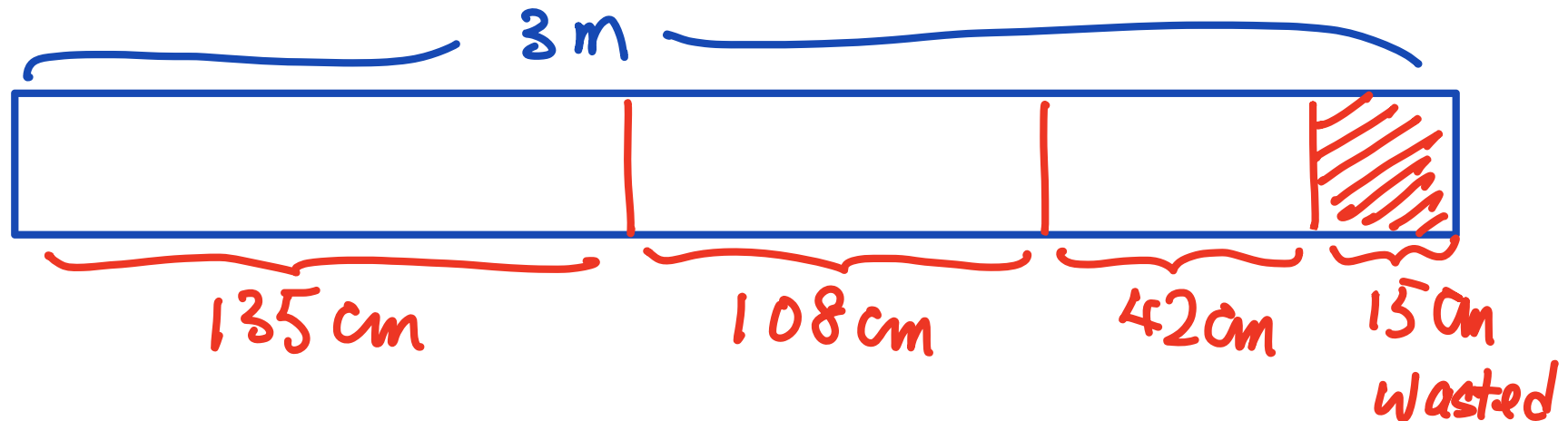
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e.g.



Cutting Stock Problem [MB, p. 26]

In order to engage linear programming one has to be generous in introducing variables. We write down all of the requested widths: 135 cm, 108 cm, 93 cm, and 42 cm. Then we list all possibilities of cutting a 3 m paper roll into rolls of some of these widths (we need to consider only possibilities for which the wasted piece is shorter than 42 cm):

$$P1: 2 \times 135$$

$$P2: 135 + 108 + 42$$

$$P3: 135 + 93 + 42$$

$$P4: 135 + 3 \times 42$$

$$P5: 2 \times 108 + 2 \times 42$$

$$P6: 108 + 2 \times 93$$

$$P7: 108 + 93 + 2 \times 42$$

$$P8: 108 + 4 \times 42$$

$$P9: 3 \times 93$$

$$P10: 2 \times 93 + 2 \times 42$$

$$P11: 93 + 4 \times 42$$

$$P12: 7 \times 42$$

$$258 \leq 135 n_1 + 108 n_2 + 93 n_3 + 42 n_4 \leq 300$$

$$0 \leq n_i, \text{ integers}$$

Cutting Stock Problem [MB p. 26]

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$$P4: 135 + 3 \times 42$$

$$P5: 2 \times 108 + 2 \times 42$$

$$P6: 108 + 2 \times 93$$

$$P7: 108 + 93 + 2 \times 42$$

$$P8: 108 + 4 \times 42$$

$$P9: 3 \times 93$$

$$P10: 2 \times 93 + 2 \times 42$$

$$P11: 93 + 4 \times 42$$

$$P12: 7 \times 42$$

$$x_i = \# \text{ of } P_i, i=1, \dots, 12$$

$$x_i \geq 0, \text{ integers}$$

$$\underline{\min \quad x_1 + x_2 + \dots + x_{12}}$$

Cutting Stock Problem

[MG p. 26]

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$$P10: 2 \times 93 + 2 \times 42$$

$$P11: 93 + 4 \times 42$$

$$P12: 7 \times 42$$

$$2x_1 + x_2 + x_3 + x_4 \geq 97$$

$$x_2 + 2x_5 + x_6 + x_7 + x_8 \geq 610$$

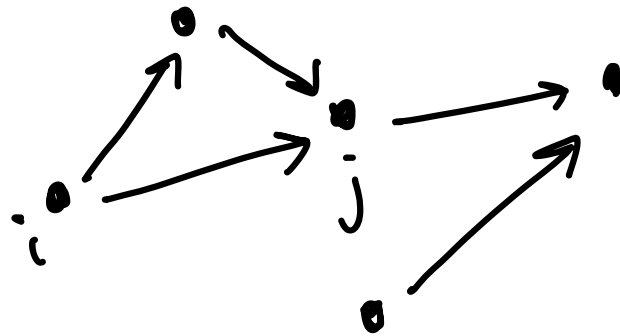
$$x_3 + 2x_6 + x_7 + 2x_9 + x_{11} \geq 395$$

$$x_2 + x_3 + x_4 + 2x_5 + 2x_7 + 4x_8 + 2x_{10} + 4x_{11} + 7x_{12} \geq 211$$

Network Flow ([V] Chapter 14)

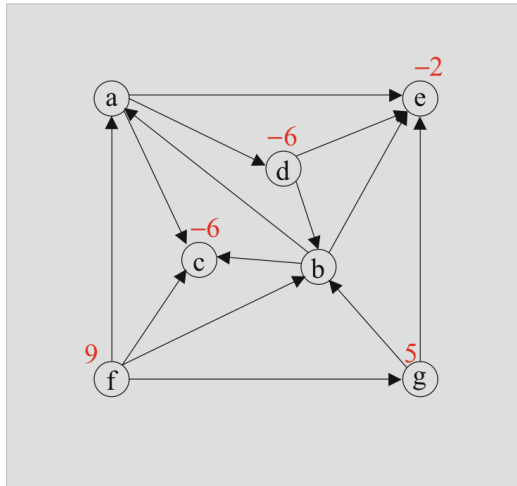
N = set of nodes $\{i\}$

A = set of (directed) arcs
 $\subseteq \{(i,j) : i,j \in N\}$



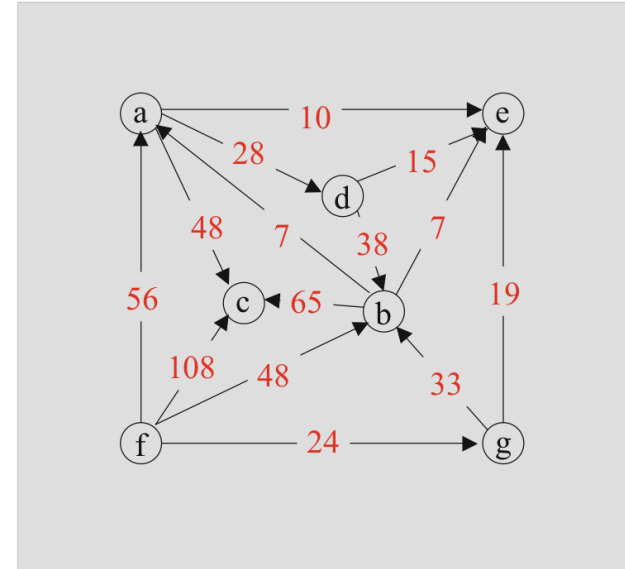
Network = (N, A)
graph (or digraph)

Network Flow ([V] Chapter 14)



b_i

FIGURE 14.1. A network having 7 nodes and 14 arcs. The numbers written next to the nodes denote the supply at the node (negative values indicate demands; missing values indicate no supply or demand).

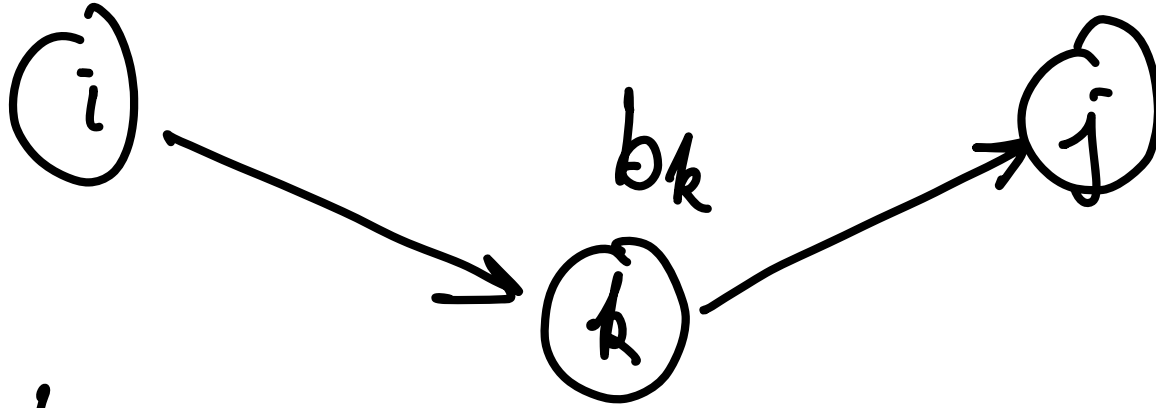


c_{ij}

FIGURE 14.2. The costs on the arcs for the network in Figure 14.1.

- (1) $b_i > 0$ (supply) ; $b_i < 0$ (demand) $\sum_{i \in N} b_i = 0$
- (2) x_{ij} = amount transported from i to j
- (3) c_{ij} = cost of transportation from i to j
- (4) $\min \sum_{(i,j) \in A} c_{ij} x_{ij}$

Network Flow ([V] Chapter 14)

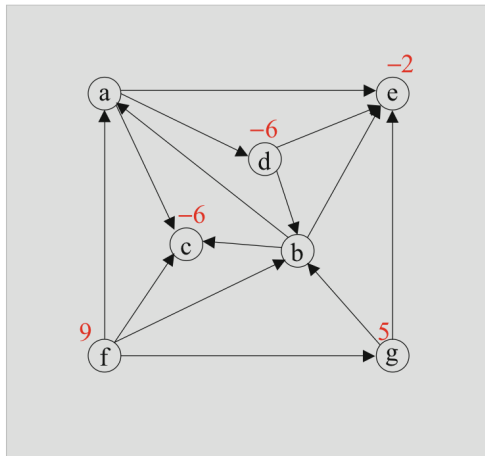


Balanced eqn at node k : $\sum_i x_{ik} + b_k = \sum_j x_{kj}$

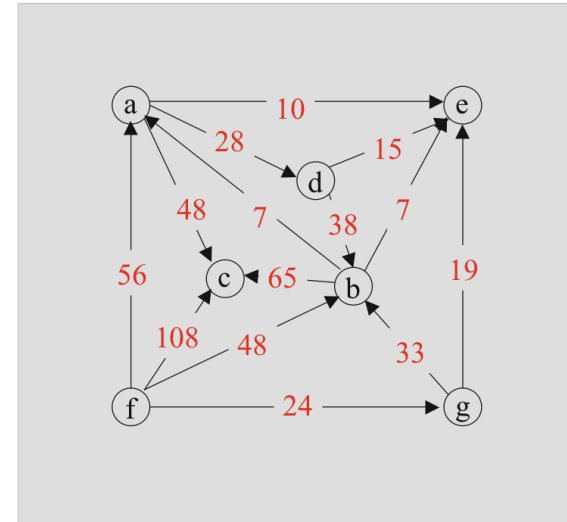
$$\sum_i x_{ik} - \sum_j x_{kj} = -b_k$$

$$\begin{aligned} \min \quad & c^T x \\ \text{s.t.} \quad & Ax = -b, \quad x \geq 0 \end{aligned}$$

Network Flow ([V] Chapter 14)



b_i



c_ij

FIGURE 14.1. A network having 7 nodes and 14 arcs. The numbers written next to the nodes denote the supply at the node (negative values indicate demands; missing values indicate no supply or demand).

FIGURE 14.2. The costs on the arcs for the network in Figure 14.1.

arcs

$$x^T = [x_{ac} \ x_{ad} \ x_{ae} \ x_{ba} \ x_{bc} \ x_{be} \ x_{db} \ x_{de} \ x_{fa} \ x_{fb} \ x_{fc} \ x_{fg} \ x_{gb} \ x_{ge}] ,$$

incidence matrix

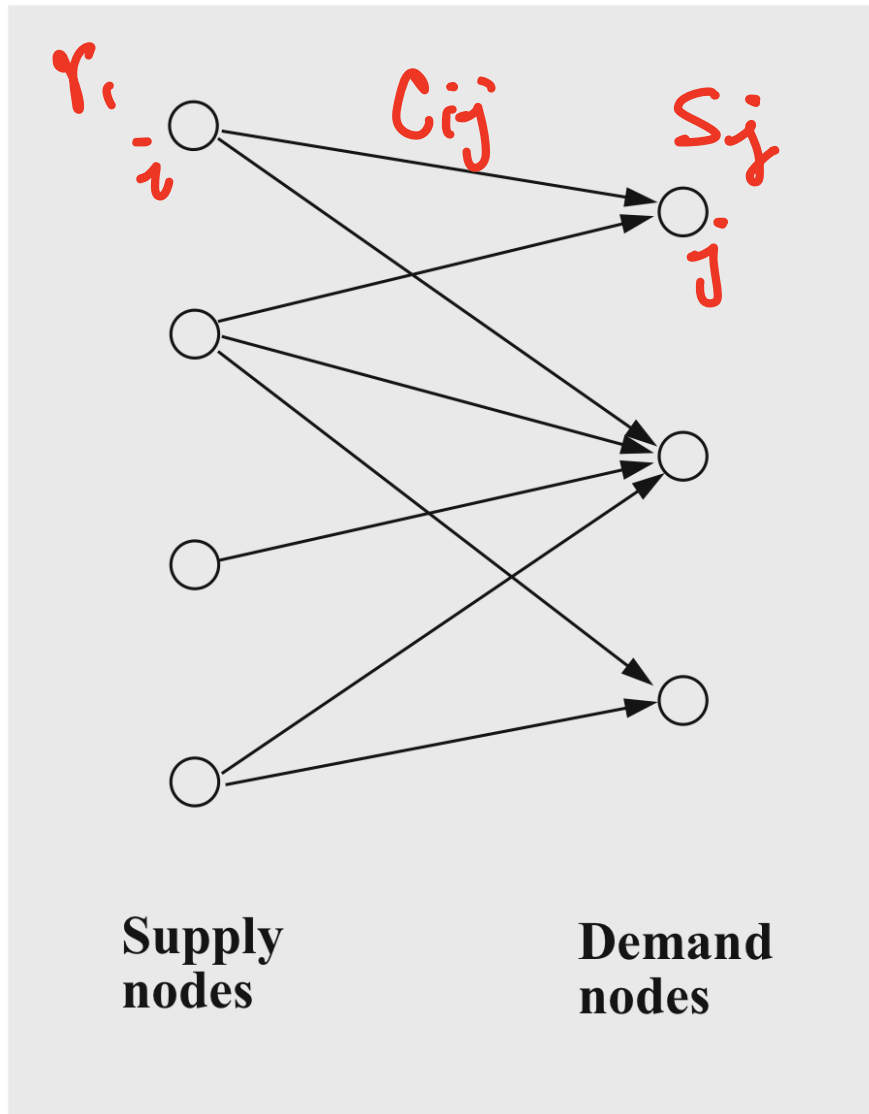
$$A = \begin{bmatrix} -1 & -1 & -1 & 1 & & & & & & & & & & \\ & & & -1 & -1 & -1 & 1 & & & & & & & \\ & 1 & & & 1 & & & & & & & & & \\ & & 1 & & & & -1 & -1 & & & & & & \\ & & & 1 & & & 1 & & 1 & & & & & \\ & & & & & & & -1 & -1 & -1 & -1 & & & \\ & & & & & & & & & 1 & -1 & -1 & & \end{bmatrix} , \quad b = \begin{bmatrix} 0 \\ 0 \\ -6 \\ -6 \\ -2 \\ 9 \\ 5 \end{bmatrix}$$

c^T

$$c^T = [48 \ 28 \ 10 \ 7 \ 65 \ 7 \ 38 \ 15 \ 56 \ 48 \ 108 \ 24 \ 33 \ 19] .$$

(a)
(b)
(c)
(d)
(e)
(f)
(g)

Transportation Problem [V] p. 258



$$\begin{aligned} & \text{minimize} && \sum_{i \in \mathcal{S}} \sum_{j \in \mathcal{D}} c_{ij} x_{ij} \\ & \text{subject to} && \sum_{j \in \mathcal{D}} x_{ij} = r_i && i \in \mathcal{S} \\ & && \sum_{i \in \mathcal{S}} x_{ij} = s_j && j \in \mathcal{D} \\ & && x_{ij} \geq 0 && i \in \mathcal{S}, j \in \mathcal{D}. \end{aligned}$$

(bipartite graph)

A Miracle

The matrices for assignment, network flow (and transportation) problems are totally uni-modular

and hence

given integer supplies and demands, the solutions are automatically integers.