

Summary of MA 421, Fall 2021

$$\begin{array}{ll} \text{(LP)} & \text{(P)} \quad \max \quad \tilde{J} = c^T x \\ & \text{s.t.} \quad Ax \leq b \\ & \quad \quad x \geq 0 \end{array} \quad \longleftrightarrow \quad \begin{array}{ll} \text{(D)} & \min \quad \tilde{J} = b^T y \\ & \text{s.t.} \quad A^T y \geq c \\ & \quad \quad y \geq 0 \end{array}$$

- (1) Simplex Method, dictionary, basic vs non-basic vars.
Various pivoting rules
- (2) Duality, complementary slackness
- (3) Primal vs Dual Simplex, applications, Phase I, II.
- (4) Sensitivity Analysis

(See also note on Tuesday, Dec 3, 2024)

7. A Dual-Based Phase I Algorithm

The dual simplex method described in the previous section provides us with a new Phase I algorithm, which if nothing else is at least more elegant than the one we gave in Chapter 2. Let us illustrate it using an example:

$$\begin{aligned}
 &\text{maximize} && -x_1 + 4x_2 \\
 &\text{subject to} && -2x_1 - x_2 \leq 4 \\
 &&& -2x_1 + 4x_2 \leq -8 \\
 &&& -x_1 + 3x_2 \leq -7 \\
 &&& x_1, x_2 \geq 0.
 \end{aligned}$$

(P)

$\zeta =$			-1	x_1	$+4$	x_2
$w_1 =$	4	$+$	2	x_1	$+$	x_2
$w_2 =$	-8	$+$	2	x_1	-4	x_2
$w_3 =$	-7	$+$		x_1	-3	x_2

(D)

$-\xi =$			-4	y_1	$+8$	y_2	$+7$	y_3
$z_1 =$	1	-2	y_1	-2	y_2	$-$		y_3
$z_2 =$	-4	$-$	y_1	$+4$	y_2	$+3$		y_3

Neither primal or dual dictionaries are feasible.

7. A Dual-Based Phase I Algorithm

The dual simplex method described in the previous section provides us with a new Phase I algorithm, which if nothing else is at least more elegant than the one we gave in Chapter 2. Let us illustrate it using an example:

$$\begin{aligned}
 &\text{maximize } -x_1 + 4x_2 \\
 &\text{subject to } -2x_1 - x_2 \leq 4 \\
 &\quad \quad \quad -2x_1 + 4x_2 \leq -8 \\
 &\quad \quad \quad -x_1 + 3x_2 \leq -7 \\
 &\quad \quad \quad x_1, x_2 \geq 0.
 \end{aligned}$$

change to $\zeta = -x_1 - x_2$ and then consider its dual

(P)

$\zeta =$		-	1	x_1	+	4	x_2
$w_1 =$	4	+	2	x_1	+		x_2
$w_2 =$	-8	+	2	x_1	-	4	x_2
$w_3 =$	-7	+		x_1	-	3	x_2

(D)

$-\xi =$		-	4	y_1	+	8	y_2	+	7	y_3
$z_1 =$	1	-	2	y_1	-	2	y_2	-		y_3
$z_2 =$	-4	-		y_1	+	4	y_2	+	3	y_3

Neither primal or dual dictionaries are feasible.

$$\begin{array}{rcl}
 \text{(D)} & -\xi = & -4y_1 + 8y_2 + 7y_3 \\
 & \hline
 z_1 = & 1 - 2y_1 - 2y_2 - & y_3 \\
 z_2 = & \cancel{-4} - & y_1 + 4y_2 + 3y_3
 \end{array}$$

1

① y_2 enters, z_1 leaves:

$$y_2 = \frac{1}{2} - y_1 - \frac{1}{2}z_1 - \frac{1}{2}y_3$$

$$z_2 = 1 - y_1 + 2 - 4y_1 - 2z_1 - 2y_3 + 3y_3$$

$$= 3 - 5y_1 - 2z_1 + y_3$$

$$-\xi = -4y_1 + 4 - 8y_1 - 4z_1 - 4y_3 + 7y_3$$

$$= 4 - 12y_1 - 4z_1 + 7y_3$$

$$\begin{array}{rcl}
 \text{(D)} & -\xi = & -4y_1 + 8y_2 + 7y_3 \\
 & \hline
 z_1 = & 1 - 2y_1 - 2y_2 - & y_3 \\
 z_2 = & \cancel{-4} - & y_1 + 4y_2 + 3y_3
 \end{array}$$

1

② y_3 enters, y_2 leaves:

$$y_3 = 1 - 2y_1 - z_1 - 2y_2$$

$$\begin{aligned}
 z_2 &= 3 - 5y_1 - 2z_1 + 1 - 2y_1 - z_1 - 2y_2 \\
 &= 4 - 7y_1 - 3z_1 - 2y_2
 \end{aligned}$$

$$\begin{aligned}
 -\xi &= 4 - 12y_1 - 4z_1 + 7 - 14y_1 - 7z_1 - 14y_2 \\
 &= 11 - 26y_1 - 11z_1 - 14y_2 \quad \text{Opt!}
 \end{aligned}$$

$$\begin{array}{rcl}
 \text{(D)} & -\xi = & -4y_1 + 8y_2 + 7y_3 \\
 & \hline
 & z_1 = & 1 - 2y_1 - 2y_2 - y_3 \\
 & z_2 = & \cancel{-4} - y_1 + 4y_2 + 3y_3
 \end{array}$$

1

Opt. Dict for Dual Problem:

$$\begin{aligned}
 -\xi &= 11 - 26y_1 - 11z_1 - 14y_2 \\
 z_3 &= 1 - 2y_1 - z_1 - 2y_2 \\
 z_2 &= 4 - 7y_1 - 3z_1 - 2y_2
 \end{aligned}$$

$$\begin{array}{rcl}
 \text{(D)} & -\xi = & -4y_1 + 8y_2 + 7y_3 \\
 & \hline
 z_1 = & 1 & -2y_1 - 2y_2 - y_3 \\
 z_2 = & \cancel{-4} & -y_1 + 4y_2 + 3y_3
 \end{array}$$

1

Opt. Dict for Primal Problem:

$$w_1 = 26 + 2w_3 + 7x_2$$

$$x_1 = 11 + w_3 + 3x_2$$

$$w_2 = 14 + 2w_3 + 2x_2$$

$$J = -x_1 + 4x_2 \quad (\text{Original objective})$$

$$= 11 + w_3 + 2x_2 + 4x_2$$

$$= 11 + w_3 + 6x_2$$

Ready for Phase II

$$\left(\begin{array}{ccc} -26 & -11 & -14 \\ -2 & -1 & -2 \\ -7 & -3 & -2 \end{array} \right)^T$$

Convex Analysis

- (1) Definition of convex sets, convex comb.
convex hull
- (2) Carathéodory Theorem
- (3) Separation Theorem
- (4) Farkas Lemma (and its variants)

LEMMA 10.5. The system $Ax \leq b$ has no solutions if and only if there is a y such that

$$\begin{aligned} (10.8) \quad & A^T y = 0 \\ & y \geq 0 \\ & b^T y < 0. \end{aligned}$$

how to find y 

Convex Analysis

- (1) Definition of convex sets, convex comb.
convex hull
- (2) Carathéodory Theorem
- (3) Separation Theorem
- (4) Farkas Lemma (and its variants)

$$\begin{array}{ll} \text{maximize} & 0 \\ \text{subject to} & Ax \leq b \end{array} \quad \text{vs} \quad \begin{array}{ll} \text{minimize} & b^T y \text{ (} < 0 \text{)} \\ \text{subject to} & A^T y = 0 \\ & y \geq 0. \end{array}$$

Applications of LP

- (1) Regression, L^1 , L^2 , L^∞
- (2) Binary Classification
- (3) Maximum inscribed circle
- (4) Formulations of problems in terms of LP...
- (5) Integer Programming
(Branch-and-Bound and Gomory Cuts)

Networks

(1) Graph, nodes, arcs, spanning tree

(2) Network flow problem: $\min C^T X$
(Transshipment) s.t. $AX = b$
 $X \geq 0$

A - incidence matrix

(3) Dual Network flow problem: $\max -b^T y$
s.t. $A^T y + z = c$
 $z \geq 0$

Given a spanning tree $\Rightarrow$ find X, y, z .
 x_{ij}, y_i, z_{ij}

Networks

(4) Primal vs Dual Network Simplex method

- (i) Spanning tree as basic variables
- (ii) Determine entering and leaving arcs
- (iii) Updating x_{ij} , y_i , z_{ij}

(5) Variants of Network Flow problems

- (i) Transportation (Bipartite graph)
- (ii) Assignment
- (iii) Inequality Constraints
- (iv) Upper bounded flows

Algorithms specific to Networks/Graphs

(1) Shortest Path (Bellman vs Dijkstra)

Dynamics programming: Bellman Egn

Let v_i = shortest distance from i to r

Then
$$v_i = \min_j \{ C_{ij} + v_j \}$$



Algorithms specific to Networks/Graphs

(1) Shortest Path (Bellman vs Dijkstra)

Dynamics programming: Bellman Egn

$$v_i^0 = \begin{cases} \infty & i \neq r \\ 0 & i = r \end{cases}$$

$$v_i^{(k)} = \min_j \{ C_{ij} + v_j^{(k)} \}, \quad v_r^{(k)} = 0$$

Algorithms specific to Networks/Graphs

11) Shortest Path (Bellman vs Dijkstra)

Initialize:

$$\mathcal{F} = \emptyset$$

$$v_j = \begin{cases} 0 & j = r, \\ \infty & j \neq r. \end{cases}$$

while ($|\mathcal{F}^c| > 0$) {

$$j = \operatorname{argmin}\{v_k : k \notin \mathcal{F}\}$$

$$\mathcal{F} \leftarrow \mathcal{F} \cup \{j\}$$

for each i for which $(i, j) \in \mathcal{A}$ and $i \notin \mathcal{F}$ {

if $(c_{ij} + v_j < v_i)$ {

$$v_i = c_{ij} + v_j$$

$$h_i = j$$

}

}

}

Algorithms specific to Networks/Graphs

(2) Max-Flow-Min-Cut

Ford-Fulkerson (Augmented Path)

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

[C] p. 368

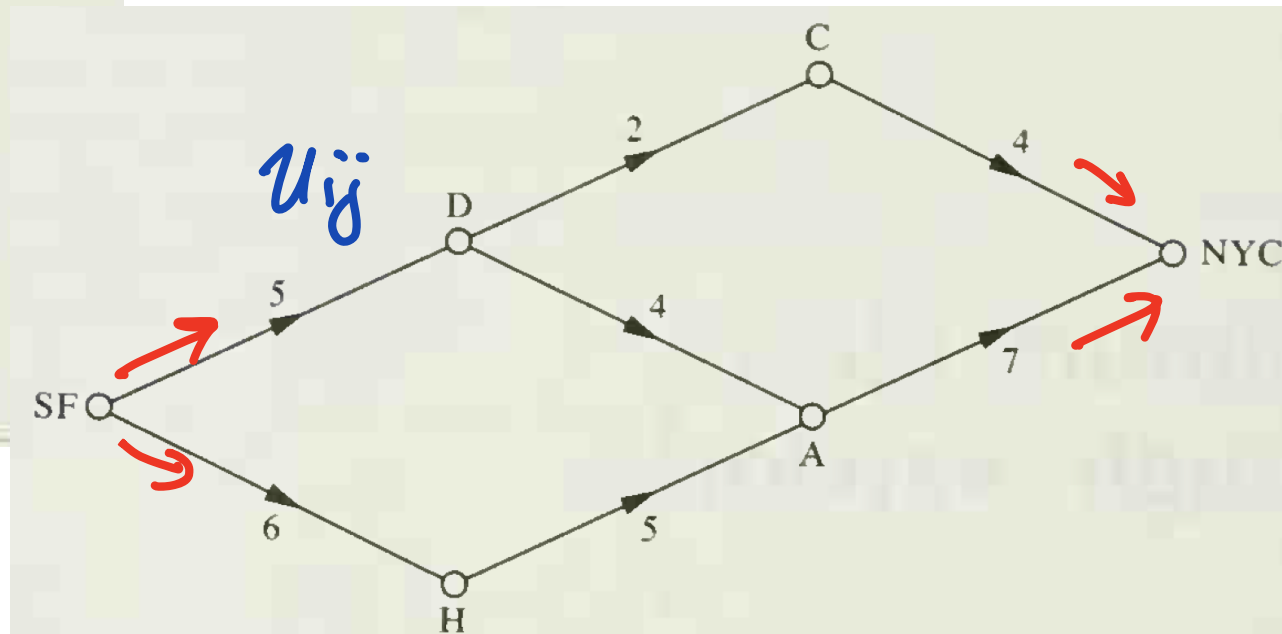


Figure 22.1

Algorithms specific to Networks/Graphs

(3) Minimum Spanning Tree (MST)

Properties of a (Spanning) tree

Theorem 7.1

- (a) Every tree with more than one node has at least one leaf.
- (b) An undirected graph is a tree if and only if it is connected and has $|\mathcal{N}| - 1$ arcs.
- (c) For any two distinct nodes i and j in a tree, there exists a unique path from i to j .
- (d) If we start with a tree and add a new arc, the resulting graph contains exactly one cycle (as long as we do not distinguish between cycles involving the same set of nodes).

Algorithms specific to Networks/Graphs

(3) Minimum Spanning Tree (MST) (Greedy Algorithm)

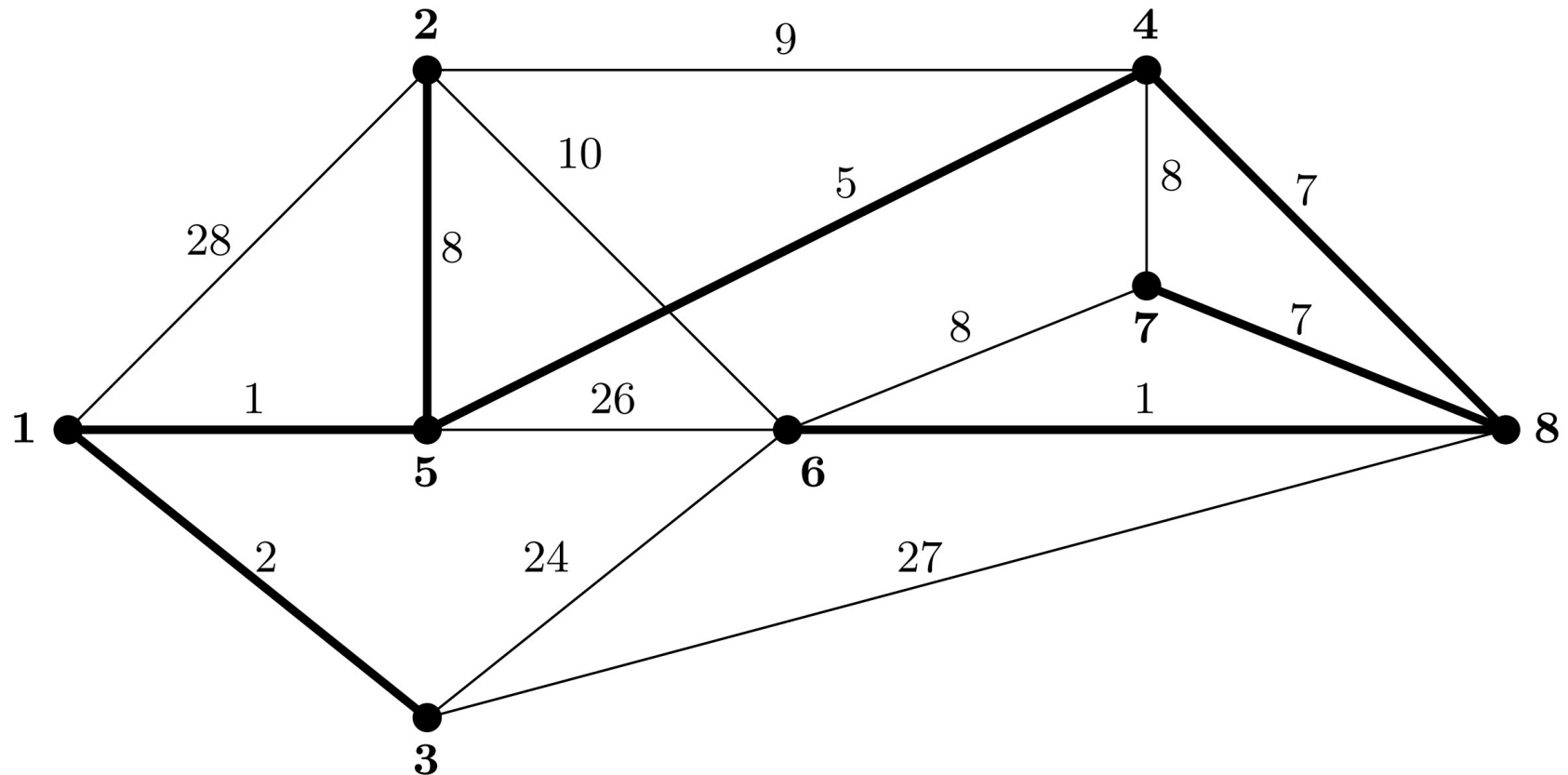
Greedy algorithm for the minimum spanning tree problem

1. The input to the algorithm is a connected undirected graph $G = (\mathcal{N}, \mathcal{E})$ and a coefficient c_e for each edge $e \in \mathcal{E}$. The algorithm is initialized with a tree $(\mathcal{N}_1, \mathcal{E}_1)$ that has a single node and no edges ($\mathcal{E}_1$ is empty).
2. Once $(\mathcal{N}_k, \mathcal{E}_k)$ is available, and if $k < n$, we consider all edges $\{i, j\} \in \mathcal{E}$ such that $i \in \mathcal{N}_k$ and $j \notin \mathcal{N}_k$. Choose an edge $e^* = \{i, j\}$ of this type whose cost is smallest. Let

$$\mathcal{N}_{k+1} = \mathcal{N}_k \cup \{j\}, \quad \mathcal{E}_{k+1} = \mathcal{E}_k \cup \{e^*\}.$$

Algorithms Specific to Networks/Graphs

(3) Minimum Spanning Tree (MST) (Greedy Algorithm)



Algorithms Specific to Networks/Graphs

(3) Minimum Spanning Tree (MST) (Greedy Algorithm)

