

Solution outline to Homework 7

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1 Problem 7.3

There are three materials 1, 2, 3. The volume variables associated with them are x_1, x_2, x_3 . Constraints are as follows

$$\begin{aligned}2x_1 + 1x_2 + 3x_3 &\leq 100 \\x_1 + x_2 + x_3 &\leq 60 \\x_1 &\leq 40 \\x_2 &\leq 30 \\x_3 &\leq 20 \\x_1, x_2, x_3 &\geq 0\end{aligned}$$

The cost function is given as

$$\text{Maximize } 1000x_1 + 1200x_2 + 12000x_3$$

2 Problem 7.8

I had already sent portions of the formulation in my email. Points denote the set of all points where each point is a 2-tuple (x_i, y_i) . Constraints of the linear program are as follows

$$\begin{aligned}\forall i \in \text{Points} \\e_i &\geq ax_i + by_i - c \\e_i &\geq -(ax_i + by_i - c) \\\forall i \in \text{Points} \\M &\geq e_i\end{aligned}$$

The cost function is given by

$$\text{Minimize } M$$

3 Problem 7.11

Dual LP

$$\begin{aligned}\text{Minimize } & 3\lambda + 5\nu \\ \text{Subject to } & \\ & 2\lambda + \nu \geq 1 \\ & \lambda + 3\nu \geq 1 \\ & \lambda, \nu \geq 0\end{aligned}$$

Solution of primal LP (Solve graphically) is $x = \frac{4}{5}$, $y = \frac{7}{5}$. Similarly solution of dual LP is $\lambda = \frac{2}{5}$, $\nu = \frac{1}{5}$. By LP duality verify that at these points the cost functions of primal and dual LP computes to the same value of 2.2.

4 Problem 7.17

- (a) Max flow : $4+2+2+3 = 11$, Min Cut : $\{s,a,b\}$, $\{c,d,t\}$
- (b) Vertices reachable from S are $\{A,B\}$ while vertices from T are $\{C\}$
- (c) Bottleneck Edge Identification: $\{AC,BC\}$
- (d) The example is shown in Figure 1
- (e) Algorithm outline for Bottleneck Edge Identification: Generate the residual graph after the ford-fulkerson

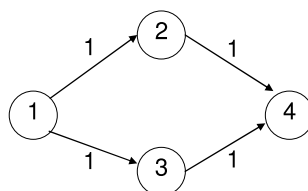


Figure 1: Counterexample for Shortest Path Tree

algorithm is run. Consider each edge $(source, sink)$ in the residual graph where $source, sink \in V$. If s is reachable from $source$ and t is reachable from $sink$ then $(source, sink)$ is a bottleneck edge. Reachability property proves the correctness of the algorithm since if s and t are not reachable from $source$ and $sink$ it is not possible to augment more flow on the path from s to t and then increasing the capacity of edge $(source, sink)$ doesn't increase the maximum flow.

5 Problem 7.25

- (a) We define flow variables f_{ij} on each edge $(i, j) \in E$ of the graph $G=(V,E)$. Based on the that maximum flow from S to T can be written as a linear program as follows

$$\begin{aligned}
 &\text{Maximize } f_{sa} + f_{sb} \\
 &\text{Subject to} \\
 &\quad f_{sa} \leq 1 \\
 &\quad f_{sb} \leq 3 \\
 &\quad f_{ab} \leq 1 \\
 &\quad f_{at} \leq 2 \\
 &\quad f_{bt} \leq 1 \\
 &\quad f_{sa} = f_{ab} + f_{at} \\
 &\quad f_{sb} + f_{ab} = f_{bt} \\
 &\quad f_{sa}, f_{sb}, f_{ab}, f_{at}, f_{bt} \geq 0
 \end{aligned}$$

(b) Dual of the linear program: We have dual variables λ_{ij} associated with each edge $(i, j) \in E$ and ν_i associated with each vertex $i \in V$ in the graph $G=(V, E)$.

$$\begin{aligned}
& \text{Minimize } \lambda_{sa} + 3\lambda_{sb} + \lambda_{ab} + 2\lambda_{at} + \lambda_{bt} \\
& \text{Subject to} \\
& \lambda_{sa} + \nu_a \geq 1 \\
& \lambda_{sb} + \nu_b \geq 1 \\
& \lambda_{ab} - \nu_a + \nu_b \geq 0 \\
& \lambda_{at} - \nu_a \geq 0 \\
& \lambda_{bt} - \nu_b \geq 0 \\
& \lambda_{sa}, \lambda_{sb}, \lambda_{ab}, \lambda_{at}, \lambda_{bt} \geq 0 \\
& \nu_a, \nu_b \text{ Unrestricted}
\end{aligned}$$

(c) General dual formulation can be written as

$$\text{Minimize } \sum_{e \in E} c_e y_e \quad (1)$$

$$\text{Subject to} \quad (2)$$

$$y_{ij} - x_i + x_j \geq 0 \forall (i, j) \in E, i, j \notin \{s, t\} \quad (3)$$

$$y_{si} + x_i \geq 1 \forall (s, i) \in E \quad (4)$$

$$y_{jt} - x_j \geq 0 \forall (j, t) \in E \quad (5)$$

$$y_{ij} \geq 0 \forall (i, j) \in E \quad (6)$$

$$(7)$$

(d) Summing up constraints 1-3 we can obtain that

$$\sum_{(i,j) \in E} y_{ij} \geq 1$$

(e) x_u is the potential defined on the vertices. y_e represents if the specified edge is on a cut. Given a graph $G=(V, E)$, for any cut $V = V1 \cup V2$. For an edge (u, v) if $u \in V1$ and $v \in V2$ then $y_{(u,v)} = 1$ otherwise $y_{(u,v)}$ is 0. For all nodes $u \in V1$ set x_u as 1 and for all nodes in $u \in V2$ set x_u as 0. The assignment satisfies the constraints given and the cost function $\sum_{e \in E} c_e y_e$ exactly sums up the capacities of edges in a cut.

6 Problem 7.31

(a) 2000 augmenting paths can be found out each sending a single unit of flow in this particular graph.

(b) We define the capacity of a s-t path as the smallest capacity of its constituent edges. We define a variable cap on each vertex.

procedure IdentifyFattestPath(G, c_e, s)

Input: Graph $G=(V, E)$, c_e capacity of edge e

source vertex s

Output: Fattest s-t path

for all vertices $v \in V$:

cap(v) = ∞

prev(v) = nil

H = makequeue(V) (using cap values as keys)

while H is not empty:

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u = deletemax(H)
for all edges (u,v) ∈ E
    if cap(v) < min{cap(u), c(u,v)}
        cap(v) = min{cap(u), c(u,v)}
        prev(v) = u
        decreasekey(H,v)

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(c) Using the duality of Maximum flow problem we know that there is a min-cut associated with the maximum flow. Now min-cut divides the vertices V into two sets V_1 and V_2 . Each edge e in the cut must be an edge of some $s - t$ path in the graph and the flow associated with this path is capacity of the edge c_e . Therefore the maximum flow must be the sum of the capacity of edges corresponding to min-cut edges e . Since edges in a min-cut are bounded by $|E|$, therefore maximum flow can be decomposed into at most $|E|$ paths.

(d) Since F is the sum of the flow over at most $|E|$ paths, therefore each flow augmentation decreases the maximum flow bound by $\frac{1}{|E|}$. After k iteration the maximum flow bound is given by

$$F(1 - \frac{1}{|E|})^k$$

Following discussion in greedy set cover algorithm, k is given by $O(|E| \log F)$.