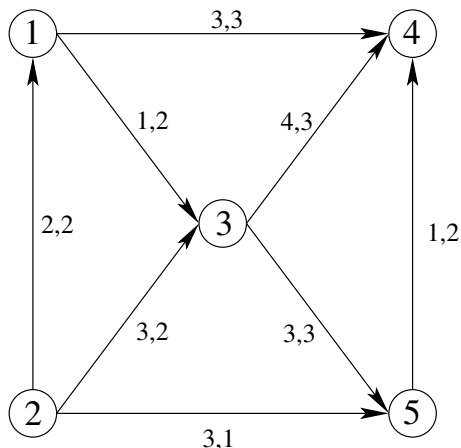


Assignment 1

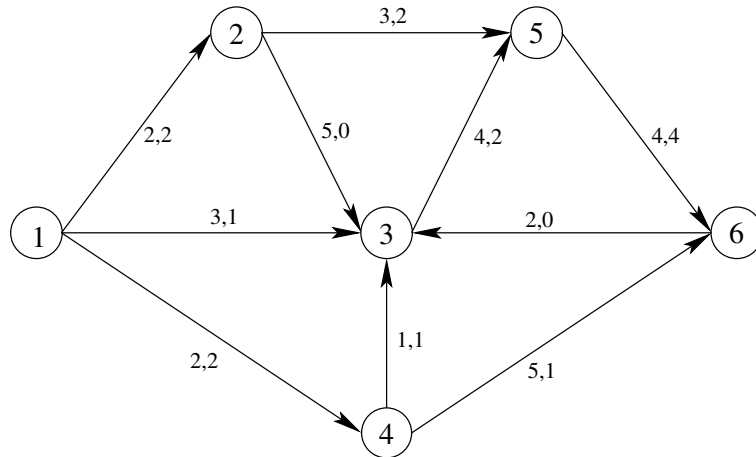
In the graph below, the arcs are labeled with cost and upper bound as follows $\xrightarrow{c,u}$. The lower bounds are all zero. Node 1 is sources of strengths 1 and node 2 is sources of strengths 2, while node 4 is a sink of strength 3.



- (1p) a) Given the flow: $x_{13} = x_{54} = 1$, $x_{23} = x_{34} = 2$ and all the other $x_{ij} = 0$. Is it a basic feasible solution?
- (3p) b) Make two iterations of the network simplex method starting from the basic feasible solution in which $x_{14} = x_{23} = x_{25} = x_{34} = x_{54} = 1$ and all the other $x_{ij} = 0$.
- (1p) c) Motivate why the obtained flow is optimal or not.
- (2p) d) Find ALL values of c_{25} for which it is guaranteed that the flow cost obtained in b) cannot be decreased.

Assignment 2

Consider the maximum flow problem defined by the graph in the figure below, in which the arcs are labeled with capacity and flow as follows $\xrightarrow{u,x}$.



The lower bounds are all zero. The source is node 1, and the sink is node 6.

- (2p) a) Find capacity of the cut $N_s = \{1, 4, 5\}$. Find flow across this cut.
- (3p) b) Use the max-flow algorithm for finding the maximum flow from 1 to 6. Start from the given flow. Present the results of using the modified Dijkstra's algorithm for finding a flow increasing path from 1 to 6. Show how arc's feasible directions are updated.
- (2p) c) Find a minimum cut in this graph.

Assignment 3

Consider the following integer knapsack problem.

$$\begin{aligned} \max \quad & x_1 + 2x_2 + 3x_3 \\ \text{s.t.} \quad & 3x_1 + 4x_2 + 5x_3 \leq a \\ & x_1, x_2, x_3 \geq 0, \quad \text{integer,} \end{aligned}$$

where a is an integer parameter.

- (1p) a) Use dynamic programming to find *ALL* optimal solutions x^* for $a=7$.
- (3p) b) Use dynamic programming to find all optimal solutions x^* and z^* for each value of $a = 3, 4, 5, 6, 7, 8$.

Assignment 4

Consider the convex function

$$f(x) = \max\{-x, x/2, x - 1\}.$$

Use the graph of this function to answer to the following questions.

- (1p) a) What is the subdifferential for $x = 0$?
- (3p) b) What is the subdifferential for $x = 2$?
- (2p) c) Use the subdifferential in $x = 0$ to motivate that it is a global minimizer.

Assignment 5

- (3p) a) Consider the problem of minimizing a strictly convex quadratic function. Show that the first two directions generated by the conjugate gradient method are conjugate.
- (3p) b) Prove that the conjugate directions are linearly independent.
- (2p) c) Consider the problem of minimizing the function $f(x) = x_1^2 - x_2^2/2 - 4x_1 - x_2$. Use the Levenberg-Marquardt method to generate a search direction at the point $\bar{x} = (0, 0)^T$.

Assignment 6 (4p)

Consider the quadratic programming problem

$$\begin{aligned} \min \quad & f(x) = x_1^2/2 + 2x_2^2 \\ \text{s.t.} \quad & x_1 \geq 2x_2 + 1 \\ & x_2 \geq 0 \end{aligned}$$

Solve it using the active set method starting from the point $x^{(0)} = (4, 1)^T$.

Assignment 7

Consider the problem

$$\begin{aligned} \min \quad & x \\ \text{s.t.} \quad & (x - 1)(x - 2) \leq 0 \end{aligned}$$

Let x_* denote the solution to this problem. The quadratic penalty method and the logarithmic barrier method reduce this constrained optimization problem to a sequence of unconstrained minimization of a function $F(x, c)$ for a series of values of the parameter c . Let $x(c)$ denote a minimizer of $F(x, c)$.

- (3p) a) Find $x(c)$ for the logarithmic barrier method and show that $x(c) \rightarrow x_*$.
- (1p) b) Consider the problem in Assignment 6. For the quadratic penalty method, find the value of $F(x, 1)$ in the point $x = (2, 1)^T$.

ANSWERS

Assignment 1

- a) It is not a basic feasible solution, because the flow is infeasible (the conservation flow is violated in nodes 3 and 5).
- b) The basic tree is composed of the edges (1,4), (2,3), (3,4), (5,4). The node prices $y = (0, -4, -1, 3, 2)^T$. Reduced costs and optimality:
 $\bar{c}_{13} = 2 \ \& \ x_{13} = l_{13} \Rightarrow \text{OK}$
 $\bar{c}_{21} = -2 \ \& \ x_{21} = l_{21} \Rightarrow \text{NOT OK}$
 $\bar{c}_{25} = -3 \ \& \ x_{25} = u_{25} \Rightarrow \text{OK}$
 $\bar{c}_{35} = 0 \ \& \ x_{35} = l_{35} \Rightarrow \text{OK}$
Then x_{21} enters the basis. This results in the cycle 2–1–4–3–2. It is possible to send 1 unit in the cycle with x_{23} exiting the basis. (The alternative is that x_{34} is exiting the basis.)
- c) Consider the case when x_{23} is exiting the basis. The new node prices $y = (0, -2, -1, 3, 2)^T$.
Reduced costs and optimality:
 $\bar{c}_{13} = 2 \ \& \ x_{13} = l_{13} \Rightarrow \text{OK}$
 $\bar{c}_{23} = 2 \ \& \ x_{23} = l_{23} \Rightarrow \text{OK}$
 $\bar{c}_{25} = -1 \ \& \ x_{25} = u_{25} \Rightarrow \text{OK}$
 $\bar{c}_{35} = 0 \ \& \ x_{35} = l_{35} \Rightarrow \text{OK}$
Thus, the optimality conditions are satisfied.
The optimal value of the total flow cost = 12.
- d) The optimality conditions give $\bar{c}_{25} = c_{25} + y_2 - y_5 \leq 0 \Rightarrow c_{25} \leq y_5 - y_2 \Rightarrow c_{25} \leq 4$.

Assignment 2

- a) The capacity of the given cut is equal to $2 + 3 + 1 + 5 + 4 = 15$. The flow across this cut is equal to $2 + 1 + 1 + 1 + 4 - 2 - 2 = 5$.
- b) The modified Dijkstra's algorithm gives the flow increasing path 1–3–4–6 whose capacity is 1. Then $f = 6$. The modified Dijkstra's algorithm shows that there is no flow increasing path. This means that the obtained flow is maximal.
- c) A minimum cut is defined by $N_s = \{1, 2, 3, 5\}$. Its capacity is equal to 6 (= max flow).

Assignment 3

- a) The dynamic programming algorithm gives the two optimal solutions:
 $x' = (0, 0, 1)^T$ and $x'' = (1, 1, 0)^T$.

- b) To find optimal solutions x^* and z^* for each value of $a = 3, 4, 5, 6, 7, 8$, the table for Stage 1 should include $s_1 = 3, 4, 5, 6, 7, 8$:

$x_1 \setminus s_1$	3	4	5	6	7	8
0	0	2	3	3	3	4
1	1	1	1	1	3	4
2	-	-	-	2	2	2
$f_1(s_1)$	1	2	3	3	3	4
$x_1^*(s_1)$	1	0	0	0	0/1	0/1

The optimal solution for each value of a :

a	x_1^*	x_2^*	x_3^*	z^*
3	1	0	0	1
4	0	1	0	2
5	0	0	1	3
6	0	0	1	3
7	0	0	1	3
7	1	1	0	3
8	0	2	0	4
8	1	0	1	4

Assignment 4

The graph of this function is defined by its equivalent presentation as

$$f(x) = \begin{cases} -x, & \text{if } x \leq 0 \\ x/2, & \text{if } 0 \leq x \leq 2 \\ x - 1, & \text{if } 2 \leq x \end{cases}$$

- a) $\partial f(0) = [-1, 0.5]$.
 b) $\partial f(2) = [0.5, 1]$.
 c) Since $0 \in \partial f(0)$, $x = 0$ is a global minimizer.

Assignment 5

- a) Consider the problem of minimizing the quadratic function

$$q(x) = x^T A x / 2 + b^T x + c,$$

where $A > 0$. Then $\nabla q(x) = Ax + b$. Let x_0 be a starting point. The first two search directions generated by the conjugate gradient method have the form

$$d_0 = -g_0, \quad d_1 = -g_1 + \beta_1(-g_0),$$

where $g_0 = \nabla q(x_0)$ and $g_1 = \nabla q(x_1)$. We need to prove that $d_1^T A d_0 = 0$, or equivalently, that $d_1^T A(x_1 - x_0) = 0$. Recalling that $A(x_1 - x_0) = g_1 - g_0$, we have

$$d_1^T A(x_1 - x_0) = (-g_1 - \beta_1 g_0)^T (g_1 - g_0).$$

The exact line search gives $g_1 \perp d_0$, that is $g_1^T g_0 = 0$. Then

$$d_1^T A(x_1 - x_0) = -\|g_1\|^2 + \beta_1 \|g_0\|^2.$$

Since $\beta_1 = \|g_1\|^2 / \|g_0\|^2$, we finally obtain $d_1^T A(x_1 - x_0) = 0$. This implies $d_1^T A d_0 = 0$, which means that d_0 and d_1 are conjugate directions.

b) For the proof, see the lecture notes.

c) In the Levenberg-Marquardt method, the search direction $d = -G^{-1} \nabla f(\bar{x})$, where $G = H + \gamma I$. The Hessian matrix $H(0,0) = \begin{pmatrix} 2 & 0 \\ 0 & -1 \end{pmatrix}$ has the eigenvalues $\lambda_1 = 2$ and $\lambda_2 = -1$ which means that $H(0,0)$ is not positively definite. Any value of $\gamma > -\min\{\lambda_1, \lambda_2\} = 1$ makes G positively definite. If to choose $\gamma = 2$, this gives $d = (1, 1)^T$.

Assignment 6

$k = 0$: $x_0 = (4, 1)^T$, $S_0 = \emptyset$, $d_0 = (-4, -1)^T$, $t = 1/2$, $x_1 = (2, 1/2)^T$.

$k = 1$: $S_1 = \{1\}$, $d_1 = (-3/2, -3/4)^T$, $v_1 = 1/2$, $t = 2/3$, $x_2 = (1, 0)^T$.

$k = 2$: $S_2 = \{1, 2\}$, $d_1 = (0, 0)^T$, $v_2 = (1, 2)^T \geq 0 \Rightarrow x = (1, 0)^T$ is the optimal solution.

Assignment 7

a) Note that the interior of the feasible region here is the interval $(1, 2)$. $F(x, c) = x - c \ln(3x - x^2 - 2)$ is a convex function of $x \in (1, 2)$ for any $c > 0$. The first derivative of this function has two roots.

The root $3/2 + c + \sqrt{c^2 + 1/4} > 2$, i.e. it is infeasible.

The root $3/2 + c - \sqrt{c^2 + 1/4} \in (1, 2)$, i.e. it is feasible.

Thus, $x(c) = 3/2 + c - \sqrt{c^2 + 1/4}$. Obviously, $x(c) \rightarrow x_* = 1$ with $c \rightarrow 0$.

b) We rewrite the inequality-type constraints in the form $g_i(x) \leq 0$, $i = 1, 2$, where $g_1(x) = 2x_2 - x_1 + 1$ and $g_2(x) = -x_2$. The quadratic penalty function $P(x) = [g_1^+(x)]^2 + [g_2^+(x)]^2$. For $x = (2 \ 1)^T$:

$$g_1(2, 1) = 1 \implies g_1^+(2, 1) = 1,$$

$$g_2(2, 1) = -1 \implies g_2^+(2, 1) = 0.$$

Thus, $P(2, 1) = 1$. For the given $x = (2 \ 1)^T$, $F(x, 1) = f(x) + 1 \cdot P(x) = 5$.