

**Optimization Methods**  
**MIT 2.098/6.255/15.093**  
**Final exam**

Date Given: December 19th, 2006

**P1. [30 pts]** Classify the following statements as true or false. All answers must be well-justified, either through a short explanation, or a counterexample. Unless stated otherwise, all LP problems are in standard form.

- (a) If there is a unique primal optimal solution to a linear programming problem, then the reduced costs of all the nonbasic variables are strictly positive.
- (b) For a network flow problem with capacity constraints, there always exists an optimal solution that is tree-structured.
- (c) When minimizing a convex function over a convex set, the optimal solution is always on the boundary of the set.
- (d) For a convex optimization problem with constraints, if a feasible point satisfies the KKT conditions then it is a global optimum.
- (e) For a nonlinear optimization problem, if Newton's method converges, then it converges to a local minimum.
- (f) For an LP in standard form, if  $\mathbf{c} \geq 0$  then the primal is either bounded or infeasible.
- (g) On very degenerate LP problems, the simplex method performs better than interior point methods.
- (h) The primal iterates  $\mathbf{x}_k$  generated by the affine scaling algorithm are always in the interior of the primal feasible set.
- (i) The feasible set of a semidefinite programming problem is always convex.
- (j) For a quadratic function  $f(x) = x^T A x + b^T x + c$ , the convergence rate of Newton's method depends on the condition number of the matrix  $A$ .

**Solution:** (a) FALSE. This is true only if the unique optimal solution is nondegenerate.

- (b) FALSE. In the capacitated case, optimal solutions do not necessarily have to be trees, a simple counterexample is a network with nodes  $\{A, B, C, D\}$  and edges  $A - B, A - C, B - D, C - D$  of unit capacity.
- (c) FALSE. Solutions can be in the interior. As an example, consider minimizing  $x^2$  on  $[-1, 1]$ .
- (d) TRUE. This is proven in the book and in the lecture notes.
- (e) FALSE. Newton's method can converge to global maxima.
- (f) TRUE. If  $c$  is nonnegative, then if the problem is feasible we have  $c^T x \geq 0$ , and thus it is bounded.
- (g) FALSE. For degenerate problems, interior point methods are a much better choice.
- (h) TRUE. By construction, the primal iterates in the affine scaling method (with  $\beta < 1$ ) are always strictly feasible.
- (i) TRUE. The feasible set of an SDP problem is convex, since it is the intersection of two convex sets (an affine subspace and the cone of PSD matrices).
- (j) FALSE. For quadratic functions, Newton's method converges exactly in one iteration.

**P2.** [25 pts] You are planning on having access to a car for the next  $N$  years, where  $N$  is a fixed number. The price of a new car is  $P$  dollars. For reliability reasons, you will only own relatively new cars, at most  $m$  years old. The yearly cost of repairing and maintaining a car during its  $k$ th year is  $r_k$ , and it satisfies  $r_1 < r_2 < \dots < r_m$  (i.e., it increases over time). At the end of any given year, you have the option of exchanging your  $k$ -year old car for a new one, with the corresponding trade-in value  $t_k$  of your old car satisfying  $t_1 > t_2 > \dots > t_m$  (i.e., depreciating over time).

You want to find the most economical sequence of buys and trade-ins, i.e., want to minimize the total cost over the  $N$ -year period. This includes all the money spent, either in buying or repairing (notice that you can sell your car at the end of the  $N$  years).

- Propose a shortest path (or network flow) formulation for this problem.
- Propose a dynamic programming formulation for this problem. Express clearly what are the state and decision variables, and the corresponding iteration.
- Use your DP formulation to solve the problem for the following data:  $N = 5$ ,  $m = 3$ ,  $P = 20000$ , the repairing costs

$$r_1 = 1200, \quad r_2 = 1600, \quad r_3 = 2400,$$

and the trade-in values

$$t_1 = 16000, \quad t_2 = 12000, \quad t_3 = 10000.$$

What is the optimal sequence of actions? Is it unique?

**Solution:** (a) Define a network whose nodes are on a  $N \times m$  grid with two additional nodes: a source node  $s$  and a sink node  $t$ . A node on the grid is indexed by (time index)  $t = 1, \dots, N$  and (car age)  $a = 1, \dots, m$ .

From node  $(t, a)$ , there is a directed edge

- to node  $(t + 1, a + 1)$  with cost  $r_a$  if  $t < N$  and  $a < m$  (car is maintained for one year)
- to node  $(t + 1, 1)$  with cost  $P - t_k + r_1$  if  $t < N$  (car is traded in)
- to a sink node  $t$  if  $t = N$  with cost  $-t_a$ .

Node  $s$  is connected to the grid node  $(1, 1)$  with cost  $P + r_1$ .

Node  $s$  a supply of 1 and node  $t$  a demand of 1.

A min-cost flow solution will correspond to an optimal purchase/maintenance plan for the car over the time horizon  $N$

- Let the time index  $t$  run from 1 to  $N$ . Define the state  $a$  at time  $t$  as the age of the car at the end of the current period. Let  $V_t(a)$  be the expected cost given that the car is  $a$ -year old at the end of the time period  $t$ .

At time  $t < N$ , if  $a = m$ , the car has to be exchanged and maintained for one year with cost  $P - t_m + r_1$ , whereas there are two available options in state  $a \in \{1, \dots, m - 1\}$ :

- maintain the car, with a cost of  $r_a$ , leading to  $(a + 1)$ -year old car at the next period
- trade-in the car for a new one and maintain it for a year, with a cost  $P - t_a + r_1$ , leading to a one-year old car in the next period.

At time  $t = N$ , the car is sold for its value  $t_a$ .

Bellman's equations for this problem are

$$\begin{aligned} V_N(a) &= -t_a, & a &= 1, \dots, m, \\ V_t(m) &= P - t_m + r_1 + V_{t+1}(1), & t &< N \\ V_t(a) &= \min(r_a + V_{t+1}(a + 1), P - t_a + r_1 + V_{t+1}(1)), & a &= 1, \dots, m - 1, t < N. \end{aligned}$$

(c) Solving Bellman's recursion yields the optimal cost of \$26,500.

**P3.** [20 pts] Consider the following transshipment problem:

The supply nodes are  $A, B$ , the demand nodes are  $D, E$ , and the transshipment node is node  $C$ . There are four unknowns  $a, b, c$  and  $d$ . The supply/demand amounts in the different nodes are:

$$A : a, \quad B : 400, \quad D : -b, \quad E : -200,$$

where as usual a positive amount indicates supply and a negative amount indicates demand. We are interested in finding an optimal (minimum cost) transshipment plan.

- (a) State conditions on  $a, b, c, d$  such that the above problem is feasible.
- (b) Consider the spanning tree given by the edges  $\{(A, D), (B, C), (C, D), (C, E)\}$ . State conditions on  $a, b, c, d$  for which the spanning tree solution will be feasible.
- (c) State conditions on  $a, b, c, d$  for which the spanning tree solution will be optimal.
- (d) State conditions on  $a, b, c, d$  for which there will be multiple solutions, including the spanning tree solution indicated above.

**Solution:** (a) For the problem to be solvable, supply and demand must balance, giving the necessary condition  $a + 400 = 200 + b$ , or equivalently,  $b - a = 200$ .

- (b) If the solution has the structure of the indicated spanning tree, then the following relations must hold:

$$a + 200 = b.$$

Here we can calculate all flows on this spanning tree and the condition again is the flow balance condition.

- (c) The optimality condition for the spanning tree solution is that all reduced costs of non-basic arcs are non-negative. In order to calculate the reduced costs, we need to calculate node potentials. Without loss of generality, set  $p_C = 0$ . Knowing the fact that  $\bar{c}_{ij} = c_{ij} - (p_i - p_j) = 0$  for all basic arcs  $(i, j)$ , we obtain  $p_B = 5$ ,  $p_D = -3$ ,  $p_A = -1$ , and  $p_E = -2$ . The reduced costs for non-basic arcs then can be calculated as follows:

$$\bar{c}_{BA} = c - 6, \bar{c}_{CA} = d - 1, \bar{c}_{BE} = 2, \bar{c}_{ED} = 10$$

Thus in addition to the feasibility condition  $b = a + 200$ , we obtain the optimality condition for the indicated spanning tree solution:

$$\begin{cases} c - 6 \geq 0 \\ d - 1 \geq 0 \end{cases} .$$

- (d) We need to have reduced costs of some non-basic arcs to be zero in order to have multiple optimal solutions, including this spanning tree solution. It means either  $c = 6$  or  $d = 1$  (in addition to the feasibility and optimality conditions mentioned previously in (c)). We can see that there is a cost-equivalent path from  $B$  to  $D$  via  $A$  as compared to the path  $B \rightarrow C \rightarrow D$  in both cases. Thus, other optimal solutions can be constructed by rerouting flows from  $B$  to  $D$  via  $A$  if either  $c = 6$  or  $d = 1$ . The final conditions for multiple optimal solutions are:  $a + 200 = b$ , and
 
$$\begin{cases} c - 6 = 0 \\ d - 1 \geq 0 \end{cases} \text{ or } \begin{cases} c - 6 \geq 0 \\ d - 1 = 0 \end{cases} .$$

**P4.** [25 pts] Consider a set of  $n$  points  $\{(x_1, y_1), \dots, (x_n, y_n)\}$  in the plane. We want to find a point  $(x, y)$  such that the sum of the Euclidean distances from this point to all the other points is minimized.

- (a) Give a nonlinear optimization formulation of this problem.
- (b) Is the objective function differentiable? Is this a convex optimization problem?
- (c) Write the corresponding optimality conditions. Give a geometric interpretation of this condition.
- (d) Are the optimality conditions necessary? Sufficient? State clearly your assumptions.
- (e) Provide a semidefinite programming formulation of this problem.

All answers and explanations must be *fully* justified.

**Solution:** (a) A simple formulation as an unconstrained nonlinear optimization problem is the following:

$$\min_{x,y} \sum_{i=1}^n \sqrt{(x-x_i)^2 + (y-y_i)^2}.$$

- (b) The objective function is differentiable everywhere, except at the points where  $(x, y)$  is equal to one of the  $(x_i, y_i)$ . The objective function is convex, since it is a sum of  $n$  convex functions.
- (c) The optimality conditions are obtained by setting the gradient equal to zero (we assume that the minimum occurs at a differentiable point).

$$\frac{\partial f}{\partial x} = \sum_{i=1}^n \frac{x-x_i}{\|(x,y)-(x_i,y_i)\|} = 0$$

$$\frac{\partial f}{\partial y} = \sum_{i=1}^n \frac{y-y_i}{\|(x,y)-(x_i,y_i)\|} = 0$$

This condition can be interpreted as requiring the sum of the normalized vectors from the point  $(x, y)$  to the  $(x_i, y_i)$  to be equal to zero.

For instance, in the case  $n = 2$ , then any point in the line segment between the two given points will be optimal. Similarly, for  $n = 3$ , the optimality condition implies that the angles between the vectors from  $(x, y)$  to the other points are all equal.

- (d) The optimality conditions are sufficient, because the problem is convex. They are necessary if the minimum occurs at a differentiable point.
- (e) A semidefinite formulation of the problem can be easily obtained if we introduce slack variables  $d_i$  and formulate the problem as:

$$\min \sum_{i=1}^n d_i \quad \text{s.t.} \quad (x-x_i)^2 + (y-y_i)^2 \leq d_i^2 \quad i = 1, \dots, n.$$

The constraints are equivalent to:

$$d_i^2 - \begin{bmatrix} x-x_i \\ y-y_i \end{bmatrix}^T \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x-x_i \\ y-y_i \end{bmatrix} \geq 0$$

Dividing by  $d_i$  and using Schur complements, we can rewrite this as:

$$\min \sum_{i=1}^n d_i \quad \text{s.t.} \quad \begin{bmatrix} d_i & x-x_i & y-y_i \\ x-x_i & d_i & 0 \\ y-y_i & 0 & d_i \end{bmatrix} \succeq 0, \quad i = 1, \dots, n.$$

which is an SDP formulation.

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15.093J / 6.255J Optimization Methods  
Fall 2009

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