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 56:271 Nonlinear Programming  
 Final Exam (Take-home) - Thursday, May 15, 1997  
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## Section A

Indicate whether each statement is *true* (+) or *false* (o). If false, briefly state why or give a counterexample.

- \_\_\_\_\_ 1. A barrier function adds a penalty to the objective when the solution approaches a boundary of the feasible region.
- \_\_\_\_\_ 2. The number of "dependent" variables of the GRG algorithm is equal to the number of equality constraints.
- \_\_\_\_\_ 3. The cubic interpolation procedure for one-dimensional minimization requires the computation of second derivatives.
- \_\_\_\_\_ 4. The function  $x^a$  is convex for  $x > 0$ .
- \_\_\_\_\_ 5. The feasible directions algorithm computes a search direction by solving an LP.
- \_\_\_\_\_ 6. The Fibonacci method for one-dimensional minimization requires the computation of derivatives.
- \_\_\_\_\_ 7. If all posynomials in a (posynomial) GP problem are condensed into single terms, then it is always possible to rewrite the resulting problem as an LP problem.
- \_\_\_\_\_ 8. A linear function is neither convex nor concave.
- \_\_\_\_\_ 9. The Gradient Projection method always follows the boundary of the feasible region, and never enters the interior.
- \_\_\_\_\_ 10. The logarithm of the dual GP objective  $v(\cdot, \cdot)$  above is a concave function of  $\cdot$  and  $\cdot$ .
- \_\_\_\_\_ 11. The logarithm of the dual GP objective  $v(\cdot, \cdot)$  above is not a concave function of both  $\cdot$  and  $\cdot$ , but becomes so if you eliminate  $k$  using  $k = \frac{\cdot}{i [k]}$ .
- \_\_\_\_\_ 12. A locally optimal solution for a signomial geometric programming problem must be globally optimal.
- \_\_\_\_\_ 13. The dual ("pseudo-dual") of a signomial GP problem may have multiple local maxima, but its global maximum will have the same objective value as the global minimum of the primal signomial GP.
- \_\_\_\_\_ 14. Penalty functions may be used for both equality and inequality constraints, but barrier functions may be used only for inequality constraints.

- \_\_\_\_\_ 16. If the Lagrangian function  $L(x, \lambda)$  has a saddlepoint  $(x^0, \lambda^0)$ , then there is zero duality gap between primal and dual optimal solutions.
- \_\_\_\_\_ 17. Consider the problem:  $\min f(x)$  subject to  $g_i(x) \leq 0, i=1,2,\dots,m, x \geq 0$  (where  $g_i$  satisfy some constraint qualification) and its Lagrangian function
- $$L(x, \lambda) = f(x) + \sum_{i=1}^m \lambda_i g_i(x)$$
- Then if  $f$  and  $g_i$  are convex, the partial derivative of  $L$  with respect to each  $x_j$  must be zero at the optimum.
- \_\_\_\_\_ 18. For the problem in the previous statement,  $\lambda_i g_i(x)$  must be zero at the optimum.
- \_\_\_\_\_ 19. For the problem in the previous statement, either  $x_j$  or  $\frac{\partial L(x, \lambda)}{\partial x_j}$  (or possibly both) must be zero at the optimum.
- \_\_\_\_\_ 20. The limit, as  $w \rightarrow 0$ , of the function  $(c/w)^w$  is \_\_\_\_\_.
- \_\_\_\_\_ 21.  $\lim_{x \rightarrow 0} \ln x = -\infty$ , but  $\lim_{x \rightarrow 0} x \ln x = 0$
- \_\_\_\_\_ 22. If we define the function  $f(x) = x \ln x$ , then  $\lim_{x \rightarrow 0} \frac{d}{dx} f(x) = -$  \_\_\_\_\_.
- \_\_\_\_\_ 23. When the dual problem is solved, the primal variables of a posynomial GP problem are often computed by exponentiating the Lagrange multipliers of the orthogonality constraints.
- \_\_\_\_\_ 24. The function  $f(x,y) = xy$  is a concave function of  $x$  and  $y$ .
- \_\_\_\_\_ 25. If an algorithm is applied to a minimization problem with optimal value zero, and the objective value is approximately halved at each iteration, then we would say that the algorithm has a linear rate of convergence.
- \_\_\_\_\_ 26. The Hessian matrix of  $(x^t Q x + c x)$  is  $2Q$ .
- \_\_\_\_\_ 27. In the QC/LD problem, the objective function is assumed to be quadratic.
- \_\_\_\_\_ 28. If  $f_i(x)$  is linear for  $i=1,2,\dots,n$ , then the function  $F(x)$  defined by  $F(x) = \max_{i=1,2,\dots,n} \{f_i(x)\}$  is a convex function.
- \_\_\_\_\_ 29. If a posynomial GP objective function continues to decrease as  $x_j \rightarrow 0$  for some primal variable  $x_j$ , then the dual problem objective is unbounded.
- \_\_\_\_\_ 30. If a posynomial GP objective function continues to decrease as  $x_j \rightarrow +\infty$  for some primal variable  $x_j$ , then the dual problem is infeasible.
- \_\_\_\_\_ 31. Every posynomial function is convex.
- \_\_\_\_\_ 32. The dual of a constrained posynomial GP problem has only linear equality and nonnegativity constraints.

- \_\_\_\_\_ 33. If a primal GP constraint is slack, then all the weights of the terms in that constraint must be zero.
- \_\_\_\_\_ 34. In the QC/LD problem, the variables are restricted to be integer-valued.
- \_\_\_\_\_ 35. The gradient of the function  $f(x) = (x^t Q x + cx)$  is  $Qx + c$ .
- \_\_\_\_\_ 36. Solving the QC/LD problem requires a one-dimensional search at each iteration.
- \_\_\_\_\_ 37. The posynomial constraint  $g_i(x) \leq 1$  has a convex feasible region.
- \_\_\_\_\_ 38. It is always possible (e.g., by a change of variable) to reformulate a posynomial GP problem so as to have a convex objective and convex feasible region.
- \_\_\_\_\_ 39. The objective of the posynomial GP problem, i.e.,
- $$v(x) = \prod_i \left( \frac{c_i}{x_i} \right)^{a_i} \prod_k x_k^{b_k}$$
- is a concave function of  $x$  and  $y$ .
- \_\_\_\_\_ 40. In the "Golden Section Search" method, more than one-third of the interval of uncertainty is eliminated at each iteration (assuming that no function values are equal).
- \_\_\_\_\_ 41. If all posynomials in a (posynomial) GP problem are condensed into single terms, then the resulting problem has a feasible region which includes the original feasible region.
- \_\_\_\_\_ 42. The Hessian matrix of a quadratic function is constant.
- \_\_\_\_\_ 43. In Wolfe's complementary pivoting algorithm for QP, if a single artificial variable is used, then the tableau does not require an objective row.
- \_\_\_\_\_ 44. The gradient projection method computes Lagrange multipliers at each iteration, and stops when they have the appropriate sign.
- \_\_\_\_\_ 45. "Quasi-Newton" search methods for unconstrained minimization require the computation of second partial derivatives.
- \_\_\_\_\_ 46. Newton's method for unconstrained minimization requires the computation of second partial derivatives.
- \_\_\_\_\_ 47. If you guess at the value of some primal GP variable and then fix it at this value, the dual GP problem becomes more difficult to solve.
- \_\_\_\_\_ 48. In a "generalized linear programming" problem, the column of coefficients of the variables must be selected as well as the variables.
- \_\_\_\_\_ 49. The function  $e^{ax}$  is convex *only* for a  $\leq 0$ .
- \_\_\_\_\_ 50. The Davidon-Fletcher-Powell (DFP) algorithm requires the computation of partial derivatives.
- \_\_\_\_\_ 51. If a posynomial function is "condensed" into a monomial, the monomial function (if not equal) is an overestimate of the posynomial function.

- \_\_\_\_\_ 52. Wolfe's method for quadratic programming requires a one-dimensional search at every iteration.
- \_\_\_\_\_ 53. If a constrained nonlinear programming problem satisfies a "constraint qualification", then a point which satisfies the Karush-Kuhn-Tucker conditions must be an optimal solution.
- \_\_\_\_\_ 54. A barrier function allows a constraint to be violated, but adds a penalty if the constraint is violated.
- \_\_\_\_\_ 55. The GRG algorithm requires the use of a one-dimensional search method.
- \_\_\_\_\_ 56. The Feasible Directions algorithm requires the use of a one-dimensional search method.
- \_\_\_\_\_ 57. If a constrained nonlinear programming problem satisfies a "constraint qualification", then the Karush-Kuhn-Tucker conditions must be satisfied by an optimal solution.
- \_\_\_\_\_ 58. The function  $e^{ax}$  is convex for *all* values of  $a$ .
- \_\_\_\_\_ 59. In Wolfe's complementary pivoting method for quadratic programming, the complementary slackness conditions are satisfied after each iteration.
- \_\_\_\_\_ 60. The tableau for Wolfe's method for quadratic programming includes columns for both primal and dual variables.
- \_\_\_\_\_ 61. The function  $x^a$  is convex for  $a > 0$ .
- \_\_\_\_\_ 62. If a nonlinear programming problem has only linear constraints, then any point which satisfies the Karush-Kuhn-Tucker conditions must be optimal.
- \_\_\_\_\_ 63. The Fletcher-Reeves method is also known as a "Quasi-Newton" method.
- \_\_\_\_\_ 64. The quadratic interpolation procedure for one-dimensional minimization requires the computation of derivatives.
- \_\_\_\_\_ 65. If the GRG algorithm were applied to a LP problem, it would produce, at each iteration, the same solution as the simplex algorithm for LP.
- \_\_\_\_\_ 66. A penalty function allows a constraint to be violated, but adds a penalty if the constraint is violated.
- \_\_\_\_\_ 67. The Lagrangian dual of a convex quadratic programming problem is a quadratic programming problem with only nonnegativity constraints on the dual variables.
- \_\_\_\_\_ 68. Powell's algorithm for unconstrained minimization requires the computation of partial derivatives.
- \_\_\_\_\_ 69. The Davidon-Fletcher-Powell method is also known as the "Conjugate Gradient" method.
- \_\_\_\_\_ 70. In the QC/LD problem, the objective function is assumed to be convex.

## Section B

Below is a list of references concerning Geometric Programming and applications. Part I lists papers which explicitly use GP, while Part II lists papers whose applications appear to be good candidates for GP.

### Part I: Selected References to Geometric Programming Applications

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- Paul, H. (1982). "An Application of Geometric Programming to Heat Exchanger Design." Computers and Industrial Engineering **6**(2): 103-114.
- Phillips, D. T. and C. S. Beightler (1970). "Optimization in Tool Engineering Using Geometric Programming." AIIE Transactions **2**(4): 355-360.
- Corstjens, M. and P. Doyle (1981). "A Model for Optimizing Retail Space Allocations." Management Science **27**(7): 822-833.
- Balachandran, V. and D. Gensch (1973). Solving the "Marketing Mix" Problem Using Geometric Programming. Northwestern University.
- Dajani, J. S., Y. Hasit, et al. (1977). "Geometric Programming in Sewer Network Design." Engineering Optimization **3**: 27-35.
- Edwards, L. S. (1975). "Optimum Limit State Design of Highway Bridge Superstructures Using Geometric Programming." Engineering Optimization **1**: 201-212.
- Unklesbay, K. and D. L. Creighton (1978). "The Optimization of Multi-pass Machining Process." Engineering Optimization **3**: 229-238.
- Mine, H. and K. Ohno (1970). "Decomposition of Mathematical Programming Problems by Dynamic Programming and its Application to BLock-Diagonal Geometric Programs." J. Jath. Anal. Appl. **32**: 370ff.
- Yu, C. H., N. C. Dasgupta, et al. (1986). "Optimization of Prestressed Concrete Bridge Girders." Engineering Optimization **10**(1): 13-24.
- Kapur, K. C. (1978). Optimization in Probabilistic Design for Engineering Systems. Second International Symposium on Large Engineering Systems, Waterloo, Sandford Educational Press.
- Wyman, F. P. (1978). "Use of Geometric Programming in the Design of an Algerian Water Conveyance System." Interfaces **8**(3): 1-6.
- Philipson, R. H. and A. Ravindran (1979). "Application of Mathematical Programming to Metal Cutting." Mathematical Programming Study **11**: 116-134.

- Hivner, W. H. and R. P. Mehta (1977). "Optimizing Computer Performance with Geometric Programming." European Journal of Operational Research **1**(2):
- Grange, F. E., G. A. Kochenberger, et al. (1992). Optimal Design of Multi-Pass Heat Exchangers with Geometric Programming.
- Woolsey, R. E. D. An Analysis of a Model of the MPS MX Missile System Using Geometric Programming. Mathematics Dept., Colorado School of Mines.
- Gopalakrishnan, B. and F. Al-Khayyal (1991). "Machine Parameter Selection for Turning based on Geometric Programming." International Journal of Production Research **29**: 1897-1908.
- Smeers, Y. and D. Tyteca (1984). "A Geometric Programming Model for the Optimal Design of Wastewater Treatment Plants." Operations Research **32**(2): 314-342.

## Questions:

- Chose one reference from Part I of the list of GP applications.
- Briefly summarize the application problem.
- Write the mathematical model.
- Identify the # of (primal) variables, # of constraints, total # of terms.
- State whether posynomial or signomial in nature.
- Describe the method used by the author(s) of the paper.
- If sample data is given for a sample problem, or if you can guess at reasonable data, solve the GP problem.

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## Part II: Selected References to Potential Geometric Programming Applications

- Dinkel, J. J. and G. A. Kochenberger (1974). "On "A Cofferdam Design Optimization"." Mathematical Programming **6**(1): 114-116.
- Neghabat, F. and R. M. Stark (1972). "A Cofferdam Design Optimization"." Mathematical Programming **3**: 263-275.
- Feiring, B. R. (1990). "An Efficient Procedure for the N-city Traveling Salesman Problem." Mathematical and Computer Modelling **13**(3): 95-98.
- Terry, W. R. and K. W. Cutright (1986). Computer Aided Design of a Broaching Process. Computers and Industrial Engineering. **11**: 576-580.
- Ulusoy, A. G. and D. M. Miller (1979). Optimal Design of Pipeline Networks Carrying Homogeneous Coal Slurry. Mathematical Programming Study. North-Holland Publishing Company. 85-107.

Cowton, C. J. and A. Wirth (1993). "On the Economics of Cutting Tools." International Journal of Production Research **31**(10): 2441-2446.

Quesada, I. and I. E. Grossmann (1992). A Global Optimization Algorithm for Heat Exchanger Networks. Engineering Design Research Center, Carnegie-Mellon University.

## Questions:

- a. Chose one reference from Part II of the list of potential GP applications.
- b. Briefly summarize the application problem.
- c. Write the mathematical model given by the author(s).
- d. If possible, reformulate the model as a GP (either posynomial or signomial) problem.
- e. Identify the # of (primal) variables, # of constraints, total # of terms.
- f. Describe the method used by the author(s) of the paper.
- g. If sample data is given for a sample problem, or if you can guess at reasonable data, solve the GP problem.