# **gpposy** A Matlab Solver for Geometric Programs in Posynomial Form

Kwangmoo Koh deneb1@stanford.edu Seungjean Kim sjkim@stanford.edu

Almir Mutapcic almirm@stanford.edu Stephen Boyd boyd@stanford.edu

May 22, 2006

### 1 Introduction

gpposy solves an optimization problem of the form

$$\begin{array}{ll} \text{minimize} & \sum_{k=1}^{K_0} b_k^{(0)} x_1^{a_{k1}^{(0)}} x_2^{a_{k2}^{(0)}} \cdots x_n^{a_{kn}^{(0)}} \\ \text{subject to} & \sum_{k=1}^{K_i} b_k^{(i)} x_1^{a_{k1}^{(i)}} x_2^{a_{k2}^{(i)}} \cdots x_n^{a_{kn}^{(i)}} \le 1, \quad i = 1, \dots, m, \\ & h_i x_1^{g_{i1}} x_2^{g_{i2}} \cdots x_n^{g_{in}} = 1, \quad i = 1, \dots, p, \\ & l \le x \le u, \end{array}$$
(1)

where the optimization variable is the vector  $x = (x_1, \ldots, x_n) \in \mathbf{R}_+^n$ . The problem data are  $a_{kj}^{(i)}, g_{ij} \in \mathbf{R}, b_k^{(i)}, h_i \in \mathbf{R}_+$ , and  $l, u \in \mathbf{R}_+^n$ . Here  $\preceq$  means componentwise inequality between vectors. This problem is called a *geometric program in posynomial form*. For more information about geometric programming, see [BV04, BKVH].

### 2 Calling sequences

The complete calling sequence of gpposy is

>> [x,status,lambda,nu] = gpposy(A,b,szs,G,h,l,u,quiet);

Input arguments represent the problem data of the problem (1). Output arguments are the optimal point (if feasible), sensitivity information (if feasible) and the solution status.

#### Input arguments

• A: matrix with n columns and  $\sum_{i=0}^{m} K_i$  rows that specifies the exponents of the objective and equality constraints, *i.e.*,

$$A = \begin{bmatrix} A^{(0)} \\ \vdots \\ A^{(m)} \end{bmatrix}, \quad A^{(i)} = \begin{bmatrix} a_{11}^{(i)} & \cdots & a_{1n}^{(i)} \\ \vdots & \ddots & \vdots \\ a_{K_i1}^{(i)} & \cdots & a_{K_in}^{(i)} \end{bmatrix} \in \mathbf{R}^{K_i \times n}, \quad i = 0, \dots, m.$$

A can be in sparse format.

• b: vector of length  $\sum_{i=0}^{m} K_i$  that specifies the coefficients of the objective and inequality constraints, *i.e.*,

$$b = \begin{bmatrix} b^{(0)} \\ \vdots \\ b^{(m)} \end{bmatrix}, \quad b^{(i)} = \begin{bmatrix} b_1^{(i)} \\ \vdots \\ b_{K_i}^{(i)} \end{bmatrix} \in \mathbf{R}_+^{K_i}, \quad i = 0, \dots, m.$$

All elements  $b_k^{(i)}$  must be positive.

- szs: vector of length m + 1 that specifies the number of terms in each objective and inequality constraints, *i.e.*,  $(K_0, \ldots, K_m)$ .
- G: matrix with n columns and p rows, that specifies the exponents of equality constraints, *i.e.*,

$$G = \begin{bmatrix} g_{11} & \cdots & g_{1n} \\ \vdots & \ddots & \vdots \\ g_{p1} & \cdots & g_{pn} \end{bmatrix} \in \mathbf{R}^{p \times n}.$$

G can be in sparse format.

• h: p-vector that contains the coefficients of equality constraints, *i.e.*,

$$h = \begin{bmatrix} h_1 \\ \vdots \\ h_p \end{bmatrix} \in \mathbf{R}^p_+.$$

All elements  $h_k$  must be positive.

- 1: *n*-vector that specifies lower bounds on x. If not given, it will be set to the default lower bounds  $(10^{-100}, \ldots, 10^{-100})$ . All elements  $l_i$  must be positive.
- u: *n*-vector that specifies upper bounds on x. If not given, it will be set to the default upper bounds  $(10^{100}, \ldots, 10^{100})$ . All elements  $u_i$  must be positive.
- quiet: boolean. Suppresses print messages during execution if true. The default value is false.

#### **Output** arguments

- x: *n*-vector. x is the optimal point of the problem if the problem is feasible, and x is the last primal iterate of phase I if the problem is infeasible.
- status: string; possible values are 'Solved', 'Infeasible' and 'Failed'.
- lambda: vector of length m + 2n; the optimal sensitivity vector (see [BKVH, §3.3]) associated with inequality constraints if the problem is feasible. The first m elements, lambda(1:m), are optimal sensitivities of the m inequality constraints, the next n elements, lambda(m+1:m+n), are those of the lowerbound constraints ( $l \leq x$ ), and the last n elements, lambda(m+n:m+2\*n), are those of the upperbound constraints ( $x \leq u$ ). If the problem is feasible, lambda is a certificate of infeasibility (see [BKVH, §5.8.1,§11.4.3]).
- nu: *p*-vector; the optimal sensitivity vector (see [BKVH,  $\S3.3$ ]) associated with equality constraints (Gx + h = 0) if the problem is feasible. If infeasible, nu is a certificate of infeasibility (see [BKVH,  $\S5.8.1, \S11.4.3$ ]).

#### Other calling sequences

Other calling sequences supported by gpposy are:

```
>> [x,status,lambda,nu] = gpposy(A,b,szs);
>> [x,status,lambda,nu] = gpposy(A,b,szs,G,h);
>> [x,status,lambda,nu] = gpposy(A,b,szs,G,h,l,u);
>> [x,status,lambda,nu] = gpposy(A,b,szs,G,h,l,u,quiet);
>> [x,status,lambda,nu] = gpposy(A,b,szs,[],[],l,u);
>> [x,status,lambda,nu] = gpposy(A,b,szs,[],[],l,u,quiet);
>> [x,status,lambda,nu] = gpposy(A,b,szs,[],[],l,u,quiet);
```

### Caveats

- The equality constraint matrix, G, must be full rank.
- If your problem is large and sparse, be sure that A and G are in sparse format.
- Equality constraints should be explicitly specified as Gx + h = 0. You cannot represent an equality constraint as a pair of opposing inequality constraints.

## 3 Example

Consider the problem

minimize 
$$x_1^{-1}x_2^{-1/2}x_3^{-1} + 2.3x_1x_3 + 4x_1x_2x_3$$
  
subject to  $(1/3)x_1^{-2}x_2^{-2} + (4/3)x_2^{1/2}x_3^{-1} \le 1$ ,  
 $0.1x_1 + 0.2x_2 + 0.3x_3 \le 1$ ,  
 $(1/2)x_1x_2 = 1$ ,

with variables  $x_1$ ,  $x_2$  and  $x_3$ . This problem has the form (1) with

$$A^{(0)} = \begin{bmatrix} -1 & -0.5 & -1 \\ 1 & 0 & 1 \\ 1 & 1 & 1 \end{bmatrix}, \quad A^{(1)} = \begin{bmatrix} -2 & -2 & 0 \\ 0 & 0.5 & -1 \end{bmatrix}, \quad A^{(2)} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix},$$
$$b^{(0)} = \begin{bmatrix} 1 \\ 2.3 \\ 4 \end{bmatrix}, \quad b^{(1)} = \begin{bmatrix} 1/3 \\ 4/3 \end{bmatrix}, \quad b^{(2)} = \begin{bmatrix} 0.1 \\ 0.2 \\ 0.3 \end{bmatrix},$$
$$G = \begin{bmatrix} 1 & 1 & 0 \end{bmatrix}, \quad h = 0.5.$$

The Matlab code for solving this problem is as follows:

% Matlab script that solves the above problem

```
>> A0 = [ -1 -0.5
                      -1 ;...
                      1 ;...
             1
                  0
                  1
                      1];
             1
>> A1 = [ -2
               -2
                      0 ;...
             0 0.5
                      -1];
>> A2 = [
             1
                 0
                      0 ;...
                  1
             0
                      0 ;...
             0
                 0
                      1];
               A1;
>> A
      = [ AO;
                     A2 ];
>> b0
            1; 2.3;
      = [
                       4];
>> b1 = [ 1/3; 4/3 ];
     = [ 0.1; 0.2; 0.3 ];
>> b2
      = [ b0; b1; b2 ];
>> b
>> G
      = [
             1
                  1
                      0];
      = [ 0.5 ];
>> h
>> szs = [ size(A0,1); size(A1,1); size(A2,1) ]; %i.e., [ 3; 2; 3 ]
>> [x,status,lambda,nu] = gpposy(A,b,szs,G,h);
```

After executing the code, you can see the result by typing x in Matlab.

>> x ans = 3.4783 0.5750 1.1030

## References

- [BKVH] S. Boyd, S.-J. Kim, L. Vandenberghe, and A. Hassibi. A tutorial on geometric programming. To appear in *Optimization and Engineering*, 2005. Available at www.stanford.edu/~boyd/gp\_tutorial.html.
- [BV04] S. Boyd and L. Vandenberghe. *Convex Optimization*. Cambridge University Press, 2004. Available at www.stanford.edu/~boyd/cvxbook.html.