

Assignment - Constrained optimization

Interior Point Method applied to linear programming problems

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1 Problem definition

The Predictor-Corrector is an Interior Point Method built to iteratively solve a nonlinear system of equations, obtained by defining the KKT condition for the pair primal-dual problem.

The linear system to solve is:

$$F(x, \lambda, s) = \begin{bmatrix} Ax - b \\ s + A^T \lambda - c \\ XSe \end{bmatrix} = 0$$

According to Newton method for nonlinear system of equation, the system can locally linearized and solved. To linearize this system is necessary to compute the Jacobian of F at each iteration and solve the linear system:

$$F'(x_k, \lambda_k, s_k) \cdot \begin{pmatrix} \Delta x_k \\ \Delta \lambda_k \\ \Delta s_k \end{pmatrix} + F(x_k, \lambda_k, s_k) = 0$$
$$\begin{pmatrix} A & 0 & 0 \\ 0 & A^T & I \\ S_k & 0 & X_k \end{pmatrix} \cdot \begin{pmatrix} \Delta x_k \\ \Delta \lambda_k \\ \Delta s_k \end{pmatrix} + F(x_k, \lambda_k, s_k) = 0$$

Where $X_k = \text{diag}(x_k)$ and $S_k = \text{diag}(s_k)$.

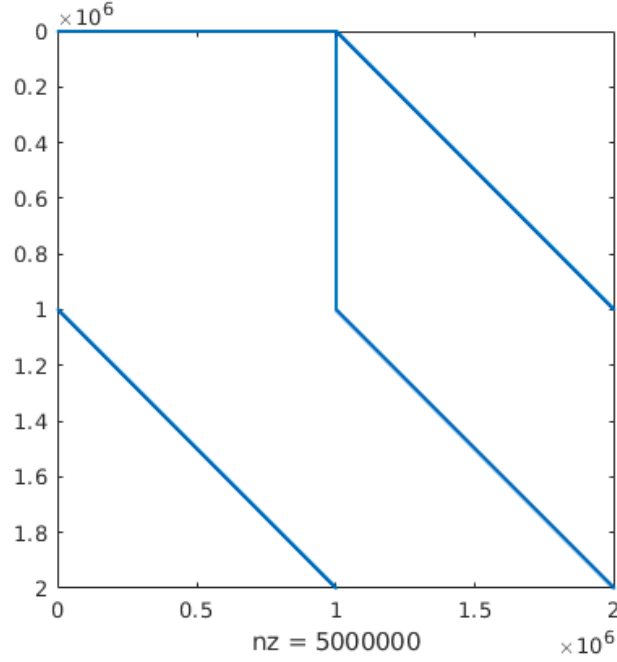
Interior point methods are variants of Newton method that guarantee that each iterate (x_k, λ_k, s_k) satisfy the constraint $x, s \geq 0$.

2 Problem data

$A = (1, \dots, 1) \in \mathbb{R}^{1 \times n}$,
 $b = 1$,

$c \in \mathbb{R}^n : c_i = a$ if i is odd, otherwise 1.
 $x \in \mathbb{R}^n$.

After having built the sparse jacobian of F, using the command `spy()` it's possible to inspect its structure:



Using the command `[R,p] = chol(J)` it's possible to notice, thanks to the value of `p`, that the matrix isn't positive definite. Hence I could not use: gradient method, conjugate gradient or Cholesky decomposition. Since this matrix isn't symmetric, I didn't use the LDL decomposition neither.

Analyzing the structure of the block matrix, I could notice that it's possible to make some reductions, rewriting the left hand part in a more compact way:

$$\begin{pmatrix} A & 0 & 0 \\ 0 & A^T & I \\ S_k & 0 & X_k \end{pmatrix} \cdot \begin{pmatrix} \Delta x_k \\ \Delta \lambda_k \\ \Delta s_k \end{pmatrix} = \begin{pmatrix} r_a \\ r_b \\ r_c \end{pmatrix}$$

Since X is invertible ($x > 0$ is guaranteed by the IPM):

$$\begin{pmatrix} A & 0 \\ -X^{-1}S & A^T \end{pmatrix} \cdot \begin{pmatrix} \Delta x_k \\ \Delta \lambda_k \end{pmatrix} = \begin{pmatrix} r_a \\ r_b - X^{-1}r_c \end{pmatrix}$$

S is invertible for the same reason:

$$\begin{aligned} AS^{-1}XA^T\Delta\lambda &= r_a + AXS^{-1}r_b - AS^{-1}r_c \\ \Delta s &= r_b - A^T\Delta\lambda \\ \Delta x &= S^{-1}r_c - XS^{-1}\Delta s \end{aligned}$$

The matrix $AS^{-1}XA^T \in \Re^{m \times m}$ is a real number, since in this case $m = 1$. This makes the computation of the system trivial, avoiding the problem of working with singular or indefinite matrices.

3 Results

n iter		a			
		2	20	200	2000
n	1,00E+04	7	9	9	10
	1,00E+06	8	10	9	11

time		a			
		2	20	200	2000
n	1,00E+04	0,1542	0,1038	0,0988	0,1635
	1,00E+06	9,1605	11,1476	10,1838	12,089

Figure 1: Starting point of all ones.

n iter		a			
		2	20	200	2000
n	1,00E+04	9	10	10	11
	1,00E+06	9	10	10	12

time		a			
		2	20	200	2000
n	1,00E+04	0,166	0,1329	0,1328	0,1449
	1,00E+06	10,2799	11,78	13,3218	13,8344

Figure 2: Starting point of all 100.

n iter		a			
		2	20	200	2000
n	1,00E+04	4	4	4	4
	1,00E+06	4	4	4	4

Time		a			
		2	20	200	2000
n	1,00E+04	0,1044	0,0677	0,0662	0,0657
	1,00E+06	5,2523	5,274	5,193	5,2321

Figure 3: Starting point as suggested on the book.