

CS/SE4-6TE3, CES 722/723: Reduced Gradient

Consider the following linearly constrained non-linear optimization (NLO) problem:

$$\begin{aligned} \text{(NLO)} \quad & \min \quad x_1 + x_2 + x_1x_2 \\ & \text{s.t.} \quad 3x_1 + 2x_2 = 5. \end{aligned}$$

Reduce this problem to an unconstrained optimization problem.

Solution: Substituting $x_2 = (5 - 3x_1)/2$ into the objective function, we get the reduced unconstrained problem

$$\min \quad f_N(x_1) = \frac{5}{2} + 2x_1 - \frac{3}{2}x_1^2.$$

Note that we get the same result choosing x_2 as the basic variable (in that case $B = 2$ and $N = 3$) and eliminating it, that ends up in

$$f_N(x_N) = f_N(x_1) = f(x_N, B^{-1}b - B^{-1}Nx_N) = f\left(x_1, \frac{5}{2} - \frac{3}{2}x_1\right) = \frac{5}{2} + 2x_1 - \frac{3}{2}x_1^2.$$

Consider the following non-linear optimization (NLO) problem:

$$\begin{aligned} \min \quad & (x_1 - x_2)^2 + (\exp(-x_1 + x_2 - x_3))^2 \\ \text{s.t.} \quad & 10x_1 + 5x_2 + 15x_3 = 60 \\ & 2x_1 + 3x_3 = 10 \end{aligned} \tag{1}$$

1. Reformulate this linearly constrained NLO problem as a unconstrained optimization problem.

Solution: Rewrite the constraints into $Ax = b$.

$$\begin{pmatrix} 10 & 5 & 15 \\ 2 & 0 & 3 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 60 \\ 10 \end{pmatrix} \tag{2}$$

So

$$A = \begin{pmatrix} 10 & 5 & 15 \\ 2 & 0 & 3 \end{pmatrix}$$

and

$$b = \begin{pmatrix} 60 \\ 10 \end{pmatrix}.$$

Given the basis

$$B = \begin{pmatrix} 10 & 5 \\ 2 & 0 \end{pmatrix},$$

thus

$$N = \begin{pmatrix} 15 \\ 3 \end{pmatrix},$$

$$x_B = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

and

$$x_N = (x_3).$$

$$\begin{aligned} Ax &= Bx_B + Nx_N = b \\ \Rightarrow x_B &= B^{-1}b - B^{-1}Nx_N \\ &= \begin{pmatrix} 10 & 5 \\ 2 & 0 \end{pmatrix}^{-1} \begin{pmatrix} 60 \\ 10 \end{pmatrix} - \begin{pmatrix} 10 & 5 \\ 2 & 0 \end{pmatrix}^{-1} \begin{pmatrix} 15 \\ 3 \end{pmatrix} (x_3) \\ &= \begin{pmatrix} 0 & 0.5 \\ 0.2 & -1 \end{pmatrix} \begin{pmatrix} 60 \\ 10 \end{pmatrix} - \begin{pmatrix} 0 & 0.5 \\ 0.2 & -1 \end{pmatrix} \begin{pmatrix} 15 \\ 3 \end{pmatrix} (x_3) \\ &= \begin{pmatrix} 5 \\ 2 \end{pmatrix} - \begin{pmatrix} 1.5 \\ 0 \end{pmatrix} (x_3) \\ &= \begin{pmatrix} 5 - 1.5x_3 \\ 2 \end{pmatrix} \end{aligned}$$

So we can change the objective function as the following:

$$\begin{aligned} f_N(x_N) &= f(x) = f(5 - 1.5x_3, 2, x_3) \\ &= (3 - 1.5x_3)^2 + \exp(0.5x_3 - 3)^2 \\ &= (3 - 1.5x_N)^2 + \exp(0.5x_N - 3)^2 \end{aligned}$$

Therefore, we reformulate the NLO problem into an unconstrained problem as

$$\min f_N(x_N),$$

where

$$f_N(x_N) = ((3 - 1.5x_N)^2 + \exp(0.5x_N - 3)^2).$$

2. Give out the null-space of A .

Solution: The null-space of A is

$$Z = \begin{pmatrix} -B^{-1}N \\ I \end{pmatrix} = \begin{pmatrix} -1.5 \\ 0 \\ 1 \end{pmatrix}$$

Clearly, $AZ = 0$.

3. Let now assume that $x \geq 0$ is also a required constraint. Apply one step of the Reduced Gradient method for this modified problem from the point $x = (5, 2, 0)^T$ with the initial basis given by x_1 and x_2 .

Solution: From the previous solution, we reformulated the problem as:

$$\begin{aligned} \min & (3 - 1.5x_N)^2 + \exp(0.5x_N - 3)^2 \\ \text{s.t.} & \quad 5 - 1.5x_N \geq 0 \\ & \quad x_N \geq 0 \end{aligned}$$

And we also get the reduced gradient at the point $(5, 2, 0)^T$ with the basis given by x_1 and x_2 :

$$\begin{aligned}
\nabla f_N(x_N)^T &= (-3(3 - 1.5x_N) + \exp(0.5x_N - 3)^2) = (-8.9975) \\
s_N &= -\nabla f_N(x_N)^T = 8.9975 \\
s_B &= -B^{-1}N s_N = -\begin{pmatrix} 1.5 \\ 0 \end{pmatrix} * (8.9975) = \begin{pmatrix} -13.4963 \\ 0 \end{pmatrix} \\
(s^1)^T &= (s_B^T, s_N^T) = (-13.4963 \quad 0 \quad 8.9975)^T \\
\bar{\lambda} &= \min_{1 \leq i \leq n, s_j < 0} \left\{ \frac{x_i}{-s_i} \right\} = 0.3705 \\
f(x^1 + \lambda s^1) &= f((5, 2, 0)^T + \lambda(-13.4963, 0, 8.9975)^T) \\
&= f(5 - 13.4963\lambda, 2, 8.9975\lambda) \\
&= (3 - 13.4963\lambda)^2 + \exp(4.4988\lambda - 3)^2
\end{aligned}$$

So we can make a line search with the Golden section:

$$\begin{aligned}
x^2 &= \arg \min_{0 \leq \lambda \leq \bar{\lambda}} f(x^1 + \lambda s^1) \\
&= \arg \min_{0 \leq \lambda \leq 0.3705} (3 - 13.4963\lambda)^2 + \exp(4.4988\lambda - 3)^2
\end{aligned}$$

We will get $\lambda = 0.2218$, then

$$x^2 = x^1 + \lambda s^1 = (2.0065, 2, 1.9957)^T,$$

and

$$f(x^2) = 0.0183 < f(x^1) = 9.0025.$$

4. Compute a projected gradient of $f(x)$ at the point $x = (5, 2, 0)^T$.

Solution:

$$\begin{aligned}
\nabla f(x) &= \begin{pmatrix} 2(x_1 - x_2) - 2 \exp(-2x_1 + 2x_2 - 2x_3) \\ -2(x_1 - x_2) + 2 \exp(-2x_1 + 2x_2 - 2x_3) \\ -2 \exp(-2x_1 + 2x_2 - 2x_3) \end{pmatrix} \\
\nabla f(5, 2, 0) &= \begin{pmatrix} 6 - 2e^{-6} \\ -6 + 2e^{-6} \\ -2e^{-6} \end{pmatrix} \\
Z^T \nabla f(5, 2, 0) &= (-1.5 \quad 0 \quad 1) \begin{pmatrix} 6 - 2e^{-6} \\ -6 + 2e^{-6} \\ -2e^{-6} \end{pmatrix} = (-9 + e^{-6}) = -8.9975
\end{aligned}$$