

# EN.550.661: Nonlinear Optimization I

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## Homework Assignment #4

*Starred exercises require the use of MATLAB.*

**Exercise 4.1:** Using the language of mathematics, describe how you would find the Cauchy point for a second-order model

$$m_k(s) = f_k + s^T g_k + \frac{1}{2} s^T B_k s$$

of  $f(x_k + s)$  within the trust-region  $\|s\| \leq \delta_k$ .

**Exercise 4.2:** Solve the trust-region subproblem

$$\underset{s \in \mathbb{R}^n}{\text{minimize}} \quad s^T g + \frac{1}{2} s^T B s \quad \text{subject to} \quad \|s\|_2 \leq \delta \quad (1)$$

in the following cases:

(a)

$$B = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{pmatrix}, \quad g = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}, \quad \text{and} \quad \delta = 2,$$

(b)

$$B = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{pmatrix}, \quad g = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}, \quad \text{and} \quad \delta = 5/12$$

Hint:  $\lambda = 2$  is a root of the nonlinear equation

$$\frac{1}{(1 + \lambda)^2} + \frac{1}{(2 + \lambda)^2} = \frac{25}{144}.$$

(c)

$$B = \begin{pmatrix} -2 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & -1 \end{pmatrix}, \quad g = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}, \quad \text{and} \quad \delta = 5/12,$$

(d)

$$B = \begin{pmatrix} -2 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & -1 \end{pmatrix}, \quad g = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}, \quad \text{and} \quad \delta = 1/2, \quad \text{and}$$

(e)

$$B = \begin{pmatrix} -2 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & -1 \end{pmatrix}, \quad g = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}, \quad \text{and} \quad \delta = \sqrt{2}.$$

**Exercise 4.3:** Sketch the solution of problem (1) with data

$$B = \begin{pmatrix} 1 & 0 \\ 0 & 3 \end{pmatrix} \quad \text{and} \quad g = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

as a function of the trust-region radius  $\delta$ , i.e., plot  $s(\lambda)$  as discussed in class. In which direction does the solution point as  $\delta$  shrinks to zero? How does the Lagrange multiplier for the trust-region constraint depend on the trust-region radius? For what value of the trust-region radius does the solution become unconstrained?

**Exercise 4.4\*:** Write a MATLAB m-function called `steihaug_CG.m` that implements the truncated linear conjugate gradient method by Steihaug based on Algorithm 4 in the Lecture 06 slides. The function call should have the form

$$[ \mathbf{p}, \text{iters}, \text{flag} ] = \text{steihaug\_CG}( \mathbf{B}, \mathbf{g}, \text{radius}, \text{tol} )$$

where the input  $\mathbf{B}$  is required to be a symmetric (possibly indefinite) matrix,  $\mathbf{g}$  is a vector, `radius`  $> 0$  is the trust-region radius, and `tol`  $\in (0, 1)$  is the stopping tolerance. On exit, the resulting approximate Steihaug truncated-CG solution should be stored in the vector  $\mathbf{p}$ , the number of iterations performed stored in `iters`, and the parameter `flag` should be set to one of the following values:  $-1$  if the algorithm terminated because of negative curvature,  $0$  if the stopping tolerance was met, and  $1$  if the algorithm returned a boundary solution simply because the CG iterations grew larger than the trust-region radius.

**Exercise 4.5\*:** Consider the problem

$$\underset{x \in \mathbb{R}^n}{\text{minimize}} \quad f(x)$$

where  $f$  is a twice continuously differentiable function.

- (a) Write a MATLAB m-function called `unc_TR.m` that implements a trust-region method with trial steps computed from the Steihaug truncated linear CG method as given in Problem 4.4 above. The function call should have the form

$$[\mathbf{x}, \mathbf{F}, \mathbf{G}, \mathbf{H}, \text{iter}, \text{status}] = \text{unc\_TR}(\text{fun}, \mathbf{x0}, \text{maxit}, \text{printlevel}, \text{tol})$$

where `fun` is of type *string* and represents the name of a Matlab m-function that computes  $f(x)$ ,  $\nabla f(x)$ , and  $\nabla^2 f(x)$  for some desired function  $f$ ; it should be of the form

$$[\mathbf{F}, \mathbf{G}, \mathbf{H}] = \text{fun}(\mathbf{x})$$

where for a given value  $x$  it returns the values of the function, gradient, and Hessian, respectively. The parameter `x0` is an initial guess at a minimizer of  $f$ , `maxit` is the maximum number of iterations allowed, `printlevel` determines the amount of printout required, and `tol` is the final stopping tolerance. In the code, if the parameter `printlevel` has the value zero, then no printing should occur; otherwise, a single line of output is printed (in column format) per iteration. On output, the parameters `x`, `F`, `G`, and `H` should contain the final iterate, function value, gradient vector, and Hessian matrix computed by the algorithm. The parameter `iter` should contain the total number of iterations performed. Finally, `status` should have the value  $0$  if the final stopping tolerance was obtained and the value  $1$  otherwise.

- (b) Write a separate MATLAB m-file with function declaration `[F,G,H] = fun(x)` that returns the value  $F$ , gradient  $G$ , and Hessian  $H$  at the point  $x \in \mathbb{R}^2$  of the function

$$f(x) = 10(x_2 - x_1^2)^2 + (x_1 - 1)^2.$$

Use your m-function `unc_TR.m` from part (a) to minimize  $f$  with starting point  $x_0 = (0, 0)$ .