

## Exercise #9

March 14, 2023

### Problem 1.

In this exercise, we study the Gauß–Newton method for solving the least-squares problem corresponding to the (overdetermined and inconsistent) system of equations

$$\begin{aligned}x + y &= 1, \\x - y &= 0, \\xy &= 2.\end{aligned}$$

To that end, we define

$$\begin{aligned}r_1(x, y) &:= x + y - 1, \\r_2(x, y) &:= x - y, \\r_3(x, y) &:= xy - 2,\end{aligned}$$

and

$$f(x, y) := \frac{1}{2} \sum_{j=1}^3 r_j(x, y)^2.$$

We denote moreover by  $J = J(x, y)$  the Jacobian of  $r$ .

- Show that the function  $f$  is non-convex, but that it has a unique minimiser  $(x^*, y^*)$ .
- Show that the matrix  $J^T J$  required in the Gauß–Newton method is positive definite for all  $x, y$ .
- Perform one step of the Gauß–Newton method (without line search) for the solution of this least-squares problem. Use the initial value  $(x_0, y_0) = (0, 0)$ .

### Problem 2.

Let

$$f(x) = x_1^4 + 2x_2^4 + x_1x_2 + x_1 - x_2 + 2.$$

Starting at the point  $x_0 = (0, 0)$  compute explicitly one step for the trust region method with the model function  $m(p) = f(x_0) + g^T p + \frac{1}{2} p^T B p$ , where  $g = \nabla f(x_0)$ ,  $B = \nabla^2 f(x_0)$ , and the trust region radius  $\Delta = 1$ .

**Problem 3.** (Problem 4.1 in N&W)

Let

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

 At  $x = (0, -1)$  draw the contour lines of the quadratic model

$$\min_p m(p) = f(x) + \langle \nabla f(x), p \rangle + \frac{1}{2} \langle p, Bp \rangle \quad \text{subject to } \|p\| \leq \Delta, \quad (1)$$

 assuming that  $B$  is the Hessian of  $f$ . Draw the family of solutions of (1) as the trust region radius varies from  $\Delta = 0$  to  $\Delta = 2$ . Repeat this at  $x = (0, 0.5)$ .

**Problem 4.** (Problem 4.5 in N&W)

 Let  $\phi(\tau) = m_k(\tau p_k^s)$ , where  $p_k^s = -\frac{\Delta_k}{\|g_k\|} g_k$  and  $m_k(p) = f_k + g_k^T p + \frac{1}{2} p^T B p$  with  $B \in \mathbb{R}^{d \times d}$  symmetric. Show that the minimizer of  $\phi(\tau)$  subject to  $\|\tau p_k^s\| \leq \Delta_k$  and  $\tau \geq 0$  is given as

$$\begin{cases} 1, & \text{if } g_k^T B_k g_k \leq 0, \\ \min\left(\frac{\|g_k\|^3}{\Delta_k g_k^T B_k g_k}, 1\right), & \text{otherwise.} \end{cases} \quad (2)$$