

## Assignment #8

**Due Wednesday 20 November 2024, at the start of class**

From the textbook<sup>1</sup> please read sections 11.3, 11.4, 11.5, 12.1, 12.2, and 12.3. Significant goals of this Assignment are to understand the benefits of a line search (section 11.5), and the motivation for the quasi-Newton approach (section 12.3). Regarding several of the problems, you will need the exact line search formula for quadratic functions; it is the conclusion of Exercise 5.3 on page 386.

DO THE FOLLOWING EXERCISES from section 11.4, pages 374–375:

- Exercise 4.4

DO THE FOLLOWING EXERCISES from section 11.5, pages 385–391:

- Exercise 5.2 *Note: the sufficient decrease condition is at the bottom of page 377.*
- Exercise 5.5 *Hint: Define  $F(\alpha) = f(x_k + \alpha p_k)$  and write down the condition at the minimizing  $\alpha$ . This proof is short.*

DO THE FOLLOWING EXERCISES from section 12.2, pages 408–411:

- Exercise 2.3

DO THE FOLLOWING EXERCISES from section 12.3, pages 420–421:

- Exercise 3.1 *Hints: Write a code to do this; it is easier than doing even one step by hand (in my opinion). The symmetric rank-one update  $B_{k+1} = B_k + \dots$  is on page 413, and again on page 414.*

**Problem P18.** Consider a one-variable problem

$$\min_{x \in \mathbb{R}} f(x)$$

where  $f : \mathbb{R} \rightarrow \mathbb{R}$  is smooth. Recall that the Newton method for this problem solves  $f'(x) = 0$  by the formulas  $p_k = -f'(x_k)/f''(x_k)$  and  $x_{k+1} = x_k + p_k$ . The *secant method for minimization* only differs from the Newton method by replacing the second derivative with a difference quotient approximation based on the last two iterates:

$$f''(x_k) \approx \frac{f'(x_k) - f'(x_{k-1})}{x_k - x_{k-1}}.$$

Thus the secant method computes the step (search vector) by

$$p_k = -\frac{(x_k - x_{k-1})f'(x_k)}{f'(x_k) - f'(x_{k-1})},$$

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<sup>1</sup>Griva, Nash, and Sofer, *Linear and Nonlinear Optimization*, 2nd ed., SIAM Press 2009.

and then it uses  $x_{k+1} = x_k + p_k$  as before.

(a) Implement the secant method. Include a stopping criterion  $|f'(x_k)| < \text{tol}$ .

(b) Use your code to accurately solve  $\min_{x \in \mathbb{R}} f(x)$ , e.g. with  $\text{tol} = 10^{-8}$ , for the following functions and initial iterates:

- i)  $f(x) = x^3 - 2 \sin x$ ,  $x_0 = 0$ ,  $x_1 = 1$
- ii)  $f(x) = 3x^4 - 4x^3 + 3x^2 - 6x$ ,  $x_0 = -1$ ,  $x_1 = 0$

(c) In part ii) above the exact minimum is at  $x_* = 1$ . Compute the errors  $e_k = x_k - x_*$ . Give evidence that the convergence is superlinear. Using the notation of section 2.5, what is your estimate of the rate (exponent)  $r$ ?

**Problem P19.** Please read the discussion in Section 12.2 of why the steepest descent method is slow when applied to minimizing a quadratic function

$$f(x) = \frac{1}{2}x^\top Qx - c^\top x.$$

The conclusion of Lemma 12.4 will make the most sense if you know that the level sets (= contours in  $\mathbb{R}^2$ ) of  $f(x)$  are generalized ellipses, and that the eccentricity of these sets is closely related to the condition number of  $Q$ . This question simply asks you to check this idea on a 2D example.

(a) Consider  $f(x) = x_1^2 + 2x_2^2 - 3x_1$ . What are  $Q$  and  $c$  in this case, and where is the minimizer  $x_*$ ? Consider a contour  $f(x) = \ell$  for some  $\ell \in \mathbb{R}$ . If nonempty, this contour is an ellipse; complete the square to put it in standard form

$$\frac{(x_1 - \gamma)^2}{\alpha^2} + \frac{(x_2 - \delta)^2}{\beta^2} = 1.$$

State  $\alpha, \beta, \gamma, \delta$  in terms of  $\ell$ . What is the ratio  $\alpha/\beta$ ?

(b) Use a contour plotter to plot some contours of the same function  $f(x)$ . Use `axis equal` or similar to make sure that the axes have the same scaling. What is the ratio of the largest to smallest dimensions of the ellipses you see? Compute  $\text{cond}(Q)$ , and relate this value to the ellipses.