

**SECTION C: CONTINUOUS OPTIMISATION**  
**REVISION CLASS 1, PART 2**

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**Problem 1** Let  $f : \mathbb{R}^2 \rightarrow \mathbb{R}$  be defined by  $f(x) = x_1x_2$  where  $x = (x_1, x_2)$ , and consider the trust region  $B_\Delta(\hat{x}) = \{x : \|x - \hat{x}\| \leq \Delta\}$  around  $\hat{x} = (1, 1/2)$ , where  $\|\cdot\|$  denotes the Euclidean norm. We are interested in solving the trust-region subproblem

$$(TR) \quad \min_{x \in B_\Delta(\hat{x})} f(x).$$

- (i) Show that the Cauchy point is  $y^c = (3/8, -3/4)$ .
- (ii) Find an explicit parameterisation of the dogleg path.
- (iii) Show that the objective function is decreasing on the first segment of the dogleg path and increasing on the second segment. Why does this phenomenon occur?
- (iv) Conclude that there exists a critical value  $\Delta_c > 0$  such that the dogleg-solution to (TR) is given by the Cauchy point for all  $\Delta \geq \Delta_c$ .
- (v) Apply Steihaug's method with exact line-searches and find a formula for its solution  $y^s(\Delta)$  as a function of  $\Delta$ .
- (vi) Show that  $f(y^s(\Delta))$  is strictly decreasing as a function of  $\Delta$ .

**Problem 2** Consider the BFGS updating formula

$$B_{k+1} = B_k - \frac{B_k \delta_k \delta_k^T B_k}{\delta_k^T B_k \delta_k} + \frac{\gamma_k \gamma_k^T}{\gamma_k^T \delta_k}$$

- (i) Explain the meaning of each of the terms appearing in this formula and explain their connection with the so-called "secant-condition".

Now let  $f$  be a strictly convex quadratic function in  $x \in \mathbb{R}^n$ ,

$$f(x) = \frac{1}{2}x^T A x + b^T x + c,$$

where  $b \in \mathbb{R}^n$  is a vector and  $A \in \mathbb{R}^{n \times n}$  a symmetric positive definite matrix. Starting with  $B_0 = I$ , we will apply BFGS steps with exact line searches.

- (ii) Let  $S_k$  be the linear subspace of  $\mathbb{R}^n$  such that  $s \in S_k$  if  $B_k s = f''(x_k)s$  and  $\nabla f(x_k)^T s = 0$ . Show that if  $s \in S_k$  then  $\delta_k^T B_k s$  and  $\gamma_k^T s$  are both zero.
- (iii) Conclude from this that  $s \in S_k$  implies  $\nabla f(x_{k+1})^T s = 0$ .
- (iv) Show that  $\delta_k^T B_k s = \gamma_k^T s = 0$  implies  $B_{k+1} s = B_k s$  and conclude that

$$S_0 \subseteq S_1 \subseteq S_2 \subseteq \dots \subseteq S_{k+1}.$$

- (v) Show that  $\delta_k$  is not in  $S_k$ .
- (vi) Show that  $\delta_k \in S_{k+1}$ , and conclude that  $\dim S_k < \dim S_{k+1}$ .
- (vii) Assuming that we stop the procedure when  $\nabla f(x_k) = 0$ , show that the algorithm terminates after at most  $n$  steps.