# AM205 Quiz 4. Optimization. Solution

#### Q1

Suppose that  $g : \mathbb{R} \to \mathbb{R}$  is a nonlinear smooth function with a fixed point  $\alpha \in \mathbb{R}$ , i.e.  $g(\alpha) = \alpha$ . Which of the following statements are true in general?

 $\Box g'(\alpha) = 1$  $\Box |g'(\alpha)| < 1$  $\Box |g'(\alpha)| \le 1$  $\boxtimes \text{ none of the above}$ 

**Answer:** For example, consider a function  $\hat{g}(x) = g(x) + c(x - \alpha)$  which has the same fixed point, but can have an arbitrary derivative depending on *c*.

## Q2

Suppose that a sequence  $x_k$  converges linearly to  $\alpha$ . Define  $y_k = (x_k - \alpha)^2$ . Which of the following statements is true in general?

- $\boxtimes$  *y*<sup>*k*</sup> converges linearly to 0
- $\Box$  *y*<sup>*k*</sup> converges superlinearly to 0

**Answer:** Linear convergence means  $\lim_{k\to\infty} |x_{k+1} - \alpha| / |x_k - \alpha| = \mu$ , where  $0 < \mu < 1$ . For  $y_k$ , we have  $\lim_{k\to\infty} |y_{k+1}| / |y_k| = \lim_{k\to\infty} (x_{k+1} - \alpha)^2 / (x_k - \alpha)^2 = \mu^2$ , so  $y_k$  converges linearly to 0.

### Q3

Consider a scalar equation f(x) = 0 with a smooth and strictly convex function  $f : \mathbb{R} \to \mathbb{R}$ . Which of the following methods are expected to converge **superlinearly**? Assume that the initial guess is chosen sufficiently close to a solution.

- $\hfill\square$  bisection method
- $\boxtimes$  Newton's method
- $\boxtimes$  secant method
- $\hfill\square$  none of the above

#### Q4

Consider a continuous function  $f : \mathbb{R} \to \mathbb{R}$ . Which of the following statements are true?

- $\boxtimes$  if *f* is coercive on  $\mathbb{R}$ , then *f* has a global minimum in  $\mathbb{R}$
- $\Box$  if *f* has a unique global minimum in  $\mathbb{R}$ , then *f* is coercive on  $\mathbb{R}$
- $\Box$  none of the above

#### Q5

The function f(x) = |x| defined on  $\mathbb{R}$  is

- $\boxtimes$  coercive
- $\boxtimes$  convex
- $\Box$  strictly convex
- $\hfill\square$  none of the above

# Q6

The Hessian of the function  $f(x, y) = x^2 + y^2$  is

- $\boxtimes$  positive definite
- $\Box$  negative definite
- $\hfill\square$  indefinite
- $\hfill\square$  none of the above

# Q7

To optimize a function  $f : \mathbb{R}^n \to \mathbb{R}$ , the BFGS algorithm relies on evaluations of

- $\Box$  the function *f*
- $\boxtimes$  the gradient  $\nabla f$
- $\Box$  the Hessian  $H_f$

## Q8

Recall the Lagrangian function  $\mathcal{L}(b, \lambda) = b^T b + \lambda^T (Ab - y)$  corresponding to an underdetermined linear least squares problem. Assume that  $A \in \mathbb{R}^{m \times n}$  has full rank and  $m \leq n$ . Suppose that this function is minimized using Newton's method with a zero initial guess  $b_0 = 0$  and  $\lambda_0 = 0$ . How many iterations would Newton's method need to satisfy  $\|\nabla \mathcal{L}\|_2 < 10^{-5}$ ?

- $\boxtimes$  one
- $\Box$  depends on  $||A||_2$
- $\Box$  depends on  $||A||_2$  and  $||y||_2$

**Answer:** Newton's method solves any quadratic optimization problem exactly after one iteration.