

## Nonlinear Optimization Exercises 2

- (2) Given a convex set  $C \subseteq \mathbb{R}^n$ . A function  $f: C \rightarrow \mathbb{R}$  is *quasiconvex*, if  $f(\alpha x_1 + (1 - \alpha)x_2) \leq \max\{f(x_1), f(x_2)\}$  for all  $x_1, x_2 \in C$  and all  $\alpha \in (0, 1)$ . Prove:  $f$  is quasiconvex if and only if all level sets  $S_f(r) = \{x \in C: f(x) \leq r\}$  are convex. Are all local minima global ones as well?
- (3) Given a convex set  $C \subseteq \mathbb{R}^n$ . A quasiconvex function  $f: C \rightarrow \mathbb{R}$  is *strictly quasiconvex*, if for each  $x_1, x_2 \in C$  with  $f(x_1) \neq f(x_2)$  we have  $f(\alpha x_1 + (1 - \alpha)x_2) < \max\{f(x_1), f(x_2)\}$  for all  $\alpha \in (0, 1)$ . Prove: If  $f$  is strictly quasiconvex, any local minimum is also a global minimum and the set of minima  $\text{Argmin}(f)$  is convex. What is lost, if in the definition of strict quasiconvexity the requirement is dropped that  $f$  is quasiconvex?
- (3) Given a strictly quasiconvex  $f: [a, b] \rightarrow \mathbb{R}$  and  $a \leq \lambda < \mu \leq b$ , prove: If  $f(\lambda) < f(\mu)$ , then  $f(\alpha) \geq f(\mu)$  for  $\alpha \in [\mu, b]$ , otherwise  $f(\alpha) \geq f(\lambda)$  for  $\alpha \in [a, \lambda]$ .
- (4) Derivative free line-search: Let a strictly quasiconvex  $f: [a, b] \rightarrow \mathbb{R}$  be given by a 0<sup>th</sup> order oracle.

- (a) Describe an algorithm with the following property: In iteration  $0 \leq k \leq n - 1$  the search interval  $[a_k, b_k]$  for the minimum is reduced by exploiting 3 after evaluating the function in

$$\lambda_k = a_k + \frac{F_{n-k-1}}{F_{n-k+1}}(b_k - a_k) \text{ and } \mu_k = a_k + \frac{F_{n-k}}{F_{n-k+1}}(b_k - a_k),$$

where  $F_0 = F_1 = 1$  and  $F_{n+2} = F_{n+1} + F_n$ ,  $n \in \mathbb{N}_0$ , is the Fibonacci sequence. What is the total number of evaluations needed?

- (b) The Golden Section Method: Now use  $\theta = \frac{\sqrt{5}-1}{2}$  and

$$\lambda_k = a_k + (1 - \theta)(b_k - a_k), \quad \mu_k = a_k + \theta(b_k - a_k).$$

By how much is the search interval reduced per iteration?

5. (8) Develop and implement in matlab a line search function of the form

```
[npoint,nval,ngrad]=linesearch(@fun,cpoint,dir,maxa,cval,cgrad,c1,c2);
```

For a continuously differentiable function  $f$  it should perform an interpolation search on  $f(x_c + \alpha p)$  for  $\alpha \in [0, \bar{\alpha}]$  so that the Wolfe conditions are satisfied for parameters  $c_1$  and  $c_2$ . The parameters of the routine are **fun** for  $f$ , **cpoint** for  $x_c$ , **dir** for the descent direction  $p$ , **maxa** for  $\bar{\alpha}$ , **cval** for  $f(x_c)$  and **cgrad** for  $\nabla f(x_c)$ . Return values are the new point, the new function value and the new gradient. All vectors/points are column vectors. For **fun** it must be possible to plug in an arbitrary function of the form

```
[val,grad]=function_name(point)
```

(type "help feval" in matlab)