

(1) Let

$$K := \mathbb{R}_-^s \times \{0\}^{m-s} = \{v \mid 0 \leq v_i, i = 1, \dots, s, v_i = 0 \ i = s+1, \dots, m\}.$$

(a) Show that  $K^\circ = \mathbb{R}_+^s \times \mathbb{R}^{m-s}$ .

(b) Given  $z \in K$ , show that  $N_K(z) = \{w \in K^\circ \mid w_i z_i = 0 \ i = 1, \dots, s\}$ .

(2) Let  $A \in \mathbb{R}^{m \times n}$  and  $b \in \mathbb{R}^m$ , and denote the  $i$ th row of  $A$  by  $a_i \in \mathbb{R}^n, i = 1, \dots, m$ . Let  $K := \mathbb{R}_-^s \times \{0\}^{m-s}$  and set  $\Omega := \{x \mid Ax - b \in K\}$  and let  $\bar{x} \in \Omega$ .

(a) Show that

$$T_\Omega(\bar{x}) = \{d \in \mathbb{R}^n \mid a_i^T d \leq 0 \ i \in I(\bar{x}) \text{ and } a_i^T d = 0 \ i = s+1, \dots, m\} = \{d \in \mathbb{R}^n \mid Ad \in T_K(A\bar{x} - b)\} =: A^{-1}T_K(A\bar{x} - b),$$

$$\text{where } I(\bar{x}) := \{i \in \{1, \dots, s\} \mid a_i^T \bar{x} = 0\}.$$

(b) Show that

$$N_\Omega(\bar{x}) = \left\{ \sum_{i \in I(\bar{x})} u_i a_i + \sum_{i=s+1}^m u_i a_i \mid 0 \leq u_i \ i \in I(\bar{x}) \right\} = \{A^T u \mid u \in N_K(A\bar{x} - b)\} =: A^T N_K(A\bar{x} - b).$$

(3) Give an example of a smooth nonlinear function  $f : \mathbb{R}^n \rightarrow \mathbb{R}$ , a pair  $x, d \in \mathbb{R}^n$ , and a number  $0 < c < 1$  for which  $f'(x; d) < 0$  but the backtracking line-search fails to terminate in a finite number of steps.

(4) Let  $F : \mathbb{R}^n \rightarrow \mathbb{R}^n$  be continuously differentiable with  $F'(x)$  invertible on  $\mathbb{R}^n$ . The Newton iteration for solving  $F(x) = 0$  is given by

$$x^{k+1} = x^k - \nabla F(x^k)^{-1} F(x^k).$$

Although this iteration converges at a quadratic rate locally, it may not converge if the initial point  $x^0$  is too far from a solution. To compensate for this deficiency, the Newton iteration is often replaced by a *damped* Newton iteration of the form

$$x^{k+1} = x^k - t_k \nabla F(x^k)^{-1} F(x^k)$$

for some choice of  $t_k > 0$ . We consider one approach to choosing  $t_k$ .

(a) Consider the function  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  given by  $f(x) = \|F(x)\|_2$ . Show that if  $x \in \mathbb{R}^n$  is such that  $F(x) \neq 0$ , then  $f$  is differentiable at  $x$  with

$$\nabla f(x) = \nabla F(x)^T \frac{F(x)}{f(x)}.$$

(b) Note that one can attempt to solve  $F(x) = 0$  by minimizing the function  $f(x)$ . This indicates that a damped Newton step size  $t_k > 0$  can be computed by doing a line search on the function  $f$  in the Newton direction  $d_N = -\nabla F(x)^{-1} F(x)$ . As a first step, show that the Newton direction  $d_N$  is a direction of descent at points  $x$  for which  $F(x) \neq 0$  by showing that

$$f'(x; d_N) = -\|F(x)\|_2.$$

(c) Show that the backtracking line search applies and takes the form

$$t_k = \max_{\substack{\gamma^s \\ \text{s.t. } s \in \{0, 1, \dots\} \\ \|F(x^k + \gamma^s d_N^k)\|_2 \leq (1 - c\gamma^s) \|F(x^k)\|_2}} \gamma^s,$$

where  $\gamma \in (0, 1)$  and  $c \in (0, 1)$  are the line search parameters.

(5) Let  $F : \mathbb{R}^n \rightarrow \mathbb{R}^m$  be continuously differentiable and consider the problem of finding a point  $x$  such that  $F(x) = 0$ . This is the same problem that led to the development of Newton's method except now we are interested in the case where  $n \neq m$ . If  $m < n$ , then there are most likely infinitely many solutions. On the other-hand, if  $m > n$ , then there are probably no solutions. In this problem we consider the case when  $m > n$ . Since it is unlikely that there exists an  $x$  satisfying  $F(x) = 0$ , we instead try to find an  $x$  that makes  $\|F(x)\|$  as small as possible. For this we define  $f(x) = \frac{1}{2} \|F(x)\|_2^2$ . It is easily shown (try it) that

$$\nabla f(x) = \nabla F(x)^T F(x)$$

$$\nabla^2 f(x) = \nabla F(x)^T \nabla F(x) + \sum_{i=1}^m F_i(x) \nabla^2 F_i(x),$$

where the functions  $F_i : \mathbb{R}^n \rightarrow \mathbb{R}$  are the component functions of  $F$ .

- (a) In this setting we still try to follow the path suggested by Newton where we successively solve for the search direction using the linearization of  $F$ . But in this case the search direction is given as the solution to the linear least squares problem

$$\mathcal{GN} \quad \min_{d \in \mathbb{R}^n} \frac{1}{2} \|F(x^k) + \nabla F(x^k)d\|_2^2 .$$

Methods using the solution to this problem as a search direction are called *Gauss-Newton* methods. Denote a solution by  $d_{GN}$ . Under what conditions does this problem have a unique solution?

- (b) When a unique solution to the problem  $\mathcal{GN}$  exists give a formula for  $d_{GN}$ .  
(c) Use this formula for  $d_{GN}$  to show that it is a descent direction for  $f$  giving a formula for  $f'(x^k; d_{GN})$ .  
(d) Show that  $d_{GN}$  is also a descent direction for the function  $h(x) = \|F(x)\|_2$ .  
(e) Assuming that  $\nabla F(x)$  is continuous, show that

$$h'(x; d) \leq \|F(x) + \nabla F(x)d\|_2 - h(x) .$$