49.14. PROBLEMS 1725

Problem 49.7. If P is a symmetric positive definite matrix, prove that $||z||_P = (z^\top Pz)^{1/2} = ||P^{1/2}z||_2$ is a norm. Prove that the normalized steepest descent direction is

$$d_{\text{nsd,k}} = -(\nabla J_{u_k}^{\top} P^{-1} \nabla J_{u_k})^{-1/2} P^{-1} \nabla J_{u_k},$$

the dual norm is $\|z\|^D = \|P^{-1/2}z\|_2$, and the steepest descent direction with respect to $\|\cdot\|_P$ is given by

$$d_{\mathrm{sd},k} = -P^{-1}\nabla J_{u_k}.$$

Problem 49.8. If $\| \|$ is the ℓ^1 -norm, then show that $d_{\text{nsd,k}}$ is determined as follows: let i be any index for which $\|\nabla J_{u_k}\|_{\infty} = |(\nabla J_{u_k})_i|$. Then

$$d_{\mathrm{nsd,k}} = -\mathrm{sign}\left(\frac{\partial J}{\partial x_i}(u_k)\right)e_i,$$

where e_i is the *i*th canonical basis vector, and

$$d_{\rm sd,k} = -\frac{\partial J}{\partial x_i}(u_k)e_i.$$

Problem 49.9. (From Boyd and Vandenberghe [29], Problem 9.12). If $\nabla^2 f(x)$ is singular (or very ill-conditioned), the Newton step $d_{\rm nt} = -(\nabla^2 J(x))^{-1} \nabla J_x$ is not well defined. Instead we can define a search direction $d_{\rm tr}$ as the solution of the problem

$$\begin{array}{ll} \text{minimize} & (1/2)\langle Hv,v\rangle + \langle g,v\rangle \\ \text{subject to} & \left\|v\right\|_2 \leq \gamma, \end{array}$$

where $H = \nabla^2 f_x$, $g = \nabla f_x$, and γ is some positive constant. The idea is to use a trust region, which is the closed ball $\{v \mid ||v||_2 \leq \gamma\}$. The point $x + d_{\text{tr}}$ minimizes the second-order approximation of f at x, subject to the constraint that

$$||x + d_{\rm tr} - x||_2 \le \gamma.$$

The parameter γ , called the *trust parameter*, reflects our confidence in the second-order approximation.

Prove that $d_{\rm tr}$ minimizes

$$\frac{1}{2}\langle Hv, v \rangle + \langle g, v \rangle + \widehat{\beta} \|v\|_{2}^{2},$$

for some $\widehat{\beta}$.

Problem 49.10. (From Boyd and Vandenberghe [29], Problem 9.9). Prove that the Newton decrement $\lambda(x)$ is given by

$$\lambda(x) = \sup_{v \neq 0} -\frac{\langle \nabla J_x, v \rangle}{(\langle \nabla^2 J_x v, v \rangle)^{1/2}}.$$