

# CNAc: Continuous Optimization

## Solutions to problem set 3 — trust-region methods

Honour School of Mathematics, Oxford University  
Hilary Term 2006, Dr Nick Gould

**Problem 1.**

The Cauchy point is  $s_k^C = -\alpha_k^C g_k$  where

$$\alpha_k^C = \arg \min_{0 < \alpha \leq \Delta_k / \|g_k\|} m_k(-\alpha g_k).$$

To find  $\alpha_k^C$ :

- if  $g_k^T B_k g_k \leq 0$ ,

$$\alpha_k^C = \frac{\Delta_k}{\|g_k\|}.$$

- otherwise,

$$\alpha_k^C = \min \left( \frac{\|g_k\|_2^2}{g_k^T B_k g_k}, \frac{\Delta_k}{\|g_k\|} \right).$$

**Problem 2.**

- (a) The unconstrained minimizer  $-(1, 0, 1/2)^T$  has  $\ell_2$ -norm  $1 < \sqrt{5}/2 < 2$ . Thus, since  $B$  is positive definite, the unconstrained minimizer solves the problem.
- (b) The unconstrained minimizer has too large a  $\ell_2$ -norm, so the solution must lie on the boundary of the constraint. The solution must be of the form  $-(1/(1+\lambda), 0, 1/(2+\lambda))^T$ . To satisfy the trust-region constraint, we then must have

$$\frac{1}{(1+\lambda)^2} + \frac{1}{(2+\lambda)^2} = \Delta^2 = \frac{25}{144}$$

which has a root  $\lambda = 2$ . Thus the required solution is  $-(1/3, 0, 1/4)^T$ .

- (c) 1(c). The Hessian is indefinite so the solution must lie on the boundary of the constraint. The solution is then of the form  $-(1/(-2+\lambda), 0, 1/(-1+\lambda))^T$ . To satisfy the trust-region constraint, we then must have

$$\frac{1}{(-2+\lambda)^2} + \frac{1}{(-1+\lambda)^2} = \Delta^2 = \frac{25}{144}$$

which has a root  $\lambda = 5$  (c.f. the previous equation with a change of variables  $\hat{\lambda} = \lambda + 3$ ) at which  $B + \lambda I$  is positive semi-definite. Thus again the solution is  $-(1/3, 0, 1/4)^T$ .

- (d) Again  $B$  is indefinite, and so the solution must be of the form  $-(\omega, 0, 1/(-1+\lambda))^T$ , where  $\omega = 0/(-2+\lambda)$  can only be nonzero if  $\lambda = 2$ —note that  $B + \lambda I$  is only positive semi-definite when  $\lambda \geq 2$ . Suppose that  $\lambda > 2$ . To satisfy the trust-region constraint, we then must have

$$\frac{1}{(-1+\lambda)^2} = \Delta^2 = \frac{1}{4}$$

which has roots  $1 \pm 2$ . The desired root is  $\lambda = 3$ , from which we deduce the solution is  $-(0, 0, 1/2)^T$ .

(e) As in (d), if we guess that  $\lambda > 2$ , we find that the roots of

$$\frac{1}{(-1 + \lambda)^2} = \Delta^2 = 2$$

are  $1 \pm 1/\sqrt{2} < 2$ . So  $\lambda$  must be 2, and the solution is of the form  $-(\omega, 0, 1)^T$ . To satisfy the trust-region constraint, we then must have

$$\omega^2 + 1 = \Delta^2 = 2,$$

and hence  $\omega = \pm 1$ . Thus the required solution is  $-(\pm 1, 0, 1)^T$ .

### Problem 3.

The problem is convex, so the required solution will satisfy  $(B + \lambda I)s = -g$  for some  $\lambda \geq 0$ . Thus

$$s = - \begin{pmatrix} \frac{1}{1+\lambda} \\ \frac{1}{3+\lambda} \end{pmatrix}.$$

We illustrate this trajectory in Figure 1.

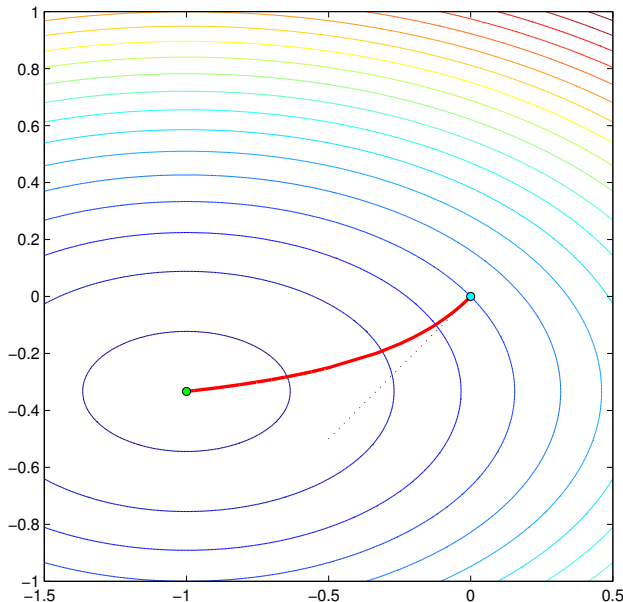


Figure 1: The trust-region minimizer of the quadratic function  $0.5s_1^2 + 1.5s_2^2 + s_1 + s_2$  as the trust-region radius increases from zero to infinity. Note that the tangent to the trajectory when the radius is zero is in the steepest-descent direction.

The Lagrange multiplier and the trust-region radius satisfy the relationship

$$\frac{1}{(1 + \lambda)^2} + \frac{1}{(3 + \lambda)^2} = \Delta^2 \tag{1}$$

for  $\lambda \geq 0$ . Letting  $\theta = \lambda + 2$ , (1) implies

$$\Delta^2(\theta^2 - 1)^2 = 2\theta^2 + 2.$$

from which trivial calculations show that

$$\theta = \frac{\sqrt{1 + \Delta^2} + \sqrt{1 + 4\Delta^2}}{\Delta},$$

i.e.,

$$\lambda = -2 + \frac{\sqrt{1 + \Delta^2} + \sqrt{1 + 4\Delta^2}}{\Delta}.$$

Using (1), the solution becomes unconstrained when  $\lambda = 0$ , that is when  $\Delta = \sqrt{10}/3$ .

**Problem 4.**

Let  $\bar{s} = Rs$ . Then the problem we wish to solve is equivalent to

$$\underset{\bar{s} \in \mathbb{R}^n}{\text{minimize}} \quad \bar{s}^T \bar{g} + \frac{1}{2} \bar{s}^T \bar{B} \bar{s} \quad \text{subject to} \quad \|\bar{s}\|_2 \leq \Delta. \quad (2)$$

where  $\bar{g} = R^{-T}g$  and  $\bar{B} = R^{-T}HR^{-1}$ . But then Theorem 3.9 gives that the global minimizer  $\bar{s}_*$  of (2) satisfies

$$(\bar{B} + \lambda_* I) \bar{s}_* = -\bar{g}, \quad (3)$$

where  $\bar{B} + \lambda_* I$  is positive semi-definite,  $\lambda_* \geq 0$  and  $\lambda_*(\|\bar{s}_*\|_2 - \Delta) = 0$ . Equation (3) is equivalent to

$$(B + \lambda_* M) s_* = -g, \quad (4)$$

while

$$\bar{B} + \lambda_* I = R^{-T}(B + \lambda_* M)R^{-1}$$

which is positive semi-definite if and only if  $B + \lambda_* M$  is. Thus the required conditions are that  $s_*$  satisfies (4) where  $B + \lambda_* M$  is positive semi-definite,  $\lambda_* \geq 0$  and  $\lambda_*(\|s_*\|_M - \Delta) = 0$ .