6.252 NONLINEAR PROGRAMMING

LECTURE 5: RATE OF CONVERGENCE

LECTURE OUTLINE

- Approaches for Rate of Convergence Analysis
- The Local Analysis Method
- Quadratic Model Analysis
- The Role of the Condition Number
- Scaling
- Diagonal Scaling
- Extension to Nonquadratic Problems
- Singular and Difficult Problems

APPROACHES FOR RATE OF CONVERGENCE ANALYSIS

- Computational complexity approach
- Informational complexity approach
- Local analysis
- Why we will focus on the local analysis method

THE LOCAL ANALYSIS APPROACH

- Restrict attention to sequences x^k converging to a local min x^*
- Measure progress in terms of an error function e(x) with $e(x^*) = 0$, such as

 $e(x) = ||x - x^*||, \qquad e(x) = f(x) - f(x^*)$

• Compare the tail of the sequence $e(x^k)$ with the tail of standard sequences

• Geometric or linear convergence [if $e(x^k) \le q\beta^k$ for some q > 0 and $\beta \in [0, 1)$, and for all k]. Holds if

$$\limsup_{k \to \infty} \frac{e(x^{k+1})}{e(x^k)} < \beta$$

- Superlinear convergence [if $e(x^k) \leq q \cdot \beta^{p^k}$ for some q > 0, p > 1 and $\beta \in [0, 1)$, and for all k].
- Sublinear convergence

QUADRATIC MODEL ANALYSIS

- Focus on the quadratic function f(x) = (1/2)x'Qx, with Q > 0.
- Analysis also applies to nonquadratic problems in the neighborhood of a nonsingular local min
- Consider steepest descent

$$x^{k+1} = x^k - \alpha^k \nabla f(x^k) = (I - \alpha^k Q) x^k$$

$$||x^{k+1}||^2 = x^{k'} (I - \alpha^k Q)^2 x^k$$

$$\leq (\max \text{ eig. } (I - \alpha^k Q)^2) ||x^k||^2$$

The eigenvalues of $(I - \alpha^k Q)^2$ are equal to $(1 - \alpha^k \lambda_i)^2$, where λ_i are the eigenvalues of Q, so

max eig of
$$(I-\alpha^k Q)^2 = \max\left\{(1-\alpha^k m)^2, (1-\alpha^k M)^2\right\}$$

where m, M are the smallest and largest eigenvalues of Q. Thus

$$\frac{\|x^{k+1}\|}{\|x^k\|} \le \max\{|1 - \alpha^k m|, |1 - \alpha^k M|\}$$

OPTIMAL CONVERGENCE RATE

• The value of α^k that minimizes the bound is $\alpha^* = 2/(M+m)$, in which case



Conv. rate for minimization stepsize (see text)

$$\frac{f(x^{k+1})}{f(x^k)} \le \left(\frac{M-m}{M+m}\right)^2$$

• The ratio M/m is called the *condition number* of Q, and problems with M/m: large are called *ill-conditioned*.

SCALING AND STEEPEST DESCENT

• View the more general method

$$x^{k+1} = x^k - \alpha^k D^k \nabla f(x^k)$$

as a scaled version of steepest descent.

• Consider a change of variables x = Sy with $S = (D^k)^{1/2}$. In the space of y, the problem is

minimize
$$h(y) \equiv f(Sy)$$

subject to $y \in \Re^n$

• Apply steepest descent to this problem, multiply with S, and pass back to the space of x, using $\nabla h(y^k) = S \nabla f(x^k)$,

$$y^{k+1} = y^k - \alpha^k \nabla h(y^k)$$
$$Sy^{k+1} = Sy^k - \alpha^k S \nabla h(y^k)$$
$$x^{k+1} = x^k - \alpha^k D^k \nabla f(x^k)$$

DIAGONAL SCALING

• Apply the results for steepest descent to the scaled iteration $y^{k+1} = y^k - \alpha^k \nabla h(y^k)$:

$$\frac{\|y^{k+1}\|}{\|y^k\|} \le \max\{|1 - \alpha^k m^k|, |1 - \alpha^k M^k|\}$$

$$\frac{f(x^{k+1})}{f(x^k)} = \frac{h(y^{k+1})}{h(y^k)} \le \left(\frac{M^k - m^k}{M^k + m^k}\right)^2$$

where m^k and M^k are the smallest and largest eigenvalues of the Hessian of h, which is

$$\nabla^2 h(y) = S \nabla^2 f(x) S = (D^k)^{1/2} Q(D^k)^{1/2}$$

• It is desirable to choose D^k as close as possible to Q^{-1} . Also if D^k is so chosen, the stepsize $\alpha = 1$ is near the optimal $2/(M^k + m^k)$.

• Using as D^k a diagonal approximation to Q^{-1} is common and often very effective. Corrects for poor choice of units expressing the variables.

NONQUADRATIC PROBLEMS

• Rate of convergence to a nonsingular local minimum of a nonquadratic function is very similar to the quadratic case (linear convergence is typical).

• If $D^k \to (\nabla^2 f(x^*))^{-1}$, we asymptotically obtain optimal scaling and superlinear convergence

• More generally, if the direction $d^k = -D^k \nabla f(x^k)$ approaches asymptotically the Newton direction, i.e.,

$$\lim_{k \to \infty} \frac{\|d^k + (\nabla^2 f(x^*))^{-1} \nabla f(x^k)\|}{\|\nabla f(x^k)\|} = 0$$

and the Armijo rule is used with initial stepsize equal to one, the rate of convergence is superlinear.

• Convergence rate to a singular local min is typically sublinear (in effect, condition number $= \infty$)