

# **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

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

- Compare the tail of the sequence  $e(x^k)$  with the tail of standard sequences
- Geometric or linear convergence [if  $e(x^k) \leq q\beta^k$  for some  $q > 0$  and  $\beta \in [0, 1)$ , and for all  $k$ ]. Holds if

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

- 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

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

$$\begin{aligned} \|x^{k+1}\|^2 &= x^{k'}(I - \alpha^k Q)^2 x^k \\ &\leq (\max \text{ eig. } (I - \alpha^k Q)^2) \|x^k\|^2 \end{aligned}$$

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

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

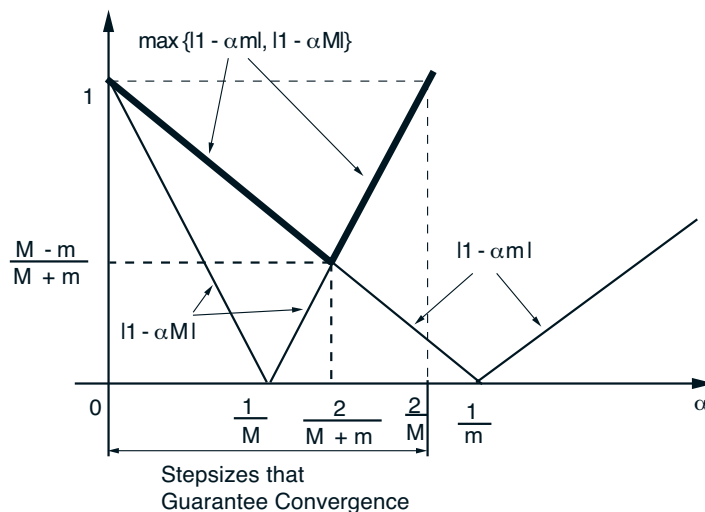
where  $m, M$  are the smallest and largest eigenvalues of  $Q$ . Thus

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

# OPTIMAL CONVERGENCE RATE

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

$$\frac{\|x^{k+1}\|}{\|x^k\|} \leq \frac{M - m}{M + m}$$



- Conv. rate for minimization stepsize (see text)

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

- 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

$$\text{minimize } h(y) \equiv f(Sy)$$

$$\text{subject to } y \in \mathbb{R}^n$$

- 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  $y^{k+1} = y^k - \alpha^k \nabla h(y^k)$ :

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

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

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  $D^k \rightarrow (\nabla^2 f(x^*))^{-1}$ , 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.,

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

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 =  $\infty$ )