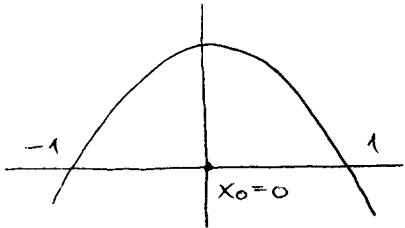


Globally convergent Modification of Newton's Method

10/12/2006

(5)

- Augment Newton with "globalization"
- To increase likelihood of convergence to some solution from an arbitrary initial guess.
- Note:
 - convergence is not (cannot) be guaranteed
 - also convergence is not necessarily be to any distinguished solution, e.g. a global minimizer
- Structure:
 - Begin with initial trial step (Newton step or something related to it)
 - Test for adequate progress
 - Modify if necessary to get a new trial step; return to test
- Criteria of which to base test
 - Suppose have $\{x_k\}$
Want criteria that ideally imply $x_k \rightarrow x_*$ where $F(x_*) = 0$
Reasonable: require $\|F(x_{k+1})\| < \|F(x_k)\|$ for each k
This isn't enough!
Example: $f(x) = 1 - x^2$

a) $x_k = x_{k-1} + \frac{1}{2^{k+1}}$
 $\Rightarrow x_0 = 0$
 $x_1 = 1/4$
 $x_2 = 3/8$
 $x_3 = 7/16$
 $\vdots$
 $x_k \rightarrow 1/2$
and $f(1/2) \neq 0$
have $f(x_{k+1}) < f(x_k)$ but steps are too short

$$b) X_k = -X_{k-1} + (-1)^{k-1} 2^{-k}$$

$$\Rightarrow X_0 = 0 ; X_1 = \frac{1}{2} ; X_2 = -\frac{3}{4} ; X_3 = \frac{7}{8} , \dots$$

see $f(X_{k+1}) < f(X_k)$ again but $\lim_{k \rightarrow \infty} X_k$ doesn't exist

Problem: steps are too long

$X_* = 1$ and $X_{**} = -1$ are solutions and are both

limit points of $\{X_k\}$

Def: X_* is limit point of $\{X_k\}$ if for any $\delta > 0$ and any k_0 , there is $k > k_0$ such that

$$\|X_k - X_*\| < \delta$$

Note: X_* is a limit point of $\{X_k\}$ then there is a subsequence $\{X_{k_j}\}$ that converges to X_*

Def: $\{X_k\}$ converges to X_* $\left(\lim_{k \rightarrow \infty} X_k = X_* \right)$ if, for any $\delta > 0$ $\|X_k - X_*\| < \delta$ for all sufficiently large k

Will base criteria on:

actual reduction in $\|F\|$:

$$\text{ared}_k \equiv \|F(X_k)\| - \|F(X_{k+1})\|$$

predicted reduction in $\|F\|$

$$\text{pred}_k \equiv \|F(X_k)\| - \|F(X_k) + F'(X_k) S_k\|$$

$$\text{where } S_k = X_{k+1} - X_k$$

(predicted by local linear model $F(X_k) + F'(X_k) \cdot S_k$)

relative predicted reduction

$$\text{relpred}_k = \frac{\text{pred}_k}{\|F(X_k)\|}$$

- Will require each $\text{pred}_k \geq 0$

- Will also require

$$\text{ared}_k \geq t \cdot \text{pred}_k \quad \left(\begin{array}{c} \text{sufficient decrease} \\ \text{condition} \end{array} \right)$$

for $t \in (0, 1)$ independent of k

- Will not need to require a minimum pred_k for every step, will require a minimum cumulative reduction over all steps

Theorem: (Eisenstat-Walter 1994)

Suppose F is continuously differentiable and $\{x_k\}$ is such that for each k (and $s_k = x_{k+1} - x_k$)

$$\text{ared}_k \geq t \cdot \text{pred}_k \geq 0$$

for $t \in (0, 1)$ independent of k .

If $\sum_{k=1}^{\infty} \text{ared}_k = \infty$, then $F(x_k) \rightarrow 0$

If also $\{x_k\} > 0$ has a limit point x_* such that $F'(x_*)$ is nonsingular, then $F(x_*) = 0$ and $x_k \rightarrow x_*$

Remarks: Theorem says exactly one of following must hold: (assuming $\sum_{k=1}^{\infty} \text{ared}_k = \infty$)

1) $\|x_k\| \rightarrow \infty$ (no limit points)

2) $\exists$ limit points, and F' is singular at each

3) $x_k \rightarrow x_*$ with $F(x_*) = 0$, $F'(x_*)$ - nonsingular

The two "bad" cases can happen if F allows them to

example: $f(x) = e^{-x^2}$

for any $x_0 \neq 0$ Newton iterates $\{x_k\}$ satisfy

$$\text{ared}_k \geq t \text{pred}_k \geq 0 \quad \text{with } t = \frac{1}{4}$$

for sufficiently large x_k , and

$$x_k \rightarrow \pm \infty$$

- Not hard to construct examples of $\{x_k\}$ with multiple limit points and F' singular at each (see Eisenstat - Wolkow '94)

- Also may not be possible to have $\sum \text{vecpred}_k = \infty$

Example: $f(x) = 1+x^2$

- If $\{x_k\}$ were such that $\sum \text{vecpred}_k = \infty$ would have $f(x_k) \rightarrow 0$

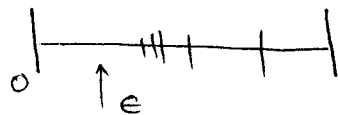
proof of theorem:

easy part:

$$\begin{aligned}
 \|F(x_0)\| &\geq \|F(x_0)\| - \|F(x_{k+1})\| \\
 &= \sum_{j=0}^k \text{aved}_j \geq t \sum_{j=0}^k \text{pred}_j \\
 &= t \sum_{j=0}^k \text{vecpred}_j \|F(x_j)\|
 \end{aligned}$$

$\bar{t} \|F(x_1)\| + \|F(x_2)\| + \dots + \|F(x_k)\|$

see: $\text{aved}_j \geq t \text{pred}_j \geq 0 \Rightarrow \|F(x_j)\|$ monotone decreasing



if $\|F(x_j)\| \rightarrow 0$ then $\exists \epsilon > 0$ such that

$$\|F(x_j)\| \geq \epsilon \text{ for all } j$$

Then

$$\|F(x_0)\| \geq t \cdot \epsilon \sum_{j=0}^k \text{vecpred}_j \text{ for all } k$$

$$\Rightarrow \sum_{j=0}^{\infty} \text{vecpred}_j < \infty$$

so: $\sum \text{vecpred}_j = \infty \Rightarrow \|F(x_j)\| \rightarrow 0$

Hard part:

Suppose $\sum \text{relpred}_k = \infty$ (so $F(x_k) \rightarrow 0$)

and x_* is a limit point of $\{x_k\}$ such that $F'(x_*)$ is nonsingular.

Trivially $F(x_*) = 0$ by continuity of F

To show: $x_k \rightarrow x_*$

By Handy Lemma (VI)

$$\|F(x) - \cancel{F(x_*)} - F'(x_*)(x - x_*)\| \leq \omega(\|x - x_*\|) \|x - x_*\|$$

where

$$\omega(\delta) \equiv \max_{\|d\| < \delta} \|F'(x_* + d) - F'(x_*)\| \rightarrow 0 \text{ as } \delta \rightarrow 0$$

Then

$$\omega(\|x - x_*\|) \|x - x_*\| \geq \|F'(x_*)(x - x_*)\| - \|F(x)\|$$

$$\Rightarrow \|F(x)\| \geq \|F'(x_*)(x - x_*)\| - \omega(\|x - x_*\|) \|x - x_*\|$$

$$\# \|v\| = \|A^{-1}Av\| \leq \|A^{-1}\| \|Av\|$$

$$\Rightarrow \|Av\| \geq \frac{\|v\|}{\|A^{-1}\|}$$

#

$$\geq \left[\frac{1}{K} - \omega(\|x - x_*\|) \right] \|x - x_*\|, \quad K = \|F'(x_*)\|$$

Suppose $\delta > 0$ so small that $\omega(\delta) < \frac{1}{2K}$

$$\text{Then } \|x - x_*\| \leq \delta \Rightarrow \|F(x)\| \geq \frac{1}{2K} \|x - x_*\|$$

In particular, if $\|x_k - x_*\| \leq \delta$ then

$$\|F(x_k)\| \geq \frac{1}{2K} \|x_k - x_*\|$$

Since $F(x_k) \rightarrow 0$, only need to show

$$\|x_k - x_*\| \leq \delta \text{ for all large } k$$

choose δ smaller if necessary so that

$$\|x - x_*$$

if x_k is such that $\|x_k - x_*\| \leq \delta$ then

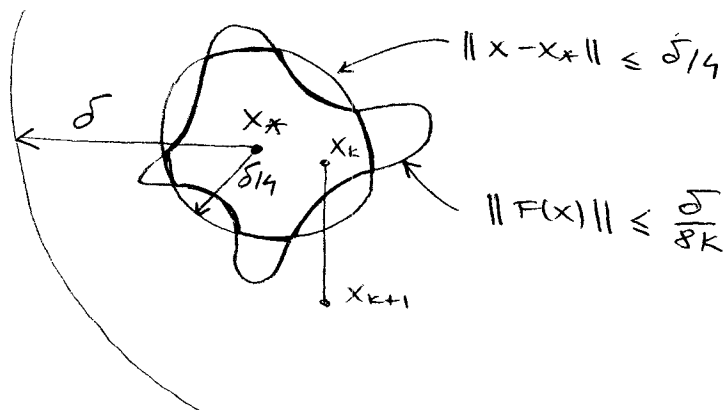
$$\|s_k\| = \|F'(x_k)^{-1} \cdot \{F'(x_k)s_k + F(x_k) - F(x_k)\}\|$$

$$\leq \|F'(x_k)^{-1}\| \cdot \left\{ \|F(x_k)\| + \underbrace{\|F(x_k) + F'(x_k)s_k\|}_{\|F(x_k)\|} \right\}$$

$\|F(x_k)\|$ since $\alpha \geq 0$

$$\leq 2k \cdot 2 \|F(x_k)\| = 4k \|F(x_k)\|$$

$$\text{Set } S = \left\{ x : \|x - x_*\| \leq \frac{\delta}{4} \text{ and } \|F(x)\| \leq \frac{\delta}{8k} \right\}$$



We'll show $x_k \in S$ for all sufficiently large k

Since x_* is a limit point $x_k \in S$ for some large k

suffice to show: $x_k \in S \Rightarrow x_{k+1} \in S$

Then

$$\|s_k\| \leq 4k \|F(x_k)\| \leq 4k \frac{\delta}{8k} = \frac{\delta}{2}$$

$$\Rightarrow \|x_{k+1} - x_*\| \leq \frac{\delta}{4} + \frac{\delta}{2} < \delta$$

But need to show $x_{k+1} \in S$

Have $\|F(x_{k+1})\| \leq \|F(x_k)\| = \frac{\delta}{8k}$

but also

$$\|x_{k+1} - x_*\| \leq 2k \|F(x_{k+1})\|$$

$$\leq 2k \|F(x_k)\|$$

$$\leq 2k \frac{\delta}{8k} = \frac{\delta}{4}$$

$x_{k+1} \in S$

$\sum_{k=0}^{\infty} \text{reepred}_k = \infty$ may be difficult/impossible to verify

Won't need to do this - will construct algorithms that determine steps such that $\text{ared}_k \geq t \cdot \text{pred}_k \geq 0$ holds for each k and $\sum \text{reepred}_k$ is automatically satisfied near points of interest.

Backtracking

Idea: For each k , initial trial step is

$$s_k = s_k^H = -F'(x_k)^{-1} F(x_k)$$

If $\text{ared}_k \geq t \cdot \text{pred}_k \geq 0$, accept s_k

If not, shorten s_k ($s_k = \lambda s_k^H$, $0 < \lambda < 1$) and repeat test

see: if $s = \lambda s^H$, $0 < \lambda < 1$ then

$$\|F + F's\| = \|F + F'\lambda s^H\|$$

$$= \|(1-\lambda)F + \lambda(F + F's^H)\|$$

$$= \|(1-\lambda)F\| = (1-\lambda)\|F\|$$

$$\Rightarrow \text{pred} = \|F\| - \|F + F's\|$$

$$= \lambda \|F\|$$

also

$$\begin{aligned}
 \text{ared} &= \|F\| - \|F(x+s)\| \\
 &= \|F\| - \|F(x+s) - F - F's + F + F's\| \\
 &\geq \underbrace{\|F\| - \|F + F's\|}_{\substack{\text{pred} \\ \| \\ \lambda \|F\|}} - \|F(x+s) - F - F's\| \quad \wedge \\
 &\qquad \qquad \qquad \omega(\|\lambda s^H\|) \|\lambda s^H\| \leq \\
 &\qquad \qquad \qquad \leq \omega(\|\lambda s^H\|) \lambda \underbrace{\|F'(x)^{-1}\| \|F(x)\|}_{\text{pred}} \\
 &\geq \lambda \|F\| - \omega(\lambda \|s^H\|) \|F'(x)^{-1}\| \lambda \|F(x)\| \\
 &= \left(1 - \omega(\lambda \|s^H\|) \|F'(x)^{-1}\|\right) \text{pred} \\
 &\geq t \cdot \text{pred} \quad , \text{ for sufficiently small } \lambda
 \end{aligned}$$

So a sufficiently short step $s = \lambda s^H$ will satisfy

$$\text{ared} \geq t \cdot \text{pred} \geq 0$$

Newton's method with Backtracking
(this is "safeguarded backtracking")

Given $t \in (0,1)$ and $0 < \theta_{\min} < \theta_{\max} < 1$

Until termination do

 solve $F'(x) \cdot s = -F(x)$ for s^H

 while $\text{ared} < t \text{pred}$ do

 choose $\theta \in [\theta_{\min}, \theta_{\max}]$

 update $s \leftarrow \theta s$

 update $x \leftarrow x + s$

Note: after p passes through while-loop with $\theta_1, \dots, \theta_p$ the θ -values have $s = \lambda_p s^H$
 $\lambda_p = \theta_1 \dots \theta_p$

Have seen

$$\text{aved} \gg (1 - \omega(\lambda_p \|S^H\|) \|F'(x)^{-1}\|) \text{pred}$$

since

$$\theta_{\min}^p \leq \lambda_p \leq \theta_{\max}^p$$

will have $\text{aved} \gg t \text{pred}$ for some finite p

Theorem: Suppose the algorithm produces $\{x_k\}$. If x_* is a limit point of $\{x_k\}$ such that $F'(x_*)$ is nonsingular, then $F(x_*) = 0$ and $x_k \rightarrow x_*$ and $s_k = x_{k+1} - x_k = -F'(x_k)^{-1} F(x_k)$ for all sufficiently large k

Note: - Don't need $\sum \text{velpred}_k = \infty$

- Free Newton steps are ultimately taken near x_* such that $F'(x_*)$ nonsingular

Possibilities: Exactly one holds

(1) $\|x_k\| \rightarrow \infty$

(2) $\{x_k\}$ has limit points and F' is singular at each

(3) $x_k \rightarrow x_*$ such that $F(x_*) = 0$, $F'(x_*)$ nonsingular and convergence is that of Newton's method (superlinear or quadratic)

proof: Have $s_k = \lambda_k S_k^H$ for $0 < \lambda_k < 1$

$$\text{and } \text{pred}_k = \lambda_k \|F_k\|$$

$$\Rightarrow \text{velpred}_k = \frac{\text{pred}}{\|F_k\|} = \lambda_k$$

$$\text{To show: } \sum_{k=1}^{\infty} \lambda_k = \infty$$

Let $\delta > 0$ be sufficiently small that if $\|x - x_*$ $\| \leq \delta$ then

$$i) \|F'(x)^{-1}\| \leq 2K \quad \text{where } K = \|F'(x_*)^{-1}\|$$

$$ii) \omega(2\delta) \equiv \max_{\|d\| \leq \delta} \|F'(x+d) - F'(x)\| \leq \frac{1-t}{2K}$$

(OK since F' is uniformly continuous on δ -balls about x_*)

suppose $\|x_k - x_*\| < \delta$ and $\|x_{k+s} - x_*\| < \delta$,

$$s = \lambda s_k^H, \quad 0 < \lambda < 1$$

$$\text{ared} = \|F(x_k)\| - \|F(x_{k+s})\|$$

$$\begin{aligned} &\geq \|F_k\| - \|F_k + F_k' s\| - \underbrace{\|F(x_{k+s}) - F_k - F_k' s\|}_{\wedge} \\ &\qquad\qquad\qquad \omega(\lambda \|s_k^H\|) \|F_k'^{-1}\| \lambda \|F_k\|_{\text{pred}} \end{aligned}$$

$$\geq \left(1 - \|F_k'^{-1}\| \omega(\|s\|)\right) \text{pred}$$

$$\geq \left(1 - 2K \omega(2\delta)\right) \text{pred}$$

$$\geq \left(1 - 2K \frac{1-t}{2K}\right) \text{pred} = t \text{pred}$$

Also,

$$\|x_k - x_*\| \leq \delta \Rightarrow \|s_k^H\| \leq \|-F'(x_k)^{-1} F(x_k)\|$$

$$\leq 2K \|F(x_k)\|$$

$$\leq 2KM$$

where $\|F(x)\| \leq M$ whenever $\|x - x_*\| \leq \delta$

Now if $\|x_k - x_*$ $\| \leq \frac{\delta}{2}$ and $s = \lambda s_k^H$ with

$0 \leq \lambda \leq \lambda_* \equiv \min \left\{ 1, \frac{\delta}{4kM} \right\}$ then

$$\|s\| \leq \lambda_* \|s_k^H\| \leq \lambda_* 2kM \leq \frac{\delta}{2}$$

$$\Rightarrow \|x_k + s - x_*\| \leq \delta \text{ and } \text{ared} \geq t_{\text{pred}}$$

So if $\|x_k - x_*\| < \frac{\delta}{2}$, backtracking terminates with $s_k = \lambda_k s_k^H$ after at most p steps where $\theta_{\max}^p \leq \lambda_*$.
Then have $\theta_{\min}^p \leq \lambda_k$ and so

$$\sum_k \text{velpred}_k = \sum_k \lambda_k \geq \sum_{k \in I} \lambda_k$$

$$I = \left\{ k : \|x_k - x_*\| \leq \frac{\delta}{2} \right\}$$

$$\geq \sum_{k \in I} \theta_{\min}^p = \infty$$

Since $\|x_k - x_*\| \leq \frac{\delta}{2}$ for infinitely many k

Results follows from theorem

For k sufficiently large, both x_k and $x_k + s^H$ are within δ of x_* . Then $\text{ared} \geq t_{\text{pred}}$ and $x_{k+1} = x_k + s_k^H$