

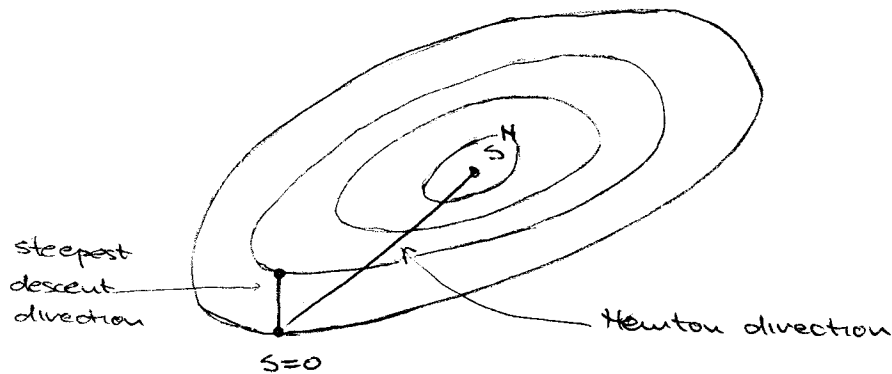
More on globalization

- Backtracking from Newton step may have shortcomings
- If F is "badly behaved"
 - many step length reductions may be required
 - reduction in $\|F\|$ may be small relative to that given by other steps of same length (but different direction)

Badly Behaved? May be

- highly nonlinear or may have ill-conditioned Jacobian (write $J(x)$ for $F'(x)$)

picture ($\|\cdot\| = \|\cdot\|_2$) of level sets $\|F + Js\| = c$ in S -space



Eccentricity is directly related to $K(J) = \|J\| \|J^{-1}\|$

- if J is very badly conditioned, Newton direction and steepest descent direction may be nearly orthogonal
- can show for an unfortunate choice of F and J can have

$$\left(\frac{\nabla f}{\|\nabla f\|_2} \right) \left(\frac{S^H}{\|S^H\|_2} \right) \approx \frac{1}{K_2(J)} \quad \text{where } f = \frac{1}{2} \|F\|^2$$

Also, in this case, J is "nearly singular" $\Rightarrow S^H$ is likely to be long

$\Rightarrow$ many step length reductions may be required to deal with nonlinearity

$\Rightarrow$ short step λS^H

Since S^H is nearly orthogonal to steepest descent direction, an equally short step in (or near) the steepest descent direction might give much greater reduction in $\|F\|$

- Trust region methods

Idea: At each step, have a trust region

$$H_\delta(x) = \{y: \|y-x\| \leq \delta\}$$

about current x defined by trust region radius $\delta > 0$ such that we "trust"

$$F(x) + J(x) \cdot S \approx F(x+S)$$

$$\text{for } \|S\| < \delta$$

Then choose $S = \arg \min_{\|w\| \leq \delta} \|F + Jw\|$

(Ideally)

finding exact
solution is real
challenge

Test S

unacceptable $\Rightarrow$ reduce δ , find new S

$$S = \arg \min_{\|w\| \leq \delta} \|F + Jw\|, \text{ repeat the test}$$

acceptable $\Rightarrow$ update $x \leftarrow x + S$, consider adjusting δ for next step

- General TR method is a long way from a practical algorithm
- Biggest issue: finding satisfactory $S \approx \arg \min_{\|w\| \leq \delta} \|F + Jw\|$

- Develop some properties of ideal S

Proposition: If J is nonsingular and $S = \arg \min_{\|w\| \leq \delta} \|F + Jw\|$

then:

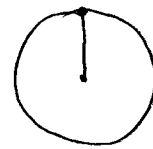
i) $\|s^H\| < \delta \Rightarrow S = s^H$

ii) $\|s^H\| > \delta \Rightarrow \|S\| = \delta$

i)



ii)



some point on the boundary but not necessarily in Newton direction

proof: i) clear since $\|F + Js^H\|$

ii) if $\|s\| < \delta$ can find ϵ such that

$$I_\epsilon = -\epsilon [F + Js]$$

where $0 < \epsilon < 1$ so small that $\|s + \epsilon\| < \delta$. Then

$$\begin{aligned} \|F + J(s + \epsilon)\| &= \|F + Js - \epsilon [F + Js]\| \\ &= \|(1 - \epsilon)(F + Js)\| \\ &= (1 - \epsilon) \|F + Js\| < \|F + Js\| \end{aligned}$$

contradiction!

In general TR method can have breakdown in white-loop if J is singular

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

At $x=0$, $J=0 \Rightarrow \|F + Js\| = \|F\|$

for every $s \Rightarrow \text{pred} = 0$ (acceptable)

But $\text{ared} = F(0) - F(0+s) = 1 - (1+s^2) = -s^2 < 0$

Can never have $\text{ared} > t \text{pred}$ for $s \neq 0$

But don't have breakdown if J nonsingular

Proposition: Suppose F is continuously differentiable near x and $J(x)$ nonsingular. Then for

$$S = \arg \min_{\|w\| \leq \delta} \|F + Jw\|$$

have

$$\text{ared} \geq [1 - \omega(\delta) \|J(x)^{-1}\|] \text{pred}$$

where

$$\omega(\delta) = \max_{\|u\| \leq \delta} \|J(x+u) - J(x)\|$$

So since $\omega(\delta) \rightarrow 0$ as $\delta \rightarrow 0$ we can have $\text{ared} \geq \text{pred}$ when δ is so small that

$$1 - \omega(\delta) \|J(x)^{-1}\| \geq t$$

i.e.,

$$\omega(\delta) \leq \frac{1-t}{\|J^{-1}\|}$$

proof:

$$\begin{aligned} \text{ared} &= \|F\| - \|F(x+s)\| \\ &= \|F\| - \|F(x+s) - F - Js + F + Js\| \\ &\geq \underbrace{\|F\| - \|F + Js\|}_{\text{pred}} - \|F(x+s) - F - Js\| \end{aligned}$$

If $\|F(x+s) - F - Js\| \leq \omega(\delta) \|J^{-1}\| \cdot \text{pred}$, then done

(assume $\|s^H\| > \delta \Rightarrow \|s\| = \delta$)

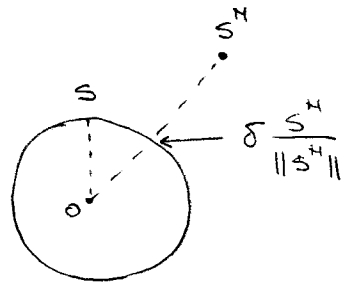
$$\begin{aligned} \text{Handy Lemma (VI)} \Rightarrow \|F(x+s) - F - Js\| &\leq \omega(\|s\|) \|s\| \\ &= \omega(\delta) \cdot \delta \end{aligned}$$

so need only to show $\delta \leq \|J^{-1}\| \text{pred}$

Have $\text{pred} = \|F\| - \|F + JS\|$

subtract something bigger
so that you can get
something smaller

$$\geq \|F\| - \underbrace{\left\| F + J \left(\delta \frac{s^H}{\|s^H\|} \right) \right\|}_{\| \quad \|}$$



$$\left(1 - \frac{\delta}{\|s^H\|} \right) F + \frac{\delta}{\|s^H\|} \underbrace{(F + JS^H)}_{\| \quad \|}$$

$$= \|F\| - \left(1 - \frac{\delta}{\|s^H\|} \right) \|F\|$$

$$= \frac{\delta}{\|s^H\|} \|F\|, \quad \|s^H\| = -J^{-1} F$$

note $\frac{\delta}{\|s^H\|} \leq 1$

$$\geq \frac{\delta}{\|J^{-1}\| \|F\|} \|F\| = \frac{\delta}{\|J^{-1}\|}$$

Nice: 1) $\text{pred} = \|F\| - \|F + JS\| \geq 0$

2) J nonsingular and $F \neq 0 \Rightarrow \text{pred} > 0$

if $s = s^H$, $\text{pred} = \|F\| > 0$

if $\|s\| = \delta < \|s^H\|$, then

$$\text{pred} \geq \|F\| - \left\| F + J \left(\delta \frac{s^H}{\|s^H\|} \right) \right\|$$

$$= \|F\| - \left(1 - \frac{\delta}{\|s^H\|} \right) \|F\|$$

$$= \frac{\delta}{\|s^H\|} \|F\| > 0$$

3) if $\|v\| = \langle v, v \rangle^{1/2}$, then s is a descent direction for $\|F\|$ at x

i.e. $\langle \nabla \|F\|, s \rangle < 0$

proof: $f(x) = \frac{1}{2} \|F(x)\|^2$

$$\left. \frac{d}{d\theta} f(x + \theta s) \right|_{\theta=0} = \left. \langle F(x + \theta s), J(x + \theta) s \rangle \right|_{\theta=0}$$

$$= \langle F, Js \rangle$$

$$= \langle F, F+JS \rangle - \langle F, F \rangle$$

$$\leq \|F\| \|F+JS\| - \|F\|^2$$

$$= -\|F\| \text{pred} < 0$$

Def: x is a stationary point of $\|F\|$ if

$$\|F+JS\| \geq \|F\| \text{ for every step}$$

(traditional def. of stationary point, point where $\nabla f = 0$)

Note: Suppose $\|v\| = \langle v, v \rangle^{1/2}$

$$\text{Then } \|F\| \leq \|F+JS\|$$

$$\Rightarrow \|F\|^2 \leq \|F+JS\|^2 = \|F\|^2 + 2\langle F, JS \rangle + \|JS\|^2$$

this holds for all s only if $\langle F, JS \rangle = 0$ for all s

$$\Rightarrow \langle J^* F, s \rangle = 0 \text{ for all } s$$

↑ adjoint operation

$$\Rightarrow J^* F = 0$$

$$\text{if } \|\cdot\| = \|\cdot\|_2 \text{ then } 0 = J^* F = J^T F = \nabla \left(\frac{1}{2} \|F\|^2 \right)$$

Theorem: Suppose $\{x_k\}$ is produced by General Trust region method.

Then every limit point is a stationary point of $\|F\|$.

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

$$s_k = x_{k+1} - x_k = -J(x_k)^{-1} F(x_k) \text{ for all sufficiently large } k$$

(Th. 4.4. of Eisenstat - Walker 1994)

Backtracking
doesn't
guarantee
this

Examples for which Newton + Backtracking iterates converge to non-stationary points

- For a practical algorithm, specify many details as with backtracking

choose t small (e.g. 10^{-4})

so that any step that reduces the norm is acceptable

$$\theta_{\min} = 0.1 \quad \theta_{\max} = 0.5$$

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

- First if $\|s\| < \delta$ then reduce δ to $\|s\|$
- Once $\|s\| = \delta$, choose θ as before to minimize an interpolating quadratic or cubic

(requires $\|v\| = \langle v, v \rangle^{1/2}$ inner-product norm)

Details similar to backtracking see Dennis & Schnaebel

- Final step: possibly adjusting δ for next step

As in Dennis & Schnaebel:

i) $\delta \leftarrow 2\delta$ if good agreement

ii) $\delta \leftarrow \delta$ if so-so agreement

iii) $\delta \leftarrow \delta/2$ if poor agreement

for i) test $\text{ared} \geq u \cdot \text{pred}$ for $t \leq u \leq 1$

ii) test $v \cdot \text{pred} \leq \text{ared} < u \cdot \text{pred}$

for $t \leq v \leq u < 1$

iii) test $\text{ared} < v \cdot \text{pred}$

Dennis - Schnaebel

$$u = 0.75$$

$$v = 0.1$$

$$\text{with } t = 10^{-4}$$

Computing the TR step

$$s = \arg \min_{\|w\| \leq \delta} \|F + Jw\|$$

- computing s exactly is not practical
- want to approximate adequately at reasonable cost
- work toward a more refined characterization of s (assume $\|\cdot\| = \|\cdot\|_2$ throughout, for convenience only any inner-product norm will work)

Lemma: If J is nonsingular then

$$s = s(\mu) \equiv -[J^T J + \mu I]^{-1} J^T F$$

for a unique $\mu \geq 0$ as follows

$$\|s^H\| \leq \delta \Rightarrow \mu = 0$$

$$\|s^H\| > \delta \Rightarrow \mu > 0 \text{ and uniquely determined by } \|s(\mu)\| = \delta$$

proof: If $\|s^H\| \leq \delta$ then $s = s^H = -J^{-1}F = -[J^T J]^{-1} J^T F$
and $\mu = 0$

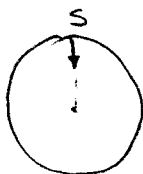
suppose $\|s^H\| > \delta$

$$\text{set } \mathcal{E}(s) = \frac{1}{2} \|F + Js\|^2$$

$$\text{have } \nabla \mathcal{E}(s) = J^T(F + Js) = J^T F + J^T J s \neq 0$$

since $s \neq s^H =$ unique point where $\nabla \mathcal{E} = 0$

$\left(\begin{array}{l} s - \text{step on the boundary of trust region} \\ \text{but not in a Newton direction,} \\ \text{so grad} \neq 0 \end{array} \right)$



since s minimizes $\mathcal{E}$ over TR , must have $\nabla \mathcal{E}(s)$ pointing inside TR and orthogonal to the boundary

$$\Rightarrow \nabla \mathcal{E}(s) = -\mu s \text{ for some } \mu > 0$$

$$\text{Then } -\mu s = J^T F + J^T J s$$

$$(J^T J + \mu I) s = -J^T F$$

so have existence of $\mu > 0$, uniqueness follows from Proposition and Corollary below

Proposition:

$$i) s(\mu) \text{ is differentiable and } s'(\mu) = -[J^T J + \mu I]^{-1} s(\mu)$$

$$ii) \varphi(\mu) \equiv \|s(\mu)\|^2 \text{ is differentiable and}$$

$$\begin{aligned}
 (*) \quad \varphi'(\mu) &= 2s(\mu)^T s'(\mu) \\
 &= -2s(\mu)^T \left[J^T J + \mu I \right]^{-1} s(\mu)
 \end{aligned}$$

Convexity: $\|s(\mu)\|$ is monotone strictly in μ with $\|s(0)\| = \|s^H\|$ and $\lim_{\mu \rightarrow \infty} \|s(\mu)\| = 0$

proof of convexity:

(*) $\Rightarrow \varphi'(\mu) < 0$ since $[J^T J + \mu I]$ is SPD

$\Rightarrow \|s(\mu)\|$ strictly monotone decreasing

Also know: $s(0) = s^H$

Finally: $s(\mu) = -[J^T J + \mu I]^{-1} J^T F \approx -\frac{1}{\mu} J^T F$
for range $\mu \rightarrow 0$ as $\mu \rightarrow \infty$

proof of proposition:

suppose $A(\mu)$ is some differentiable function of μ and that $A(\mu)$ is invertible. Then $A(\mu + \Delta)$ is invertible for small Δ and

$$\begin{aligned}
 \frac{1}{\Delta} [A(\mu + \Delta)^{-1} - A(\mu)^{-1}] &= A(\mu + \Delta)^{-1} \left[\frac{A(\mu) - A(\mu + \Delta)}{\Delta} \right] A(\mu)^{-1} \\
 &\longrightarrow -A(\mu)^{-1} A'(\mu) A(\mu)^{-1}
 \end{aligned}$$

$$\Rightarrow \frac{d}{d\mu} A(\mu)^{-1} = -A(\mu)^{-1} A'(\mu) A(\mu)^{-1}$$

$$\text{Have } s(\mu) = - \underbrace{[J^T J + \mu I]^{-1}}_{\substack{A(\mu) \\ A'(\mu) I}} J^T F$$

$$\Rightarrow s'(\mu) = - \underbrace{[J^T J + \mu I]^{-1} I [J^T J + \mu I]^{-1}}_{-s(\mu)} J^T F = -[J^T J + \mu I]^{-1} s(\mu)$$

$$\text{For } \varphi(\mu) = \|s(\mu)\|^2 = s(\mu)^T s(\mu)$$

$$\Rightarrow \varphi'(\mu) = 2s(\mu)^T s'(\mu) = -2s(\mu)^T [J^T J + \mu I]^{-1} s(\mu)$$