

SOLUTIONS TO MIDTERM SPRING 2011

Q1. (a) $\{x_k\}$ is such that

$$\begin{aligned} x_{k+1} &= x_k - \frac{f(x_k)}{f'(x_k)} \\ &= x_k - \frac{(x_k^2 - 1/5)}{2x_k} = \frac{x_k^2 + 1/5}{2x_k} \end{aligned}$$

(b) Order of convergence

$$\begin{aligned} \lim_{k \rightarrow \infty} \frac{\|x_{k+1} - 1/\sqrt{5}\|}{\|x_k - 1/\sqrt{5}\|^2} &= \lim_{k \rightarrow \infty} \frac{\left| \frac{x_k^2 + 1/5}{2x_k} - 1/\sqrt{5} \right|}{|x_k - 1/\sqrt{5}|^2} \\ &= \lim_{k \rightarrow \infty} \frac{|x_k^2 - 2/\sqrt{5}x_k + 1/5|}{|2x_k| |x_k - 1/\sqrt{5}|^2} \\ &= \lim_{k \rightarrow \infty} \frac{1}{|2x_k|} = \frac{\sqrt{5}}{2} \end{aligned}$$

Therefore the order of convergence is quadratic.

(c) $\left. \frac{d}{d\alpha} |f(x_k + \alpha p_k)| \right|_{\alpha=0} = \left. \frac{d}{d\alpha} (\sqrt{f^2(x_k + \alpha p_k)}) \right|_{\alpha=0}$ ①

$$= \left. \left(\frac{1}{2\sqrt{f^2(x_k + \alpha p_k)}} \cdot 2f(x_k + \alpha p_k) p_k \right) \right|_{\alpha=0}$$

(since $p_k = -\frac{f(x_k)}{f'(x_k)}$) $= \frac{f(x_k)}{|f(x_k)|} f'(x_k) \left(\frac{-f(x_k)}{f'(x_k)} \right) = -\frac{f^2(x_k)}{|f(x_k)|} < 0$

Q2. (a)

$$\nabla f(x_*) = 0 \quad \text{and} \quad \nabla^2 f(x_*) \succ 0$$

\implies
 x_* is a local minimizer

(b)

$$f(x) = \frac{1}{2} x^T \begin{bmatrix} 6 & -1 \\ -1 & 2 \end{bmatrix} x + x^T \begin{bmatrix} 7 \\ -3 \end{bmatrix} + 4$$

where $x = [x_1 \ x_2]^T$.

$$\nabla f(x) = \begin{bmatrix} 6 & -1 \\ -1 & 2 \end{bmatrix} x + \begin{bmatrix} 7 \\ -3 \end{bmatrix}$$

Stationary point x_* satisfies:

$$\nabla f(x_*) = 0 \iff \begin{bmatrix} 6 & -1 \\ -1 & 2 \end{bmatrix} x_* = \begin{bmatrix} -7 \\ 3 \end{bmatrix}$$

$$\iff x_* = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$$

Hessian is positive definite, since it has positive eigenvalues.

$$\begin{aligned} \det \left(\begin{bmatrix} 6-\lambda & -1 \\ -1 & 2-\lambda \end{bmatrix} \right) = 0 &\iff (6-\lambda)(2-\lambda) - 1 = 0 \\ &\iff \lambda^2 - 8\lambda + 11 = 0 \\ &\iff \lambda_{1,2} = \frac{8 \pm \sqrt{64-44}}{2} > 0 \end{aligned}$$

By second order sufficient conditions
(since $\nabla f(x_*) = 0$ and $\nabla^2 f(x_*) \succ 0$)

$x_* = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$ is a local minimizer.

(2)

(c) Let x_* be such that $\nabla f(x_*) = 0$.
Then by Taylor's thm

$$f(x) = f(x_*) + \underbrace{\nabla f(x_*)^T (x - x_*)}_0 + \frac{1}{2} \underbrace{(x - x_*)^T \nabla^2 f(\tilde{x}) (x - x_*)}_{q(x)}$$

Since $\nabla^2 f(\tilde{x}) \underset{(\text{= } A)}{\geq} 0$, $q(x) \geq 0$ for all x .

Consequently for all x

$$f(x) - f(x_*) = \frac{q(x)}{2} \geq 0$$

$$\implies f(x) \geq f(x_*)$$

meaning x_* is a global minimizer. \square

Q3.

$$H_{k+1} = H_k + \sigma u u^T$$

$$(*) \quad \underbrace{H_{k+1} y_k}_{s_k} = H_k y_k + \sigma u \underbrace{(u^T y_k)}_{\text{scalar}}$$

$$\implies \boxed{u = \alpha (s_k - H_k y_k)} \text{ for some scalar } \alpha$$

Plug this again in (*)

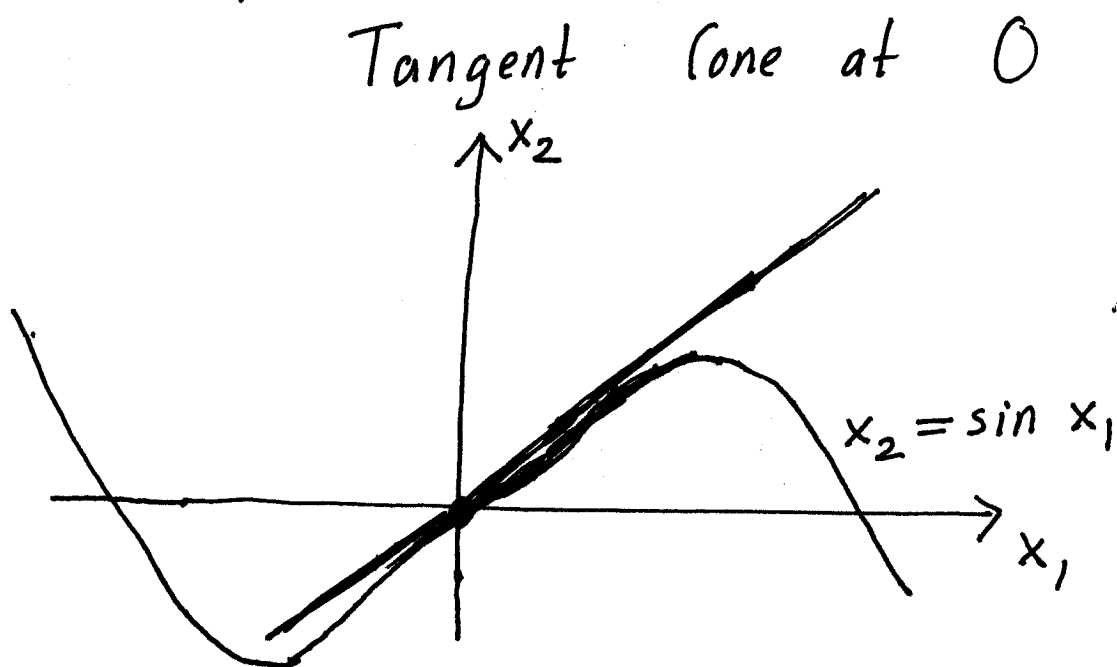
$$s_k - H_k y_k = \sigma \alpha^2 (s_k - H_k y_k) ((s_k - H_k y_k)^T y_k)$$

$$\implies \boxed{\sigma \alpha^2 = 1 / ((s_k - H_k y_k)^T y_k)}$$

Consequently

$$\begin{aligned} H_{k+1} &= H_k + \sigma \alpha^2 (s_k - H_k y_k) (s_k - H_k y_k)^T \\ &= H_k + \frac{(s_k - H_k y_k) (s_k - H_k y_k)^T}{((s_k - H_k y_k)^T y_k)} \end{aligned}$$

Q4. (a)



Consists of the set of tangent vectors to feasible paths at 0 equivalently all vectors tangent to $x_2 = \sin x_1$ at 0.

$$\begin{aligned} T^0(0) &= \left\{ \alpha \begin{bmatrix} x_1 \\ \sin x_1 \end{bmatrix} \Big|_{x_1=0} : \alpha \in \mathbb{R} \right\} \\ &= \left\{ \alpha \begin{bmatrix} 1 \\ \cos x_1 \end{bmatrix} \Big|_{x_1=0} : \alpha \in \mathbb{R} \right\} \\ &= \text{span} \left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right\} \end{aligned}$$

Null space of the constraint Jacobian at 0

$$J(x) = [-\cos x_1 \quad 1]$$

$$\Rightarrow J(0) = [-1 \quad 1]$$

$$\begin{aligned}\Rightarrow \text{Null}(J(0)) &= \{p \in \mathbb{R}^2 : J(0)p = 0\} \\ &= \{\alpha \begin{bmatrix} 1 \\ 1 \end{bmatrix} : \alpha \in \mathbb{R}\} \\ &= \text{span} \left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right\}\end{aligned}$$

$$T^0(0) = \text{Null}(J(0))$$

CONSTRAINT
QUALIFICATION
HOLDS

(b)

First order optimality condition

~~For~~ If $x \in \mathbb{R}^2$ ~~is~~ is a local minimizer, then

$$(i) \nabla f(x) = J(x)^T \lambda \Rightarrow \begin{bmatrix} 6x_1 \\ 2(x_2 - 1) \end{bmatrix} = \begin{bmatrix} -\cos x_1 \\ 1 \end{bmatrix} \text{ for some } \lambda$$

$$(ii) c(x) = x_2 - \sin(x_1) = 0$$

At $\bar{x} = 0$

$$(ii) c(\bar{x}) = 0 - \sin(0) = 0$$

but

$$(i) \nabla f(\bar{x}) = \begin{bmatrix} 0 \\ -2 \end{bmatrix} \neq \overbrace{\begin{bmatrix} -\cos 0 \\ 1 \end{bmatrix}}^{J(\bar{x})^T} \lambda = \begin{bmatrix} -1 \\ 1 \end{bmatrix} \lambda \text{ for all } \lambda$$

Consequently $\bar{x} = (0, 0)$ is not a local minimizer. (by first order optimality condition)