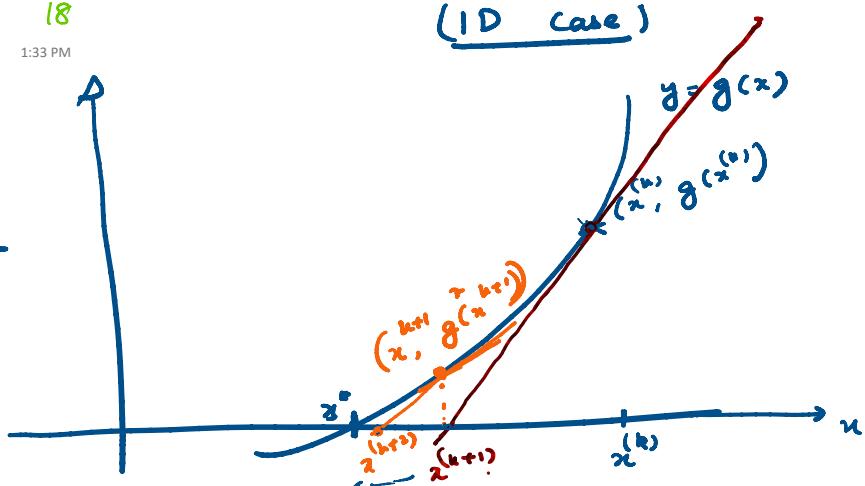


## Lecture 18

Thursday, 10 March 2022

1:33 PM

### Newton's



$$\begin{aligned} \min f(x) \\ \frac{f'(x^*)}{\frac{f'(x^*)}{g(x^*)}} &= 0 \\ \downarrow \\ g(x^*) &= 0 \end{aligned}$$

$x = x^{(k+1)}$

$$\checkmark y - g(x^{(k)}) = g'(x^{(k)}) (x - x^{(k)})$$

$$0 \approx y = g(x^{(k)}) + g'(x^{(k)}) (x - x^{(k)})$$

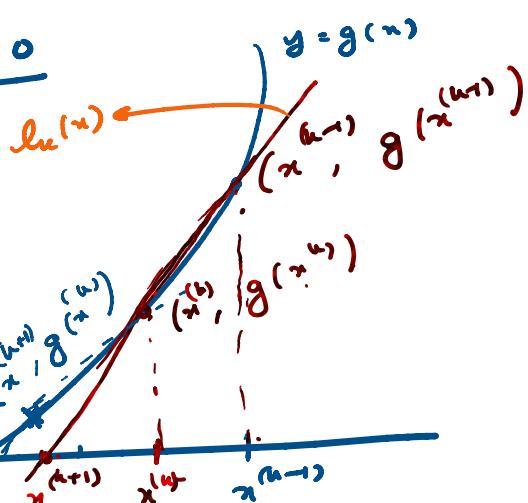
$$x^{(k+1)} = x^{(k)} - \underbrace{\left(g'(x^{(k)})\right)^{-1}}_{x^{(k+1)} = x^{(k)} - (f''(x^{(k)}))^{-1} f'(x^{(k)})} g(x^{(k)})$$

multivariate

$$\begin{aligned} \underline{x}^{(k+1)} &= \underline{x}^{(k)} - [\nabla g(\underline{x}^{(k)})]^{-1} g(\underline{x}^{(k)}) \\ \underline{x}^{(k+1)} &= \underline{x}^{(k)} - [D^2 f(\underline{x}^{(k)})]^{-1} \nabla f(\underline{x}^{(k)}) \end{aligned}$$

Solve:

$$g(x) = 0$$



$$x^{(k+1)} = x^{(k)} - \left[ \frac{x^{(k)} - x^{(k-1)}}{g(x^{(k)}) - g(x^{(k-1)})} \right] g(x^{(k)}).$$

l\_{k+1}(x) passes ...

$$g(\underline{x}^{(k)}) - g(\underline{x}^{(k)})$$

$$l_k(\underline{x}) = g(\underline{x}^{(k)}) + B_k (\underline{x} - \underline{x}^{(k)})$$

$$\rightarrow l_{k+1}(\underline{x}) = g(\underline{x}^{(k+1)}) + B_{k+1} (\underline{x} - \underline{x}^{(k+1)})$$

$$\underline{l}_{k+1}(\underline{x}^{(k)}) = g(\underline{x}^{(k+1)}) + B_{k+1} (\underline{x}^{(k)} - \underline{x}^{(k+1)})$$

$$\underline{g}(\underline{x}^{(k)}) - g(\underline{x}^{(k+1)}) = B_{k+1} (\underline{x}^{(k)} - \underline{x}^{(k+1)})$$

$$\Delta g(\underline{x}^{(k)}) = B_{k+1} \Delta \underline{x}^{(k)}$$

\$l\_k(\underline{x})\$ passes thru' \$(\underline{x}^{(k)}, g(\underline{x}^{(k)}))\$ and \$(\underline{x}^{(k)}, g(\underline{x}^{(k)}))\$

\$l\_{k+1}(\underline{x})\$ passes thru' \$(\underline{x}^{(k+1)}, g(\underline{x}^{(k+1)}))\$ and \$(\underline{x}^{(k)}, g(\underline{x}^{(k)}))\$

Undetermined System:

$$(B_{k+1})_{n \times n} \Delta \underline{x}^{(k)} = \Delta g(\underline{x}^{(k)})$$

Secant Condition

Lecture 13  $\rightarrow (B_{k+1} = H_{k+1}^{-1})$ .

$$\nabla f(\underline{x}^{(k+1)}) - \nabla f(\underline{x}^{(k)}) = B_{k+1} (\underline{x}^{(k+1)} - \underline{x}^{(k)})$$

$$[-]_{n \times 1} = [=]_{n \times n} [.]_{n \times 1}$$

$n \times n$   
 $1 + 2 + \dots + n = \frac{n(n+1)}{2}$

We want

$$\min \|B_{k+1} - B_k\|$$

s.t.

$$B_{k+1}^T = B_{k+1}$$

$$B_{k+1} \Delta \underline{x}^{(k)} = \Delta g^{(k)}$$

$$\min \|B_{k+1}^{-1} - B_k^{-1}\|$$

s.t.

$$\left\{ \begin{array}{l} (B_{k+1}^{-1})^T = B_{k+1}^{-1} \\ \Delta \underline{x}^{(k)} = B_{k+1}^{-1} \Delta g^{(k)} \end{array} \right.$$

(Many notations)

Quasi-Newton

- $d^{(k)} = -H_k^{-1} g^{(k)}$  (Fessian inverse in Newton)
- $d^k = \arg \min_{d \geq 0} f(\underline{x}^{(k)} + \alpha d^{(k)})$
- $\underline{x}^{(k+1)} = \underline{x}^{(k)} + d^{(k)}$

Motivation for Lagrange Method

$\underbrace{1}_{\text{constant}} \dots \underbrace{T}_{+ B} \underbrace{V V^T}$

Construct updates

Many options

$$B_{k+1} = B_k + \alpha \underline{u} \underline{u}^T + \beta \underline{v} \underline{v}^T$$

(Rank - two update)

(BFGS Method)

Broyden-Fletcher-Goldfarb-Shanno  
(1970)

$$\underline{u} = \begin{bmatrix} u_1 \\ \vdots \\ u_n \end{bmatrix} \quad \underline{v} = \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix}$$

$\underline{u}, \underline{v}$  l.i.

Rank 1

$$\underline{u} \underline{u}^T = \begin{bmatrix} u_1^2 & u_1 u_2 & \dots & u_1 u_n \\ u_2 u_1 & u_2^2 & \dots & u_2 u_n \\ \vdots & \vdots & \ddots & \vdots \\ u_n u_1 & u_n u_2 & \dots & u_n^2 \end{bmatrix}$$

$$B_{k+1} \Delta \underline{x}^{(k)} = \Delta \underline{g}^{(k)}$$

$$B_{k+1}^T = B_{k+1}$$

$$B_{k+1} \Delta \underline{x}^{(k)} = B_k \Delta \underline{x}^{(k)} + \alpha \underline{u} \underline{u}^T \Delta \underline{x}^{(k)} + \beta \underline{v} \underline{v}^T \Delta \underline{x}^{(k)} = \Delta \underline{g}^{(k)}$$

Choose

$$\begin{aligned} \sqrt{\underline{u}} &= \Delta \underline{g}^{(k)} \\ \sqrt{\underline{v}} &= B_k \Delta \underline{x}^{(k)} \end{aligned}$$

Qn: Are they l.i.? Exercise.  
(Prove?)

$$B_k \Delta \underline{x}^{(k)} + \alpha \underbrace{\Delta \underline{g}^{(k)} (\Delta \underline{g}^{(k)})^T \Delta \underline{x}^{(k)}}_{\text{Constant}} + \beta \underbrace{\frac{B_k \Delta \underline{x}^{(k)} (\Delta \underline{x}^{(k)})^T B_k \Delta \underline{x}^{(k)}}{\Delta \underline{g}^{(k)}}}_{\text{Constant}} = \Delta \underline{g}^{(k)}$$

$$\Rightarrow \underbrace{\Delta \underline{g}^{(k)}}_{\underline{u}} \left[ 1 - \alpha (\Delta \underline{g}^{(k)})^T \Delta \underline{x}^{(k)} \right] = \underbrace{B_k \Delta \underline{x}^{(k)}}_{\sqrt{\underline{v}}} \left[ 1 + \beta (\Delta \underline{x}^{(k)})^T B_k \Delta \underline{x}^{(k)} \right]$$

$$\Rightarrow \alpha = \frac{1}{(\Delta \underline{g}^{(k)})^T \Delta \underline{x}^{(k)}} \quad \beta = -\frac{1}{(\Delta \underline{x}^{(k)})^T B_k \Delta \underline{x}^{(k)}}$$

$$B_{k+1} = B_k + \frac{(\Delta \underline{g}^{(k)}) (\Delta \underline{g}^{(k)})^T}{(\Delta \underline{g}^{(k)})^T \Delta \underline{x}^{(k)}} - \frac{B_k \Delta \underline{x}^{(k)} (\Delta \underline{x}^{(k)})^T B_k}{(\Delta \underline{x}^{(k)})^T B_k \Delta \underline{x}^{(k)}}$$

Rank - two update