$$y = f(\mathbf{x}) = f(x_1, \cdots, x_N)$$

$$\mathbf{x}^* = [x_1^2, \cdots, x_N^*]^T \in \mathbb{R}^N$$

$$f(\mathbf{x})$$
, $f(\mathbf{x})$,

$$f(\mathbf{x}^*) = \min_{\mathbf{x}} f(\mathbf{x})$$

$$y = f(\mathbf{x}, \mathbf{a})$$

$$\mathcal{D} = \{ (\mathbf{x}_n, y_n), \quad n = 1, \dots, N \}$$

$$\mathbf{a} = [a_1, \cdots, a_M]^T$$

$$y = f(x) = ax + b$$

$$o(\mathbf{a}) = \sum_{n=1}^{N} [y_n - f((\mathbf{x}_n, \mathbf{a}))]^2$$

$$o(\mathbf{a}) = L(\mathbf{a}|\mathcal{D}) = p(\mathcal{D}|\mathbf{a}) = \prod_{n=1}^{N} p(y_n|\mathbf{x}_n, \mathbf{a})$$

 $\mathbf{g}_f = \nabla_{\mathbf{x}} f(\mathbf{x}) = \frac{df(\mathbf{x})}{d\mathbf{x}} = \left[\frac{\partial f(\mathbf{x})}{\partial x_1}, \cdots, \frac{\partial f(\mathbf{x})}{\partial x_N}\right]^T$

$$\mathbf{g}_f = \nabla_{\mathbf{x}} f(\mathbf{x}) = d f(\mathbf{x}) / d\mathbf{x}$$

$$\mathbf{x} = [\mathbf{x}_1, \cdots, \mathbf{x}_N]^T$$

$$\mathbf{g}_f = \mathbf{0}$$

$$\mathbf{H}_{f} = \begin{bmatrix} \frac{\partial^{2} f(\mathbf{x})}{\partial x_{1}^{2}} & \cdots & \frac{\partial^{2} f(\mathbf{x})}{\partial x_{1} \partial x_{N}} \\ \vdots & \ddots & \vdots \\ \frac{\partial^{2} f(\mathbf{x})}{\partial x_{N} \partial x_{1}} & \cdots & \frac{\partial^{2} f(\mathbf{x})}{\partial x_{N}^{2}} \end{bmatrix}_{\mathbf{x} = \mathbf{x}^{*}}$$

$$f_1(x,y) = x^2 + y^2$$

$$f_2(x,y) = -x^2 - y^2$$

$$f_3(x,y) = x^2 - y^2$$

$$\mathbf{g}_{f_1} = \begin{bmatrix} 2x \\ 2y \end{bmatrix}, \quad \mathbf{g}_{f_2} = \begin{bmatrix} -2x \\ -2y \end{bmatrix}, \quad \mathbf{g}_{f_3} = \begin{bmatrix} 2x \\ -2y \end{bmatrix}$$

$$\mathbf{H}_{f_1} = \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix} > 0; \quad \mathbf{H}_{f_2} = \begin{bmatrix} -2 & 0 \\ 0 & -2 \end{bmatrix} < 0; \quad \mathbf{H}_{f_3} = \begin{bmatrix} 2 & 0 \\ 0 & -2 \end{bmatrix}$$

$$\mathbf{x}^* = [0, \, 0]^T$$

$$f_1(x,y) = f_2(x,y) = f_3(x,y) = 0$$

$$f_1(\mathbf{x}^*)$$

$$\mathbf{g}_f(\mathbf{x}) = \nabla_{\mathbf{x}} f(\mathbf{x}) = \frac{df(\mathbf{x})}{d\mathbf{x}} = \begin{bmatrix} \partial f/\partial x_1 \\ \vdots \\ \partial f/\partial x_N \end{bmatrix} =$$

$$f(x) = 0$$

$$\mathbf{x}_{n+1} = \mathbf{x}_n - \mathbf{J}_f^{-1}(\mathbf{x}_n) \mathbf{f}(\mathbf{x}_n)$$

$$\mathbf{g}_f(\mathbf{x}) = \mathbf{0}$$

$$\mathbf{J}_g(\mathbf{x}) = \mathbf{H}_f(\mathbf{x})$$

$$\mathbf{g}_f(\mathbf{x})$$

$$\mathbf{x}_{n+1} = \mathbf{x}_n - \mathbf{J}_g^{-1}(\mathbf{x}_n) \, \mathbf{g}_f(\mathbf{x}_n) = \mathbf{x}_n - \mathbf{H}_g^{-1}(\mathbf{x}_n) \, \mathbf{g}_f(\mathbf{x}_n)$$

$$\mathbf{J}_g = \mathbf{H}_f$$

$$\mathbf{g}(\mathbf{x}) = [g_1(\mathbf{x}), \cdots, g_N(\mathbf{x})]^T = \mathbf{0}$$

$$f(\mathbf{x}) = \frac{1}{2}\mathbf{g}^T(\mathbf{x})\mathbf{g}(\mathbf{x}) = \frac{1}{2}||\mathbf{g}(\mathbf{x})||^2 = \frac{1}{2}\sum_{i=1}^N |g_i(\mathbf{x})|^2$$

$$\frac{\partial f(\mathbf{x})}{\partial x_j} = \frac{\partial}{\partial x_j} \left(\frac{1}{2} \sum_{i=1}^{N} (g_i(\mathbf{x}))^2 \right) = \sum_{i=1}^{N} g_i(\mathbf{x}) \frac{\partial g_i(\mathbf{x})}{\partial x_j} = 0, \quad (j = 1, \dots, N)$$

$$f(\mathbf{x}) = \mathbf{g}^T \mathbf{g}$$

$$\nabla_{\mathbf{x}} f(\mathbf{x}) = \frac{d}{d\mathbf{x}} \left[\frac{1}{2} \mathbf{g}^T(\mathbf{x}) \mathbf{g}(\mathbf{x}) \right] = \frac{d}{d\mathbf{x}} \mathbf{g}(\mathbf{x}) \ \mathbf{g}(\mathbf{x}) = \mathbf{J}_{\mathbf{g}}(\mathbf{x}) \mathbf{g}(\mathbf{x}) = \mathbf{0}$$

$$g(x) = 0$$

$$f(x) = 0$$

$$x_0 < x_1 < x_2$$

$$f(x_0) > f(x_1) < f(x_2)$$

$$f(x_3) = f_{3a} > f(x_1)$$

$$x_3 - x_0 = a + c$$

$$x_0 < x_1 < x_3$$

$$f(x_3) = f_{3b} < f(x_1)$$

$$x_2 - x_1 = b$$

$$x_1 < x_3 < x_2$$

$$x_2 - x_0 = a + b$$

a+c=b,

$$\frac{+b}{b} = \frac{a+c}{a} = \frac{b}{a}$$

$$\frac{a+b}{b} = \frac{b}{b-c} =$$

$$a^2 + ab - b^2 = 0$$

$$a_{1,2} = \frac{-1 \pm \sqrt{5}}{2} b = \begin{cases} 0.618 \, b \\ -1.618 \, b \end{cases}$$

$$\frac{a}{b} = 0.618,$$
 $\frac{b}{a+b} = \frac{b}{0.618b+b} = \frac{1}{1.618} = 0.618$

$$|x_2 - x_0| < \epsilon(|x_1| + |x_3|)$$

$$f(x_0) < f(x_1) < f(x_2)$$

$$f(x_0) > f(x_1) > f(x_2)$$

$$x_0^{new} = x_1^{old}, \quad x_1^{new} = x_2^{old}, \quad x_2^{new} = x_2^{old} + c(x_2^{old} - x_1^{old})$$

$$f(x_2) > f(x_1)$$

$$d = x_2 - x_1$$

 c^p d \sim

$$f(a) > f(b) < f(c)$$

$$q(x) = f(a)\frac{(x-b)(x-c)}{(a-b)(a-c)} + f(b)\frac{(x-c)(x-a)}{(b-c)(b-a)} + f(c)\frac{(x-a)(x-b)}{(c-a)(c-b)}$$

$$q(x_{min})$$

$$q'(x) = f(a)\frac{(x-b) + (x-c)}{(a-b)(a-c)} + f(b)\frac{(x-c) + (x-a)}{(b-c)(b-a)} + f(c)\frac{(x-a) + (x-b)}{(c-a)(c-b)} = 0$$

$$(a-b)(b-c)(c-a)$$

$$f(a)(c-b)(2x-b-c) + f(b)(a-c)(2x-c-a) + f(c)(b-a)(2x-a-b)$$

$$2x[f(a)(c-b) + f(b)(a-c) + f(c)(b-a)] - [f(a)(c^2 - b^2) + f(b)(a^2 - c^2) + f(c)(b^2 - a^2)] = 0$$

$$\frac{1}{2} \frac{f(a)(c^2 - b^2) + f(b)(a^2 - c^2) + f(c)(b^2 - a^2)}{f(a)(c - b) + f(b)(a - c) + f(c)(b - a)}$$

$$b + \frac{1}{2} \frac{f(a)(c-b)(c+b-2b) + f(b)(a-c)(a+c-2b) + f(c)(b-a)(b+a-2b)}{f(a)(c-b) + f(b)(a-c) + f(c)(b-a)}$$

$$b + \frac{1}{2} \frac{f(a)(c-b)^2 + f(b)(a-c)(a+c-2b) - f(c)(b-a)^2}{f(a)(c-b) + f(b)(a-c) + f(c)(b-a)}$$

$$b + \frac{1}{2} \frac{[f(a) - f(b)](c - b)^2 - [f(c) - f(b)](b - a)^2}{[f(a) - f(b)](c - b) + [f(c) - f(b)](b - a)}$$

$$f(x_{min})$$

$$\{a, x_{min}, b\}$$

 $\{b, x_{min}, c\}$

$$[f(a) - f(b)](c - b)^{2} = [f(c) - f(b)](b - a)^{2}$$

$$x_{min} = b$$

 \mathbf{x}_0, \cdots $, \mathbf{x}_N$

$$\mathbf{x}_i = \mathbf{x}_0; \ (i = 1, \cdots, N)$$

$$S = \left\{ \sum_{i=0}^{N} c_i \mathbf{x}_i \mid \sum_{i=0}^{N} c_i = 1, \quad c_i \ge 0 \right\}$$

$$\{\mathbf{e}_1,\cdots,\mathbf{e}_N\}$$

$$\mathbf{x}_i = \mathbf{x}_0 + c\mathbf{e}_i$$

0 < \mathcal{B}

$$f(\mathbf{x}_0) \le f(\mathbf{x}_1) \le \dots \le f(\mathbf{x}_N)$$

 \mathbf{X}_{N-1} \mathbf{X}_{S}

$$\mathbf{c} = \frac{1}{N} \sum_{i=1}^{N} \mathbf{x}_i$$

$$\mathbf{x}_r = \mathbf{c} + \alpha(\mathbf{c} - \mathbf{x}_h)$$

$$f(\mathbf{x}_r)$$

$$f(\mathbf{x}_s < f(\mathbf{x}_r) < f(\mathbf{x}_l)$$

$$f(\mathbf{x}_r) < f(\mathbf{x}_l)$$

$$\mathbf{x}_e = \mathbf{c} + \gamma(\mathbf{x}_r - \mathbf{c})$$

$$f(\mathbf{x}_r) > f(\mathbf{x}_s)$$

$$bfx_r - \mathbf{c}$$

$$\mathbf{x}_c = \mathbf{c} + \beta(\mathbf{x}_h - \mathbf{c})$$

$$f(\mathbf{x}_c) < f(\mathbf{x}_h)$$

 $f(x_0) + f'(x_0)(x - x_0) + \frac{1}{2}f''(x_0)(x - x_0)^2 + \dots + \frac{1}{n!}f^{(n)}(x_0)(x - x_0)^n + \dots$

 $f(x_0) + f'(x_0)(x - x_0) + \frac{1}{2}f''(x_0)(x - x_0)^2 = q(x)$

$$f(x) = q(t)$$

$$f(x) = q(x)$$

$$\frac{d}{dx}q(x) = \frac{d}{dx}\left[f(x_0) + f'(x_0)(x - x_0) + \frac{1}{2}f''(x_0)(x - x_0)^2\right]$$

$$f'(x_0) + f''(x_0)(x - x_0) = 0$$

$$x^* = x = x_0 - \frac{f'(x_0)}{f''(x_0)} = x_0 + \Delta x_0$$

$$\Delta x_0 = -f'(x_0)/f''(x_0)$$

$$f''(x^*) > 0$$

$$f''(x^*) < 0$$

$$f(x) \neq q(x)$$

$$x_{n+1} = x_n + \Delta x_n = x_n - \frac{f'(x_n)}{f''(x_n)}$$

$$q(x_{n+1})$$

$$x_{n+1} = x_n - f'(x_n)/f''(x_n)$$

$$f'(x) = 0$$

$$q(x) = f(x_0) + f'(x_0)(x - x_0) + \frac{1}{2}f''(x_0)(x - x_0)^2$$

$$q(x) = \frac{a}{2}(x-b)^2 + c$$

$$\begin{cases} a = f''(x_0) \\ b = x_0 - f'(x_0)/f''(x_0) \\ c = f(x_0) - f'^2(x_0)/2f''(x_0) \end{cases}$$

$$q(x) = q(b) = c$$

$$f(x) = 2x^3 - 4x^2 + x$$
, $f'(x) = 6x^2 - 8x + 1$, $f''(x) = 12x - 8$

1.333

$$f(x^*) = -1.037$$

$$f(x_0) = f(3) = 21,$$
 $f'(x_0) = f'(3) = 31,$ $f''(x_0) = f''(3) = 28$

$$f''(x_0) > 0$$

$$\begin{cases} a = f''(3) = 28 \\ b = 3 - f'(3)/f''(3) = 1.8929 \\ c = f(3) - f'^{2}(3)/2f''(3) = 3.8393 \end{cases}$$

a = b = 1.8929

$$f(x_1) = f(1.8929) = 1.1251$$

$$f(x_0) = f(3) = 21$$

$$x = b = x_0 - f'(x_0)/f''(x_0)$$

$$x_1 = x_0 - f'(x_0)/f''(x_0)$$

$$f(\mathbf{x}) = f(x_1, \cdots, x_N)$$

$$f(\mathbf{x}) \approx f(\mathbf{x}_0) + \mathbf{g}_0^T(\mathbf{x} - \mathbf{x}_0) + \frac{1}{2}(\mathbf{x} - \mathbf{x}_0)^T \mathbf{H}_0(\mathbf{x} - \mathbf{x}_0) = q(\mathbf{x})$$

$$\mathbf{g}_{f}(\mathbf{x}_{0}) = \frac{d}{d\mathbf{x}} f(\mathbf{x}_{0}) = \begin{bmatrix} \frac{\partial f(\mathbf{x}_{0})}{\partial x_{1}} \\ \vdots \\ \frac{\partial f(\mathbf{x}_{0})}{\partial x_{N}} \end{bmatrix}$$

$$\mathbf{H}_{f}(\mathbf{x}_{0}) = \frac{d}{d\mathbf{x}}\mathbf{g}(\mathbf{x}_{0}) = \frac{d^{2}}{d\mathbf{x}^{2}}f(\mathbf{x}_{0}) = \begin{bmatrix} \frac{\partial^{2}f(\mathbf{x}_{0})}{\partial x_{1}^{2}} \\ \vdots \\ \frac{\partial^{2}f(\mathbf{x}_{0})}{\partial x_{N}\partial x_{1}} \end{bmatrix}$$

 $\frac{\partial^2 f(\mathbf{x}_0)}{\partial x_1 \partial x_N}$

 $\frac{\partial^2 f(\mathbf{x}_0)}{\partial \mathbf{x}_0}$

$$f(\mathbf{x}) = q(\mathbf{x})$$

$$\frac{d}{d\mathbf{x}}\left[f(\mathbf{x}_0) + \mathbf{g}_0^T(\mathbf{x} - \mathbf{x}_0) + \frac{1}{2}(\mathbf{x} - \mathbf{x}_0)^T \mathbf{H}_0(\mathbf{x} - \mathbf{x}_0)\right] = \mathbf{g}_0 + \mathbf{H}_0(\mathbf{x} - \mathbf{x}_0) = \mathbf{0}$$

$$\mathbf{x}^* = \mathbf{x} = \mathbf{x}_0 - \mathbf{H}(\mathbf{x}_0)^{-1} \mathbf{g}(\mathbf{x}_0)$$

$$\mathbf{H}^* = \mathbf{H}(\mathbf{x}^*)$$

$$H^* > 0$$

$$f(\mathbf{x}) \neq q(\mathbf{x})$$

$$\mathbf{x}_{n+1} = \mathbf{x}_n + \Delta \mathbf{x}_n = \mathbf{x}_n - \mathbf{H}_n^{-1} \mathbf{g}_n = \mathbf{x}_n + \mathbf{d}_n$$

$$\mathbf{g}_n = \mathbf{g}(\mathbf{x}_n)$$

$$\mathbf{H}_n = \mathbf{H}(\mathbf{x}_n)$$

$$\mathbf{d}_n = \Delta \mathbf{x}_n = -\mathbf{H}_n^{-1} \mathbf{g}_n$$

$$\mathbf{g}_n = f'(x_n)$$

$$\mathbf{H}_n = f''(x_n)$$

$$\mathbf{x}_{n+1} = \mathbf{x}_n - \mathbf{H}_n^{-1} \, \mathbf{g}_n$$

$$\mathbf{x}_{n+1} = \mathbf{x}_n + \delta_n \ \Delta \mathbf{x}_n = \mathbf{x}_n - \delta_n \ \mathbf{H}_n^{-1} \mathbf{g}_n$$

$$q(\mathbf{x}) = \frac{1}{2}(\mathbf{x} - \mathbf{b})^T \mathbf{A}(\mathbf{x} - \mathbf{b}) + c$$

$$\begin{cases}
\mathbf{A} = \mathbf{H}_0 \\
\mathbf{b} = \mathbf{x}_0 - \mathbf{H}^{-1} \mathbf{g}_0 \\
c = f(\mathbf{x}_0) - \mathbf{g}_0^T \mathbf{H}_0^{-1} \mathbf{g}_0 / 2
\end{cases}$$

$$\mathbf{g}_0 = \mathbf{g}_f(\mathbf{x}_0) = \mathbf{A}(\mathbf{x}_0 - \mathbf{b}) = \mathbf{A}\mathbf{e}_0, \qquad \mathbf{H}_0 = \mathbf{H}_f(\mathbf{x}_0) = \mathbf{A}$$

$$\mathbf{e}_0 = \mathbf{x}_0 - \mathbf{b}$$



$$\mathbf{x} = \mathbf{b}$$

$$q(\mathbf{b}) = c$$

$$q(x,y) = \frac{1}{2}[x, y] \begin{bmatrix} a & b/2 \\ b/2 & c \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \frac{1}{2}(ax^2 + bxy + cy^2)$$

$$\mathbf{g} = \begin{bmatrix} ax + by/2 \\ bx/2 + cy \end{bmatrix}, \qquad \mathbf{H} = \begin{bmatrix} a & b/2 \\ b/2 & c \end{bmatrix}, \quad \det \mathbf{H} = ac - b^2/4$$

 $\det \mathbf{H} = \lambda_1 \lambda_2$

$$f(0, 0) = 0$$

$$\lambda_1 = 2$$

$$\lambda_2 = 1$$

 $\det \mathbf{H} = \lambda_1 \lambda_2$

$$\lambda_2 = -0.562$$

 $\det \mathbf{H} = \lambda_1 \lambda_2$

$$\begin{cases} f_1(x_1, x_2, x_3) = 3x_1 - (x_2 x_3)^2 - 3/2 \\ f_2(x_1, x_2, x_3) = 4x_1^2 - 625 x_2^2 + 2x_2 - 1 \\ f_3(x_1, x_2, x_3) = exp(-x_1 x_2) + 20x_3 + 9 \end{cases}$$

$$(x_1 = 0.5, x_2 = 0, x_3 = -0.5)$$

$$o(\mathbf{x}) = \mathbf{f}^T(\mathbf{x})\mathbf{f}(\mathbf{x})$$

$$\mathbf{x}_0 = \mathbf{0}$$

n	$\mathbf{x} = [x_1, x_2, x_3]$	$ \mathbf{f}(\mathbf{x}) $
0	0.000000, 0.000000, 0.000000	1.016120e + 01
1	0.500000, 0.500000, -0.500000	1.552502e + 02
2	0.499550, 0.250800, -0.493801	3.881300e + 01
3	$0.500096, \ 0.126206, \ -0.496852$	9.702208e + 00
4	0.500025, 0.063914, -0.498405	2.425198e + 00
5	$0.500010, \ 0.032778, \ -0.499181$	6.059054e - 01
6	0.500005, 0.017231, -0.499570	1.510777e - 01
7	0.500003, 0.009498, -0.499763	3.737330e - 02
8	0.500002, 0.005712, -0.499857	8.959365e - 03
9	0.500001, 0.003968, -0.499901	1.900145e - 03
10	0.500001, 0.003326, -0.499917	2.577603e - 04
11	$0.500001, \ 0.003206, \ -0.499920$	8.932714e - 06
12	$0.500001, \ 0.003202, \ -0.499920$	1.238536e - 08
13	$0.500001, \ 0.003202, \ -0.499920$	2.371437e - 14

$$o(\mathbf{x}) = ||\mathbf{f}(\mathbf{x}^*)||^2 \approx 10^{-28}$$

$$\mathbf{x}^* = \begin{bmatrix} 0.5000008539707297 \\ 0.0032017070323056 \\ -0.4999200212218281 \end{bmatrix}$$

$$x_1 = x_0 + \Delta x_0 = x_0 - \delta f'(x_0),$$

$$\Delta x_0 = -\delta f'(x_0)$$

$$f(x_1) \approx f(x_0) + f'(x_0)\Delta x_0 = f(x_0) - |f'(x_0)|^2 \delta < f(x_0)$$

$$x_{n+1} = x_n - \delta_n \ f'(x_n)$$

$$f'(x^*) = 0$$

$$\mathbf{g}(\mathbf{x}) = df(\mathbf{x})/d\mathbf{x}$$

$$\mathbf{x}_{n+1} = \mathbf{x}_n + \Delta \mathbf{x}_n$$

$$\Delta \mathbf{x} = -\delta \mathbf{g} \ (\delta > 0)$$

$$f(\mathbf{x}_1) = f(\mathbf{x}_0 + \Delta \mathbf{x}) \approx f(\mathbf{x}_0) + \mathbf{g}_0^T \Delta \mathbf{x} = f(\mathbf{x}_0) - \delta \mathbf{g}_0^T \mathbf{g}_0 = f(\mathbf{x}_0) - \delta ||\mathbf{g}||^2 < f(\mathbf{x}_0)$$

$$\mathbf{x}_{n+1} = \mathbf{x}_n - \delta_n \, \mathbf{g}_n = (\mathbf{x}_{n-1} - \delta_{n-1} \, \mathbf{g}_{n-1}) - \delta_n \, \mathbf{g}_n = \dots = \mathbf{x}_0 - \sum_{i=0}^n \delta_n \, \mathbf{g}_i$$

$$\mathbf{d}_n = -\mathbf{g}_n$$

$$\mathbf{d}_n = -\mathbf{H}_n^{-1} \mathbf{g}_n$$

$$\mathbf{H} = \mathbf{I}$$

$$f(\mathbf{x}) = \mathbf{x}^T \mathbf{A} \mathbf{x} + \mathbf{b}^T \mathbf{x} + c = \begin{bmatrix} x_1 & x_2 \end{bmatrix} \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} b_1 & b_2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + c$$

 a_1 П.

$$\mathbf{A} = \begin{bmatrix} 2 & 1 \\ 1 & 1 \end{bmatrix}, \quad \mathbf{b} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}, \quad c = 0$$

 $f(x_1, x_2) = \frac{1}{2} \begin{bmatrix} x_1 & x_2 \end{bmatrix} \begin{bmatrix} 2 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \frac{1}{2} (2x_1^2 + 2x_1x_2 + x_2^2)$

$$f(x_1, x_2) = 0$$

$$\mathbf{x}_0 = [1, \ 2]^T$$

 \mathbf{g}_0

$$\begin{bmatrix} 2x_1 + x_2 \\ x_1 + x_2 \end{bmatrix}, \quad \mathbf{H} = \mathbf{A} = \begin{bmatrix} 2 & 1 \\ 1 & 1 \end{bmatrix}, \quad \mathbf{H}^{-1} = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$$

 $\mathbf{g} =$

$$\mathbf{d}_0 = -\mathbf{H}^{-1}\mathbf{g}_0 = -[1, \ 2]^T$$

$$\mathbf{x}_1 = \mathbf{x}_0 - \mathbf{H}^{-1}\mathbf{g}_0 = \left[egin{array}{cc} 1 \\ 2 \end{array}
ight] - \left[egin{array}{cc} 1 & -1 \\ -1 & 2 \end{array}
ight] \left[egin{array}{cc} 4 \\ 3 \end{array}
ight] = \left[egin{array}{cc} 0 \\ 0 \end{array}
ight]$$

$$\mathbf{d}_0 = -\mathbf{g}_0 = -[4, \ 3]^T$$

$$\mathbf{x}_1 = \mathbf{x}_0 - \delta \mathbf{g}_0 = \begin{bmatrix} 1 \\ 2 \end{bmatrix} - \delta \begin{bmatrix} 4 \\ 3 \end{bmatrix} = \begin{bmatrix} 1 - \delta 4 \\ 2 - \delta 3 \end{bmatrix}$$

$$\mathbf{x}_{n+1} = \mathbf{x}_n + \delta_n \mathbf{d}_n$$

$$f(\mathbf{x}_{n+1})$$

$$\theta = \cos^{-1} \frac{\mathbf{d}_n^T \mathbf{g}_n}{||\mathbf{d}_n|| ||\mathbf{g}_n||} > \frac{\pi}{2}$$
 i.e., $\mathbf{d}_n^T \mathbf{g}_n < 0$

$$\mathbf{d}_n = -\mathbf{H}_n^{-1} \mathbf{g}_n$$

$$\mathbf{d}_n^T \mathbf{g}_n = -\mathbf{g}_n^T \mathbf{H}_n^{-1} \mathbf{g}_n < 0, \qquad \mathbf{d}_n^T \mathbf{g}_n = -\mathbf{g}_n^T \mathbf{g}_n = -||\mathbf{g}_n||^2 < 0$$

$$f(\mathbf{x}_{n+1}) = f(\mathbf{x}_n + \delta_n \mathbf{d}_n)$$

$$\frac{d}{d\delta_n} f(\mathbf{x}_{n+1}) = \frac{d}{d\delta_n} f(\mathbf{x}_n + \delta_n \mathbf{d}_n) = \left(\frac{d f(\mathbf{x}_{n+1})}{d\mathbf{x}}\right)^T \frac{d(\mathbf{x}_n + \delta_n \mathbf{d}_n)}{d\delta_n} = \mathbf{g}_{n+1}^T \mathbf{d}_n = 0$$

$$\mathbf{g}_{n+1} = f'(\mathbf{x}_{n+1})$$

$$\mathbf{x}_{n+1} = \mathbf{x}_n + \delta \mathbf{d}_n$$

$$f(\mathbf{x}_{n+1}) = f(\mathbf{x}_n + \delta \mathbf{d}_n)$$

$$f(\mathbf{x}_{n+1}) = f(\mathbf{x}_n + \delta \mathbf{d}_n) \approx \left[f(\mathbf{x}_n + \delta \mathbf{d}_n) \right]_{\delta=0} + \delta \left[\frac{d}{d\delta} f(\mathbf{x}_n + \delta \mathbf{d}_n) \right]_{\delta=0} + \frac{\delta^2}{2} \left[\frac{d^2}{d\delta^2} f(\mathbf{x}_n + \delta \mathbf{d}_n) \right]_{\delta=0}$$

$$[f(\mathbf{x}_n + \delta \mathbf{d}_n)]_{\delta=0}$$

$$\left[\frac{d}{d\delta}f(\mathbf{x}_n + \delta\mathbf{d}_n)\right]_{\delta=0}$$

$$\mathbf{g}(\mathbf{x}_n + \delta \mathbf{d}_n)^T \mathbf{d}_n \bigg|_{\delta=0} = \mathbf{g}_n^T \mathbf{d}_n$$

$$\left[\frac{d^2}{d\delta^2}f(\mathbf{x}_n + \delta \mathbf{d}_n)\right]_{\delta=0}$$

$$\left[\frac{d}{d\delta}\mathbf{g}(\mathbf{x}_n + \delta\mathbf{d}_n)^T\right]_{\delta=0}\mathbf{d}_n = \left[\frac{d}{d\mathbf{x}}\mathbf{g}(\mathbf{x})\ \frac{d}{d\delta}(\mathbf{x}_n + \delta\mathbf{d}_n)\right]_{\delta=0}^T\mathbf{d}_n$$

$$(\mathbf{H}_n \mathbf{d}_n)^T \mathbf{d}_n = \mathbf{d}_n^T \mathbf{H}_n \mathbf{d}_n$$

$$\mathbf{H}_n = \mathbf{H}_n^T$$

$$f(\mathbf{x}_n + \delta \mathbf{d}_n) \approx f(\mathbf{x}_n) + \delta \mathbf{g}_n^T \mathbf{d}_n + \frac{\delta^2}{2} \mathbf{d}_n^T \mathbf{H}_n \mathbf{d}_n$$

$$f(\mathbf{x}_n + \delta \mathbf{d}_n)$$

$$\frac{d}{d\delta}f(\mathbf{x}_n + \delta \mathbf{d}_n) \approx \frac{d}{d\delta} \left(f(\mathbf{x}_n) + \delta \mathbf{g}_n^T \mathbf{d}_n + \frac{\delta^2}{2} \mathbf{d}_n^T \mathbf{H}_n \mathbf{d}_n \right) = \mathbf{g}_n^T \mathbf{d}_n + \delta \mathbf{d}_n^T \mathbf{H}_n \mathbf{d}_n = 0$$

$$\delta_n = -\frac{\mathbf{g}_n^T \mathbf{d}_n}{\mathbf{d}_n^T \mathbf{H}_n \mathbf{d}_n}$$

$$\delta_n = -\frac{\mathbf{g}_n^T \mathbf{d}_n}{\mathbf{d}_n^T \mathbf{H}_n \mathbf{d}_n} = \frac{\mathbf{g}_n^T (\mathbf{H}_n^{-1} \mathbf{g}_n)}{(\mathbf{H}_n^{-1} \mathbf{g}_n)^T \mathbf{H}_n (\mathbf{H}_n^{-1} \mathbf{g}_n)} = \frac{\mathbf{g}_n^T (\mathbf{H}_n^{-1} \mathbf{g}_n)}{(\mathbf{H}_n^{-1} \mathbf{g}_n)^T \mathbf{g}_n} = 1$$

$$\mathbf{x}_{n+1} = \mathbf{x}_n + \delta_n \mathbf{d}_n = x_n - \delta_n \mathbf{H}_n^{-1} \mathbf{g}_n = \mathbf{x}_n - \mathbf{H}_n^{-1} \mathbf{g}_n$$

$$\mathbf{g}_{n+1}^T \mathbf{d}_n = -\mathbf{g}_{n+1}^T \mathbf{g}_n = 0,$$

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\mathbf{g}_{n+1}
                \perp \mathbf{g}_n
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$$\mathbf{d}_{n+1} = -\mathbf{g}_{n+1}$$

$$\delta_n = -\frac{\mathbf{g}_n^T \mathbf{d}_n}{\mathbf{d}_n^T \mathbf{H}_n \mathbf{d}_n} = \frac{\mathbf{g}_n^T \mathbf{g}_n}{\mathbf{g}_n^T \mathbf{H}_n \mathbf{g}_n} = \frac{||\mathbf{g}_n||^2}{\mathbf{g}_n^T \mathbf{H}_n \mathbf{g}_n}$$

$$f''(\delta)\big|_{\delta=0}$$

$$\left[\frac{d^2}{d\delta^2}f(\mathbf{x} + \delta\mathbf{d})\right]_{\delta=0}$$

$$\left[\frac{d}{d\delta}f'(\mathbf{x} + \delta\mathbf{d})\right]_{\delta=0} = \lim_{\sigma \to 0} \frac{f'(\mathbf{x} + \sigma\mathbf{d}) - f'(\mathbf{x})}{\sigma}$$

$$\frac{\mathbf{g}^{T}(\mathbf{x} + \sigma \mathbf{d}) \, \mathbf{d} - \mathbf{g}^{T}(\mathbf{x}) \, \mathbf{d}}{\sigma} = \frac{\mathbf{g}_{\sigma}^{T} \mathbf{d} - \mathbf{g}^{T} \mathbf{d}}{\sigma}$$

$$\mathbf{g} = \mathbf{g}(\mathbf{x})$$

$$\mathbf{g}_{\sigma} = \mathbf{g}(\mathbf{x} + \sigma \mathbf{d})$$

$$\mathbf{d}_n^T \mathbf{H}_n \mathbf{d}_n$$

$$\frac{d}{d\delta}f(\mathbf{x}_n + \delta \mathbf{d}_n) \approx \mathbf{g}_n^T \mathbf{d}_n + \frac{\delta}{\sigma} \left(\mathbf{g}_{\sigma n}^T \mathbf{d}_n - \mathbf{g}_n^T \mathbf{d}_n \right) = 0$$

$$\delta_n = -\frac{\sigma \mathbf{g}_n^T \mathbf{d}_n}{(\mathbf{g}_{\sigma n}^T \mathbf{d}_n - \mathbf{g}_n^T \mathbf{d}_n)} = -\frac{\sigma \mathbf{g}_n^T \mathbf{d}_n}{(\mathbf{g}_{\sigma n} - \mathbf{g}_n)^T \mathbf{d}_n}$$

$$\delta = ||\mathbf{g}_n||^2 / \mathbf{g}_n^T \mathbf{H}_n \mathbf{g}_n$$

$$\delta_n = -\frac{\sigma \mathbf{g}_n^T \mathbf{d}_n}{(\mathbf{g}_{\sigma n} - \mathbf{g}_n)^T \mathbf{d}_n} = -\frac{\sigma \mathbf{g}_n^T \mathbf{g}_n}{(\mathbf{g}_{\sigma n} - \mathbf{g}_n)^T \mathbf{g}_n} = \frac{\sigma ||\mathbf{g}_n||^2}{||\mathbf{g}_n||^2 - \mathbf{g}_{\sigma n}^T \mathbf{g}_n}$$

$$\mathbf{x}_{n+1} = \mathbf{x}_n - \delta_n \mathbf{g}_n = \mathbf{x}_n + \frac{\sigma ||\mathbf{g}_n||^2}{\mathbf{g}_{\sigma n}^T \mathbf{g}_n - ||\mathbf{g}_n||^2} \mathbf{g}_n$$

n	$\mathbf{x} = [x_1, x_2, x_3]$	$ \mathbf{f}(\mathbf{x}) $
0	0.0000, 0.0000, 0.0000	1.032500e + 02
10	0.4246, -0.0073, -0.5002	1.535939e - 01
20	0.5015, 0.0064, -0.4998	2.448241e - 05
30	0.5009, 0.0057, -0.4998	9.178209e - 06
40	0.5006, 0.0052, -0.4998	4.17587e - 06
50	0.5004, 0.0049, -0.4999	2.122594e - 06
60	0.5003, 0.0047, -0.4999	1.182466e - 06
100	0.5001, 0.0043, -0.4999	1.805871e - 07
150	0.5000, 0.0041, -0.4999	2.720744e - 08
200	0.5000, 0.0041, -0.4999	4.934027e - 09
250	0.5000, 0.0040, -0.4999	9.630085e - 10
300	0.5000, 0.0040, -0.4999	1.939747e - 10
350	0.5000, 0.0040, -0.4999	3.961646e - 11
400	0.5000, 0.0040, -0.4999	8.140393e - 12
450	0.5000, 0.0040, -0.4999	1.677968e - 12
500	0.5000, 0.0040, -0.4999	3.462695e - 13
550	0.5000, 0.0040, -0.4999	7.136628e - 14
600	$0.5000, \ 0.0040, \ -0.4999$	1.474205e - 14

$$o(\mathbf{x}) = ||\mathbf{f}(\mathbf{x}^*)||^2 \approx 10^{-14}$$

 $o(\mathbf{x}) \approx 10^{-28}$

$$\mathbf{x}^* = \begin{bmatrix} 0.5000013623816102\\ 0.0040027495837189\\ -0.4999000311539049 \end{bmatrix}$$

$$f(\mathbf{x}_{n+1}) = f(\mathbf{x}_n + \delta \mathbf{d}_n) \le f(\mathbf{x}_n) + c_1 \delta \mathbf{d}^T \mathbf{g}_n$$

$$\mathbf{g}_{n+1}^T \mathbf{d}_n \ge c_2 \; \mathbf{g}_n^T \mathbf{d}_n$$

$$\mathbf{g}_n^T \mathbf{d}_n < 0$$

$$|\mathbf{g}_{n+1}^T \mathbf{d}_n| < |c_2 \; \mathbf{g}_n^T \mathbf{d}_n|$$

 $0 < c_1 < c_2 < 1$

$$\phi(\delta) = f(\mathbf{x}_n + \delta \mathbf{d}_n)$$

$$L_0(\delta) = a + b\delta$$

$$a = L_0(0) = \phi(\delta)\big|_{\delta=0} = f(\mathbf{x}_n)$$

 $b = \frac{d}{d\delta} f(\mathbf{x}_n + \delta \mathbf{d}_n)$

 $= \mathbf{g}_n^T \mathbf{d}_n < 0$

$$f(\mathbf{x}_n + \delta \mathbf{d}_n) < f(\mathbf{x}_n)$$

$$L_0(\delta) = a + b\delta = f(\mathbf{x}_n) + \mathbf{g}_n^T \mathbf{d}_n \delta$$

$$L_1(\delta) = f(\mathbf{x}_n)$$

$$L(\delta) = f(\mathbf{x}_n) + c_1 \, \mathbf{g}_n^T \mathbf{d}_n \delta$$

$$0 < c_1 \mathbf{g}_n^T \mathbf{d}_n < \mathbf{g}_n^T \mathbf{d}_n$$

$$f(\mathbf{x}_{n+1}) = f(\mathbf{x}_n + \delta \mathbf{d}_n) \le f(\mathbf{x}_n) + c_1 \mathbf{g}_n^T \mathbf{d}_n \delta < f(\mathbf{x}_n)$$

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\mathbf{g}_{n+}
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$$f(\delta_n) = f(\mathbf{x}_n - \delta_n \mathbf{g}_n)$$

$$\mathbf{g}_{n+1}^T \mathbf{g}_n = 0$$

$$\mathbf{x}_{n+1} = \mathbf{x}_n - \delta_n \mathbf{g}_n + \alpha_n (\mathbf{x}_n - \mathbf{x}_{n-1})$$

$$f(x,y) = (a-x)^2 + b(y-x^2)^2$$

100typically a

$$f(1, 1) = 0$$

$$\mathbf{x}_{n+1} = \mathbf{x}_n - \delta \mathbf{J}(\mathbf{x}_n)^{-1} \mathbf{f}(\mathbf{x}_n) = \mathbf{x}_n - \delta \mathbf{J}_n^{-1} \mathbf{f}_n$$

$$\mathbf{J}_n = \mathbf{J}(\mathbf{x}_n)$$

$$\mathbf{f}(\mathbf{x}_n)$$

$$\mathbf{x}_{n+1} = \mathbf{x}_n - \delta \mathbf{H}(\mathbf{x}_n)^{-1} \mathbf{g}(\mathbf{x}_n) = \mathbf{x}_n - \delta \mathbf{H}_n^{-1} \mathbf{g}_n$$

$$f(\mathbf{x}) = f(\mathbf{x}_{n+1}) + (\mathbf{x} - \mathbf{x}_{n+1})^T \mathbf{g}_{n+1} + \frac{1}{2} (\mathbf{x} - \mathbf{x}_{n+1})^T \mathbf{H}_{n+1} (\mathbf{x} - \mathbf{x}_{n+1}) + O(||\mathbf{x} - \mathbf{x}_{n+1}||^3)$$

$$\frac{d}{d\mathbf{x}}f(\mathbf{x}) = \mathbf{g}(\mathbf{x}) = \mathbf{g}_{n+1} + \mathbf{H}_{n+1}(\mathbf{x} - \mathbf{x}_{n+1}) + O(||\mathbf{x} - \mathbf{x}_{n+1}||^2)$$

$$\mathbf{g}(\mathbf{x}_n) = \mathbf{g}_n$$

$$\mathbf{g}_{n+1} - \mathbf{g}_n$$

$$\mathbf{H}_{n+1}(\mathbf{x}_{n+1} - \mathbf{x}_n) + O(||\mathbf{x}_{n+1} - \mathbf{x}_n||^2)$$

$$\mathbf{B}_{n+1}(\mathbf{x}_{n+1}-\mathbf{x}_n)$$

$$\mathbf{s}_n = \mathbf{x}_{n+1} - \mathbf{x}_n, \quad \mathbf{y}_n = \mathbf{g}_{n+1} - \mathbf{g}_n$$

$$\mathbf{B}_{n+1}\mathbf{s}_n=\mathbf{y}_n,$$

$$\mathbf{B}_{n+1}^{-1}\mathbf{y}_n = \mathbf{s}_n$$

$$\mathbf{d}_n = -\mathbf{B}_n^{-1} \mathbf{g}_n$$

$$\mathbf{B}_{n+1} = \mathbf{B}_n + \Delta \mathbf{B}_n$$

$$\mathbf{B}_{n+1}^{-1} = \mathbf{B}_n^{-1} + \Delta \mathbf{B}_n^{-1}$$

$$\mathbf{B}_n = \mathbf{I}$$

$$\mathbf{B}_{n+1} = \mathbf{B}_n + \mathbf{u}\mathbf{u}^T$$

$$\mathbf{B}_{n+1}\mathbf{s}_n = \mathbf{B}_n\mathbf{s}_n + \mathbf{u}\mathbf{u}^T\mathbf{s}_n = \mathbf{y}_n,$$

$$\mathbf{u}(\mathbf{u}^T\mathbf{s}_n) = \mathbf{y}_n - \mathbf{B}_n\mathbf{s}_n$$

$$\mathbf{w} = \mathbf{y}_n - \mathbf{B}_n \mathbf{s}_n$$

$$\mathbf{u} = c\mathbf{w}$$

$$\mathbf{u}\mathbf{u}^T\mathbf{s}_n = c^2\mathbf{w}\mathbf{w}^T\mathbf{s}_n = \mathbf{w}$$

$$c = 1/(\mathbf{w}^T \mathbf{s}_n)^{1/2}$$

$$\mathbf{u} = c\mathbf{w} = \frac{\mathbf{y}_n - \mathbf{B}_n \mathbf{s}_n}{(\mathbf{w}^T \mathbf{s}_n)^{1/2}} = \frac{\mathbf{y}_n - \mathbf{B}_n \mathbf{s}_n}{((\mathbf{y}_n - \mathbf{B}_n \mathbf{s}_n)^T \mathbf{s}_n)^{1/2}}$$

$$\mathbf{B}_{n+1} = \mathbf{B}_n + \mathbf{u}\mathbf{u}^T = \mathbf{B}_n + \frac{(\mathbf{y}_n - \mathbf{B}_n\mathbf{s}_n)(\mathbf{y}_n - \mathbf{B}_n\mathbf{s}_n)^T}{(\mathbf{y}_n - \mathbf{B}_n\mathbf{s}_n)^T\mathbf{s}_n}$$

$$\left(\mathbf{B}_n + \mathbf{u}\mathbf{u}^T\right)^{-1} = \mathbf{B}_n^{-1} - \frac{\mathbf{B}_n^{-1}\mathbf{u}\mathbf{u}^T\mathbf{B}_n^{-1}}{1 + \mathbf{u}^T\mathbf{B}_n^{-1}\mathbf{u}}$$

$$\mathbf{B}_n^{-1} + rac{(\mathbf{s}_n - \mathbf{B}_n^{-1}\mathbf{y}_n)(\mathbf{s}_n - \mathbf{B}_n^{-1}\mathbf{y}_n)^T}{(\mathbf{s}_n - \mathbf{B}_n^{-1}\mathbf{y}_n)^T\mathbf{y}_n}$$

$$\mathbf{B}_{n+1}^{-1} = \mathbf{B}_n^{-1} + \mathbf{u}\mathbf{u}^T$$

$$\mathbf{B}_{n+1}^{-1}\mathbf{y}_n = \mathbf{B}_n^{-1}\mathbf{y}_n + \mathbf{u}\mathbf{u}^T\mathbf{y}_n = \mathbf{s}_n,$$

$$\mathbf{w}^{T}\mathbf{y}_{n} = (\mathbf{s}_{n} - \mathbf{B}_{n}^{-1}\mathbf{y}_{n})^{T}\mathbf{y}_{n} = \mathbf{s}_{n}^{T}\mathbf{y}_{n} - \mathbf{y}_{n}^{T}\mathbf{B}_{n}^{-1}\mathbf{y}_{n} > 0$$

$$\mathbf{B}_{n+1} = \mathbf{B}_n + \alpha \mathbf{u} \mathbf{u}^T + \beta \mathbf{v} \mathbf{v}^T$$

$$\mathbf{B}_{n+1}\mathbf{s}_n = (\mathbf{B}_n + \alpha \mathbf{u} \mathbf{u}^T + \beta \mathbf{v} \mathbf{v}^T)\mathbf{s}_n = \mathbf{B}_n \mathbf{s}_n + \mathbf{u}(\alpha \mathbf{u}^T \mathbf{s}_n) + \mathbf{v}(\beta \mathbf{v}^T \mathbf{s}_n) = \mathbf{y}_n$$

$$\mathbf{u}(\alpha \mathbf{u}^T \mathbf{s}_n) + \mathbf{v}(\beta \mathbf{v}^T \mathbf{s}_n) = \mathbf{y}_n - \mathbf{B}_n \mathbf{s}_n$$

$$\mathbf{u} = \mathbf{y}_n, \quad \alpha = \frac{1}{\mathbf{s}_n^T \mathbf{y}_n}, \quad \mathbf{v} = \mathbf{B}_n \mathbf{s}_n, \quad \beta = -\frac{1}{\mathbf{v}^T \mathbf{s}_n} = -\frac{1}{\mathbf{s}_n^T \mathbf{B}_n \mathbf{s}_n}$$

$$\mathbf{B}_{n+1} = \mathbf{B}_n + \alpha \mathbf{u} \mathbf{u}^T + \beta \mathbf{v} \mathbf{v}^T$$

$$\mathbf{B}_{n+1} = \mathbf{B}_n + \alpha \mathbf{u} \mathbf{u}^T + \beta \mathbf{v} \mathbf{v}^T = \mathbf{B}_n + \frac{\mathbf{y}_n \mathbf{y}_n^T}{\mathbf{y}_n^T \mathbf{s}_n} - \frac{\mathbf{B}_n \mathbf{s}_n \mathbf{s}_n^T \mathbf{B}_n}{\mathbf{s}_n^T \mathbf{B}_n \mathbf{s}_n}$$

$$\mathbf{U} = [\mathbf{u}_1 \ \mathbf{u}_2]$$

$$\mathbf{V} = [\mathbf{v}_1 \ \mathbf{v}_2]$$

$$\mathbf{u}_1 = \mathbf{v}_1 = \frac{\mathbf{y}_n}{(\mathbf{s}_n^T \mathbf{y}_n)^{1/2}}, \qquad \mathbf{u}_2 = -\mathbf{v}_2 = \frac{\mathbf{B}_n \mathbf{s}_n}{(\mathbf{s}_n^T \mathbf{B}_n \mathbf{s}_n)^{1/2}}$$

$$\mathbf{B}_{n+1} = \mathbf{B}_n + \mathbf{u}_1 \mathbf{v}_1^T + \mathbf{u}_2 \mathbf{v}_2^T = (\mathbf{B}_n + \mathbf{U}\mathbf{V}^T)^{-1}$$

$$(\mathbf{B}_n + \mathbf{U}\mathbf{V}^T)^{-1} = \mathbf{B}_n^{-1} - \mathbf{B}_n^{-1}\mathbf{U}(\mathbf{I} + \mathbf{V}^T\mathbf{B}_n^{-1}\mathbf{U})^{-1}\mathbf{V}^T\mathbf{B}_n^{-1}$$

$$\mathbf{B}_n^{-1} - \mathbf{B}_n^{-1} \mathbf{U} \mathbf{C}^{-1} \mathbf{V}^T \mathbf{B}_n^{-1} = \mathbf{B}_n^{-1} - \mathbf{B}_n^{-1} [\mathbf{u}_1 \ \mathbf{u}_2] \mathbf{C}^{-1} (\mathbf{B}_n^{-1} [\mathbf{v}_1 \ \mathbf{v}_2])^T$$

$$\mathbf{C} = \begin{bmatrix} c_{11} & c_{12} \\ c_{21} & c_{22} \end{bmatrix} = \mathbf{I} + \mathbf{V}^T \mathbf{B}_n^{-1} \mathbf{U} = \mathbf{I} + [\mathbf{v}_1 \ \mathbf{v}_2]^T \mathbf{B}_n^{-1} [\mathbf{u}_1 \ \mathbf{u}_2]$$

$$c_{11} = 1 + \mathbf{v}_1^T \mathbf{B}_n^{-1} \mathbf{u}_1 = 1 + \frac{\mathbf{y}_n^T \mathbf{B}_n^{-1} \mathbf{y}_n}{\mathbf{s}_n^T \mathbf{y}_n}$$

$$c_{22} = 1 + \mathbf{v}_2 \mathbf{B}_n^{-1} \mathbf{u}_2 = 1 - \frac{\mathbf{s}_n^T \mathbf{B}_n \mathbf{B}_n^{-1} \mathbf{B}_n \mathbf{s}_n}{\mathbf{s}_n^T \mathbf{B}_n \mathbf{s}_n} = 0$$

$$c_{12} = \mathbf{v}_1^T \mathbf{B}_n^{-1} \mathbf{u}_2 = \frac{\mathbf{y}_n^T \mathbf{B}_n^{-1} \mathbf{B}_n \mathbf{s}_n}{(\mathbf{s}_n^T \mathbf{y}_n)^{1/2} (\mathbf{s}_n^T \mathbf{B}_n \mathbf{s}_n)^{1/2}} = \frac{(\mathbf{s}_n^T \mathbf{y}_n)^{1/2}}{(\mathbf{s}_n^T \mathbf{B}_n \mathbf{s}_n)^{1/2}}$$

$$c_{21} = \mathbf{v}_2^T \mathbf{B}_n^{-1} \mathbf{u}_1 = -c_{12}$$

$$\mathbf{C} = \begin{bmatrix} c_{11} & c_{12} \\ -c_{12} & 0 \end{bmatrix}, \qquad \mathbf{C}^{-1} = \begin{bmatrix} 0 & -1/c_{12} \\ 1/c_{12} & c_{11}/c_{12}^2 \end{bmatrix}$$

$$\mathbf{B}_{n}^{-1} - [\mathbf{B}_{n}^{-1}\mathbf{u}_{1} \ \mathbf{B}_{n}^{-1}\mathbf{u}_{2}] \begin{bmatrix} 0 & -1/c_{12} \\ 1/c_{12} & c_{11}/c_{12}^{2} \end{bmatrix} [\mathbf{B}_{n}^{-1}\mathbf{v}_{1} \ \mathbf{B}_{n}^{-1}\mathbf{v}_{2}]^{T}$$

$$\mathbf{B}_{n}^{-1} - \frac{1}{c_{12}} [\mathbf{B}_{n}^{-1} \mathbf{u}_{2} \mathbf{v}_{1}^{T} \mathbf{B}_{n}^{-1} - \mathbf{B}_{n}^{-1} \mathbf{u}_{1} \mathbf{v}_{2}^{T} \mathbf{B}_{n}^{-1}] - \frac{c_{11}}{c_{12}^{2}} \mathbf{B}_{n}^{-1} \mathbf{u}_{2} \mathbf{v}_{2}^{T} \mathbf{B}_{n}^{-1}$$

$$\frac{1}{c_{12}}\mathbf{B}_{n}^{-1}\mathbf{u}_{2}\mathbf{v}_{1}^{T}\mathbf{B}_{n}^{-1} = \frac{(\mathbf{s}_{n}^{T}\mathbf{B}_{n}\mathbf{s}_{n})^{1/2}}{(\mathbf{s}_{n}^{T}\mathbf{y}_{n})^{1/2}}\mathbf{B}_{n}^{-1}\frac{\mathbf{B}_{n}\mathbf{s}_{n}}{(\mathbf{s}_{n}^{T}\mathbf{B}_{n}\mathbf{s}_{n})^{1/2}}\frac{\mathbf{y}_{n}^{T}}{(\mathbf{s}_{n}^{T}\mathbf{y}_{n})^{1/2}}\mathbf{B}_{n}^{-1} = \frac{\mathbf{s}_{n}\mathbf{y}_{n}^{T}\mathbf{B}_{n}^{-1}}{\mathbf{s}_{n}^{T}\mathbf{y}_{n}}$$

$$-\frac{1}{c_{12}}\mathbf{B}_{n}^{-1}\mathbf{u}_{1}\mathbf{v}_{2}^{T}\mathbf{B}_{n}^{-1} = \frac{(\mathbf{s}_{n}^{T}\mathbf{B}_{n}\mathbf{s}_{n})^{1/2}}{(\mathbf{s}_{n}^{T}\mathbf{y}_{n})^{1/2}}\mathbf{B}_{n}^{-1}\frac{\mathbf{y}_{n}}{(\mathbf{s}_{n}^{T}\mathbf{y}_{n})^{1/2}}\frac{\mathbf{s}_{n}^{T}\mathbf{B}_{n}}{(\mathbf{s}_{n}^{T}\mathbf{B}_{n}\mathbf{s}_{n})^{1/2}}\mathbf{B}_{n}^{-1} = \frac{\mathbf{B}_{n}^{-1}\mathbf{y}_{n}\mathbf{s}_{n}^{T}}{\mathbf{s}_{n}^{T}\mathbf{y}_{n}}$$

$$\frac{c_{11}}{c_{12}^2} \mathbf{B}_n^{-1} \mathbf{u}_2 \mathbf{v}_2^T \mathbf{B}_n^{-1}$$

$$-\left(1+\frac{\mathbf{y}_n^T\mathbf{B}_n^{-1}\mathbf{y}_n}{\mathbf{s}_n^T\mathbf{y}_n}\right)\frac{\mathbf{s}_n^T\mathbf{B}_n\mathbf{s}_n}{\mathbf{s}_n^T\mathbf{y}_n}\mathbf{B}_n^{-1}\frac{\mathbf{B}_n\mathbf{s}_n}{(\mathbf{s}_n^T\mathbf{B}_n\mathbf{s}_n)^{1/2}}\frac{\mathbf{s}_n^T\mathbf{B}_n}{(\mathbf{s}_n^T\mathbf{B}_n\mathbf{s}_n)^{1/2}}\mathbf{B}_n^{-1}$$

$$-\left(1 + \frac{\mathbf{y}_n^T \mathbf{B}_n^{-1} \mathbf{y}_n}{\mathbf{s}_n^T \mathbf{y}_n}\right) \frac{\mathbf{s}_n \mathbf{s}_n^T}{\mathbf{s}_n^T \mathbf{y}_n}$$

$$\mathbf{B}_{n+1}^{-1} = \mathbf{B}_n^{-1} - \frac{\mathbf{B}_n^{-1}\mathbf{y}_n\mathbf{s}_n^T + \mathbf{s}_n\mathbf{y}_n^T\mathbf{B}_n^{-1}}{\mathbf{s}_n^T\mathbf{y}_n} + \left(1 + \frac{\mathbf{y}_n^T\mathbf{B}_n^{-1}\mathbf{y}_n}{\mathbf{s}_n^T\mathbf{y}_n}\right) \frac{\mathbf{s}_n\mathbf{s}_n^T}{\mathbf{s}_n^T\mathbf{y}_n}$$

$$\mathbf{B}_{n+1}^{-1} = \mathbf{B}_n^{-1} + \alpha \mathbf{u} \mathbf{u}^T + \beta \mathbf{v} \mathbf{v}^T$$

$$\mathbf{B}_{n+1}^{-1}\mathbf{y}_n = (\mathbf{B}_n^{-1} + \alpha \mathbf{u} \mathbf{u}^T + \beta \mathbf{v} \mathbf{v}^T)\mathbf{y}_n = \mathbf{B}_n^{-1}\mathbf{y}_n + \mathbf{u}(\alpha \mathbf{u}^T \mathbf{y}_n) + \mathbf{v}(\beta \mathbf{v}^T \mathbf{y}_n) = \mathbf{s}_n$$

$$\mathbf{u}(\alpha \mathbf{u}^T \mathbf{y}_n) + \mathbf{v}(\beta \mathbf{v}^T \mathbf{y}_n) = \mathbf{s}_n - \mathbf{B}_n^{-1} \mathbf{y}_n$$

$$\mathbf{u} = \mathbf{s}_n, \quad \alpha = \frac{1}{\mathbf{s}^T \mathbf{y}_n}, \quad \mathbf{v} = \mathbf{B}_n^{-1} \mathbf{y}_n, \quad \beta = -\frac{1}{\mathbf{v}^T \mathbf{y}_n} = -\frac{1}{\mathbf{y}_n^T \mathbf{B}_n^{-1} \mathbf{y}_n}$$

$$\mathbf{B}_{n+1}^{-1} = \mathbf{B}_n^{-1} + \alpha \mathbf{u} \mathbf{u}^T + \beta \mathbf{v} \mathbf{v}^T$$

$$\mathbf{B}_{n+1}^{-1} = \mathbf{B}_n^{-1} + \alpha \mathbf{u} \mathbf{u}^T + \beta \mathbf{v} \mathbf{v}^T = \mathbf{B}_n^{-1} + \frac{\mathbf{s}_n \mathbf{s}_n^T}{\mathbf{s}_n^T \mathbf{y}_n} - \frac{\mathbf{B}_n^{-1} \mathbf{y}_n \mathbf{y}_n^T \mathbf{B}_n^{-1}}{\mathbf{y}_n^T \mathbf{B}_n^{-1} \mathbf{y}_n}$$

$$\mathbf{B}_{n+1} = \mathbf{B}_n - \frac{\mathbf{B}_n \mathbf{s}_n \mathbf{y}_n^T + \mathbf{y}_n \mathbf{s}_n^T \mathbf{B}_n}{\mathbf{y}_n^T \mathbf{s}_n} + \left(1 + \frac{\mathbf{s}_n^T \mathbf{B}_n \mathbf{s}_n}{\mathbf{y}_n^T \mathbf{s}_n}\right) \frac{\mathbf{y}_n \mathbf{y}_n^T}{\mathbf{y}_n^T \mathbf{s}_n}$$

 $\mathbf{z}^T \mathbf{B}_n \mathbf{z} > 0$

 $\mathbf{g}_{n+1}^T \mathbf{d}_n \ge c_2 \; \mathbf{g}_n^T \mathbf{d}_n$

$$\left(\mathbf{g}_{n+1} - c_2 \; \mathbf{g}_n\right)^T \mathbf{d}_n \ge 0$$

$$\delta \mathbf{d}_n = \mathbf{x}_{n+1} - \mathbf{x}_n = \mathbf{s}_n$$

$$(\mathbf{g}_{n+1} - c_2 \mathbf{g}_n)^T \mathbf{s}_n = \mathbf{g}_{n+1}^T \mathbf{s}_n - c_2 \mathbf{g}_n^T \mathbf{s}_n \ge 0$$

$$\mathbf{y}_n^T \mathbf{s}_n = (\mathbf{g}_{n+1} - \mathbf{g}_n)^T \mathbf{s}_n = \mathbf{g}_{n+1}^T \mathbf{s}_n - \mathbf{g}_n^T \mathbf{s}_n$$

$$\mathbf{y}_n^T \mathbf{s}_n - (\mathbf{g}_{n+1}^T \mathbf{s}_n - c_2 \mathbf{g}_n^T \mathbf{s}_n) = (c_2 - 1) \mathbf{g}_n^T \mathbf{s}_n > 0$$

$$\mathbf{g}_n^T \mathbf{s}_n < 0$$

$$\mathbf{y}_n^T \mathbf{s}_n \ge \mathbf{g}_{n+1}^T \mathbf{s}_n - c_2 \mathbf{g}_n^T \mathbf{s}_n \ge 0$$

$$\mathbf{y}_n^T \mathbf{s}_n > 0$$

$$\mathbf{B}_n = \mathbf{L}\mathbf{L}^T$$

$$\mathbf{a} = \mathbf{L}^T \mathbf{z}$$

$$\mathbf{b} = \mathbf{L}^T \mathbf{s}$$

$$\mathbf{a}^T \mathbf{a} = \mathbf{z}^T \mathbf{B}_n \mathbf{z}, \quad \mathbf{b}^T \mathbf{b} = \mathbf{s}^T \mathbf{B}_n \mathbf{s}, \quad \mathbf{a}^T \mathbf{b} = \mathbf{z}^T \mathbf{B}_n \mathbf{s}$$

$$\mathbf{z}^T \left[\mathbf{B}_n - \frac{\mathbf{B}_n \mathbf{s}_n \mathbf{s}_n^T \mathbf{B}_n}{\mathbf{s}_n^T \mathbf{B}_n \mathbf{s}_n} \right] \mathbf{z} = \mathbf{z}^T \mathbf{B}_n \mathbf{z} - \frac{(\mathbf{z}^T \mathbf{B}_n \mathbf{s}_n)^2}{\mathbf{s}_n^T \mathbf{B}_n \mathbf{s}_n} = \mathbf{a}^T \mathbf{a} - \frac{(\mathbf{a}^T \mathbf{b})^2}{\mathbf{b}^T \mathbf{b}} \ge 0$$

$$\mathbf{s}_n^T \mathbf{y}_n \ge 0$$

$$\mathbf{z}^T \left(\frac{\mathbf{s}_n \mathbf{s}_n^T}{\mathbf{s}_n^T \mathbf{y}_n} \right) \mathbf{z} \ge 0$$

$$\mathbf{z}^{T}\mathbf{B}_{n+1}\mathbf{z} = \mathbf{z}^{T}\left(\mathbf{B}_{n} - \frac{\mathbf{B}_{n}\mathbf{s}_{n}\mathbf{s}_{n}^{T}\mathbf{B}_{n}}{\mathbf{s}_{n}^{T}\mathbf{B}_{n}\mathbf{s}_{n}}\right)\mathbf{z} + \mathbf{z}^{T}\left(\frac{\mathbf{y}_{n}\mathbf{y}_{n}^{T}}{\mathbf{s}_{n}^{T}\mathbf{y}_{n}}\right)\mathbf{z} \geq 0$$

$$f(\mathbf{x}) = \frac{1}{2}\mathbf{x}^T \mathbf{A} \mathbf{x} - \mathbf{b}^T \mathbf{x} + c$$

$$\mathbf{A} = \mathbf{A}^T$$

$$\mathbf{g}(\mathbf{x}) = \frac{d}{d\mathbf{x}} f(\mathbf{x}) = \frac{d}{d\mathbf{x}} \left(\frac{1}{2} \mathbf{x}^T \mathbf{A} \mathbf{x} - \mathbf{b}^T \mathbf{x} + c \right) = \mathbf{A} \mathbf{x} - \mathbf{b}$$

$$\mathbf{H}(\mathbf{x}) = \frac{d^2}{d\mathbf{x}^2} f(\mathbf{x}) = \frac{d}{d\mathbf{x}} \mathbf{g} = \frac{d}{d\mathbf{x}} (\mathbf{A}\mathbf{x} - \mathbf{b}) = \mathbf{A}$$

$$\mathbf{g}(\mathbf{x}) = \mathbf{A}\mathbf{x} - \mathbf{b} = \mathbf{0}$$

$$\mathbf{x}^* = \mathbf{A}^{-1}\mathbf{b}$$

$$f(\mathbf{x}^*) = \frac{1}{2} (\mathbf{A}^{-1} \mathbf{b})^T \mathbf{A} (\mathbf{A}^{-1} \mathbf{b}) - \mathbf{b}^T (\mathbf{A}^{-1} \mathbf{b}) + c = -\frac{1}{2} \mathbf{b}^T \mathbf{A}^{-1} \mathbf{b} + c$$

$$\mathbf{g}(\mathbf{x}^*) = \mathbf{A}\mathbf{x}^* - \mathbf{b} = \mathbf{0}$$

$$\varepsilon = ||\mathbf{g}_n||^2$$

$$\mathbf{u}^T \mathbf{A} \mathbf{v} = (\mathbf{A} \mathbf{v})^T \mathbf{u} = \mathbf{v}^T \mathbf{A} \mathbf{u} = 0$$

$$\mathbf{v}^T \mathbf{u} = 0$$

$$\{\mathbf{d}_0,\cdots,\mathbf{d}_{N-1}\}$$

$$\mathbf{d}_i^T \mathbf{A} \mathbf{d}_j = 0$$

$$\mathbf{v}_1 = \left[egin{array}{c} 1 \\ 0 \end{array}
ight], \quad \mathbf{v}_2 = \left[egin{array}{c} 0 \\ 1 \end{array}
ight], \quad \mathbf{A} = \left[egin{array}{c} \end{array}
ight]$$

 $\begin{bmatrix} 3 & 1 \\ 1 & 2 \end{bmatrix}$

$$\mathbf{u}_1 = \mathbf{v}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \qquad \mathbf{u}_2 = \mathbf{v}_2 - \frac{\mathbf{u}_1^T \mathbf{A} \mathbf{v}_2}{\mathbf{u}_1^T \mathbf{A} \mathbf{u}_1} \mathbf{u}_1 = \begin{bmatrix} -1/3 \\ 1 \end{bmatrix}$$

v

$$\mathbf{p}_{\mathbf{v}_1}(\mathbf{x}) = \frac{\mathbf{v}_1^T \mathbf{x}}{\mathbf{v}_1^T \mathbf{v}_1} \mathbf{v}_1 = 2 \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 2 \\ 0 \end{bmatrix}, \quad \mathbf{p}_{\mathbf{v}_2}(\mathbf{x}) = \frac{\mathbf{v}_2^T \mathbf{x}}{\mathbf{v}_2^T \mathbf{v}_2} \mathbf{v}_2 = 3 \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 3 \end{bmatrix}$$

$$\mathbf{p}_{\mathbf{u}_1}(\mathbf{x}) = \frac{\mathbf{u}_1^T \mathbf{A} \mathbf{x}}{\mathbf{u}_1^T \mathbf{A} \mathbf{u}_1} \mathbf{u}_1 = 3 \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 3 \\ 0 \end{bmatrix}, \qquad \mathbf{p}_{\mathbf{u}_2}(\mathbf{x}) = \frac{\mathbf{u}_2^T \mathbf{A} \mathbf{x}}{\mathbf{u}_2^T \mathbf{A} \mathbf{u}_2} \mathbf{u}_2 = 3 \begin{bmatrix} -1/3 \\ 1 \end{bmatrix} = \begin{bmatrix} -1 \\ 3 \end{bmatrix}$$

$$2\begin{bmatrix} 1\\0 \end{bmatrix} + 3\begin{bmatrix} 0\\1 \end{bmatrix} = 2\mathbf{v}_1 + 3\mathbf{v}_2 = \mathbf{p}_{\mathbf{v}_1}(\mathbf{x}) + \mathbf{p}_{\mathbf{v}_2}(\mathbf{x})$$

$$3\begin{bmatrix} 1\\0 \end{bmatrix} + 3\begin{bmatrix} -1/3\\1 \end{bmatrix} = 3\mathbf{u}_1 + 3\mathbf{u}_2 = \mathbf{p}_{\mathbf{u}_1}(\mathbf{x}) + \mathbf{p}_{\mathbf{u}_2}(\mathbf{x})$$

$$\{-\mathbf{g}_0,\cdots,-\mathbf{g}_n,\cdots\}$$

$$\mathbf{d}_i^T \mathbf{A} \mathbf{d}_j = 0 \ (i \neq j)$$

$$\mathbf{x}_{n+1} = \mathbf{x}_n + \delta_n \mathbf{d}_n = \dots = \mathbf{x}_0 + \sum_{i=0}^n \delta_i \mathbf{d}_i$$

$$\mathbf{x}^* = \mathbf{x}_{n+1} - \mathbf{e}_{n+1} = \mathbf{x}_n - \mathbf{e}_n$$

$$\mathbf{e}_{n+1} = \mathbf{e}_n + \delta_n \mathbf{d}_n = \dots = \mathbf{e}_0 + \sum_{i=0}^n \delta_i \mathbf{d}_i$$

$$f(\mathbf{x}) = \mathbf{x}^T \mathbf{A} \mathbf{x} / 2 - \mathbf{b}^T \mathbf{x} + c$$

$$\mathbf{g}_i = f'(\mathbf{x}_i) = \mathbf{A}\mathbf{x}_i - \mathbf{b} = \mathbf{A}(\mathbf{x}^* + \mathbf{e}_i) - \mathbf{b} = \mathbf{A}\mathbf{x}^* - \mathbf{b} + \mathbf{A}\mathbf{e}_i = \mathbf{A}\mathbf{e}_i$$

$$-\frac{\mathbf{g}_{i}^{T}\mathbf{d}_{i}}{\mathbf{d}_{i}^{T}\mathbf{A}\mathbf{d}_{i}} = -\frac{\mathbf{e}_{i}^{T}\mathbf{A}\mathbf{d}_{i}}{\mathbf{d}_{i}^{T}\mathbf{A}\mathbf{d}_{i}} = -\frac{\mathbf{d}_{i}^{T}\mathbf{A}\mathbf{e}_{i}}{\mathbf{d}_{i}^{T}\mathbf{A}\mathbf{d}_{i}} = -\frac{\mathbf{d}_{i}^{T}\mathbf{A}\left(\mathbf{e}_{0} + \sum_{j=0}^{i-1}\delta_{j}\mathbf{d}_{j}\right)}{\mathbf{d}_{i}^{T}\mathbf{A}\mathbf{d}_{i}}$$

$$-\frac{\mathbf{d}_i^T \mathbf{A} \mathbf{e}_0 + \sum_{j=0}^{i-1} \delta_j \mathbf{d}_i^T \mathbf{A} \mathbf{d}_j}{\mathbf{d}_i^T \mathbf{A} \mathbf{d}_i} = -\frac{\mathbf{d}_i^T \mathbf{A} \mathbf{e}_0}{\mathbf{d}_i^T \mathbf{A} \mathbf{d}_i}$$

$$\mathbf{d}_i^T \mathbf{A} \mathbf{d}_j = 0$$

$$\mathbf{e}_{n+1} = \mathbf{e}_n + \delta_n \mathbf{d}_n = \mathbf{e}_n - \left(\frac{\mathbf{d}_n^T \mathbf{A} \mathbf{e}_0}{\mathbf{d}_n^T \mathbf{A} \mathbf{d}_n}\right) \mathbf{d}_n$$

$$\mathbf{e}_0 = \mathbf{x}_0 - \mathbf{x}^*$$

$$\mathbf{e}_0 = \sum_{i=0}^{N-1} c_i \mathbf{d}_i = \sum_{i=0}^{N-1} \mathbf{p}_{\mathbf{d}_i}(\mathbf{e}_0) = \sum_{i=0}^{N-1} \left(\frac{\mathbf{d}_i^T \mathbf{A} \mathbf{e}_0}{\mathbf{d}_i^T \mathbf{A} \mathbf{d}_i} \right) \mathbf{d}_i$$

$$\mathbf{p}_{\mathbf{d}_i}(\mathbf{e}_0)$$

$$\mathbf{p}_{\mathbf{d}_i}(\mathbf{e}_0) = c_i \mathbf{d}_i = \left(\frac{\mathbf{d}_i^T \mathbf{A} \mathbf{e}_0}{\mathbf{d}_i^T \mathbf{A} \mathbf{d}_i}\right) \mathbf{d}_i$$

$$c_i = \frac{\mathbf{d}_i^T \mathbf{A} \mathbf{e}_0}{\mathbf{d}_i^T \mathbf{A} \mathbf{d}_i} = -\delta_i$$

$$\mathbf{e}_{n+1} = \mathbf{e}_0 + \sum_{i=0}^{n} \delta_i \mathbf{d}_i = \sum_{i=0}^{N-1} c_i \mathbf{d}_i - \sum_{i=0}^{n} c_i \mathbf{d}_i = \sum_{i=n+1}^{N-1} c_i \mathbf{d}_i = \sum_{i=n+1}^{N-1} \mathbf{p}_{\mathbf{d}_i}(\mathbf{e}_0)$$

$$\mathbf{p}_{\mathbf{d}_n}(\mathbf{e}_0)$$

$$\mathbf{x}_N = \mathbf{x}^* + \mathbf{e}_N = \mathbf{x}^*$$

$$\mathbf{d}_k^T \mathbf{A} \ (k \le n)$$

$$\mathbf{d}_k^T \mathbf{A} \mathbf{e}_{n+1} = \mathbf{d}_k^T \mathbf{g}_{n+1} = \sum_{i=n+1}^{N-1} c_i \mathbf{d}_k^T \mathbf{A} \mathbf{d}_j = 0$$

$$\mathbf{d}_0, \cdots, \mathbf{d}_n$$

$$\{\mathbf{v}_0,\cdots,\mathbf{v}_{N-1}\}$$

$$\mathbf{d}_n = \mathbf{v}_n - \sum_{j=0}^{n-1} \mathbf{p}_{\mathbf{d}_j}(\mathbf{v}_n) = \mathbf{v}_n - \sum_{m=0}^{n-1} \left(\frac{\mathbf{d}_m^T \mathbf{A} \mathbf{v}_n}{\mathbf{d}_m^T \mathbf{A} \mathbf{d}_m} \right) \mathbf{d}_m = \mathbf{v}_n - \sum_{m=0}^{n-1} \beta_{nm} \mathbf{d}_m$$

$$\beta_{nm} = \mathbf{d}_m^T \mathbf{A} \mathbf{v}_n / (\mathbf{d}_m^T \mathbf{A} \mathbf{d}_m)$$

$$\beta_{nm}\mathbf{d}_m$$

$$\mathbf{v}_n = -\mathbf{g}_n$$

$$\mathbf{d}_n = -\mathbf{g}_n - \sum_{m=0}^{n-1} \beta_{nm} \mathbf{d}_m$$

$$\beta_{nm} = \frac{\mathbf{d}_m^T \mathbf{A} \mathbf{v}_n}{\mathbf{d}_m^T \mathbf{A} \mathbf{d}_m} = -\frac{\mathbf{d}_m^T \mathbf{A} \mathbf{g}_n}{\mathbf{d}_m^T \mathbf{A} \mathbf{d}_m} \qquad (m < n)$$

$$\mathbf{g}_n = \sum_{i=0}^n \alpha_i \mathbf{d}_i$$

 $\mathbf{g}_{n+1}\mathbf{g}_k = \mathbf{g}_{n+1}^T \left(\sum_{i=0}^k \alpha_i \mathbf{d}_i\right) = \sum_{i=0}^k \alpha_i \ \mathbf{g}_{n+1}^T \mathbf{d}_i$

$$\mathbf{g}_{n+1}^T \mathbf{d}_k = 0$$

 \mathbf{g}_0, \cdot $, \mathbf{g}_n$

$$\mathbf{g}_k^T (k \ge n)$$

$$\mathbf{g}_k^T \mathbf{d}_n = -\mathbf{g}_k^T \mathbf{g}_n - \sum_{m=0}^{n-1} \beta_{mn} \, \mathbf{g}_k^T \mathbf{d}_m = -\mathbf{g}_k^T \mathbf{g}_n = \begin{cases} -||\mathbf{g}_n||^2 & n=k \\ 0 & n < k \end{cases}$$

$$\mathbf{g}_k^T \mathbf{d}_m = 0$$

 $m < n \le k$

$$\mathbf{g}_n^T \mathbf{d}_n = -||\mathbf{g}_n||^2$$

$$\delta_n = -\frac{\mathbf{g}_n^T \mathbf{d}_n}{\mathbf{d}_n^T \mathbf{A} \mathbf{d}_n} = \frac{||\mathbf{g}_n||^2}{\mathbf{d}_n^T \mathbf{A} \mathbf{d}_n}$$

$$\mathbf{g}_{m+1} = \mathbf{A}\mathbf{x}_{m+1} - \mathbf{b} = \mathbf{A}(\mathbf{x}_m + \delta_m \mathbf{d}_m) - \mathbf{b} = (\mathbf{A}\mathbf{x}_m - \mathbf{b}) + \delta_m \mathbf{A}\mathbf{d}_m = \mathbf{g}_m + \delta_m \mathbf{A}\mathbf{d}_m$$

$$\mathbf{g}_n^T \mathbf{g}_{m+1} = \mathbf{g}_n^T \mathbf{g}_m + \delta_m \mathbf{g}_n^T \mathbf{A} \mathbf{d}_m = \delta_m \mathbf{g}_n^T \mathbf{A} \mathbf{d}_m$$

$$\mathbf{g}_n^T \mathbf{g}_m = 0$$

$$\mathbf{g}_n^T \mathbf{A} \mathbf{d}_m$$

$$\mathbf{g}_n^T \mathbf{A} \mathbf{d}_m = \frac{1}{\delta_m} \mathbf{g}_n^T \mathbf{g}_{m+1} = \begin{cases} & ||\mathbf{g}_n||^2 / \delta_{n-1} & m = n-1 \\ & 0 & m < n-1 \end{cases}$$

$$\beta_{nm} = -\frac{\mathbf{d}_m^T \mathbf{A} \mathbf{g}_n}{\mathbf{d}_m^T \mathbf{A} \mathbf{d}_m} = \begin{cases} -||\mathbf{g}_n||^2 / \delta_{n-1} \mathbf{d}_{n-1}^T \mathbf{A} \mathbf{d}_{n-1} & m = n-1 \\ 0 & m < n-1 \end{cases}$$

$$\delta_{n-1} = ||\mathbf{g}_{n-1}||^2 / \mathbf{d}_{n-1}^T \mathbf{A} \mathbf{d}_{n-1}$$

$$\beta_{nm} = \beta_m$$

$$\beta_n = -\frac{||\mathbf{g}_n||^2}{||\mathbf{g}_{n-1}||^2}$$

$$\mathbf{d}_0 = -\mathbf{g}_0$$

$$\delta_n = \frac{||\mathbf{g}_n||^2}{\mathbf{d}_n^T \mathbf{A} \mathbf{d}_n}, \quad \mathbf{x}_{n+1} = \mathbf{x}_n + \delta_n \mathbf{d}_n$$

$$\mathbf{g}_{n+1} = \frac{d}{d\mathbf{x}} f(\mathbf{x}_{n+1})$$

$$\beta_{n+1} = \frac{||\mathbf{g}_{n+1}||^2}{||\mathbf{g}_n||^2}$$

$$\mathbf{d}_{n+1} = -\mathbf{g}_{n+1} + \beta_{n+1}\mathbf{d}_n$$

 $f(x,y) = \mathbf{x}^T \mathbf{A} \mathbf{x} = [x_1, x_2] \begin{bmatrix} 3 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$

$$\mathbf{x}_0 = [1.5, -0.75]^T$$

\overline{n}	$\mathbf{x} = [x_1, x_2]$	$f(\mathbf{x})$
0	1.500000, -0.750000	2.812500
1	0.250000, -0.750000	0.468750e - 01
2	0.250000, -0.125000	7.812500e - 02
3	0.041667, -0.125000	1.302083e - 02
4	0.041667, -0.020833	2.170139e - 03
5	0.006944, -0.020833	3.616898e - 04
6	0.006944, -0.003472	6.028164e - 05
7	0.001157, -0.003472	1.004694e - 05
8	0.001157, -0.000579	1.674490e - 06
9	0.000193, -0.000579	2.790816e - 07
10	0.000193, -0.000096	4.651361e - 08
11	0.000032, -0.000096	7.752268e - 09
12	0.000032, -0.000016	1.292045e - 09
13	0.000005, -0.000016	2.153408e - 10

\overline{n}	$\mathbf{x} = [x_1, x_2]$	$f(\mathbf{x})$
0	1.500000, -0.750000	2.812500e + 00
1	0.250000, -0.750000	4.687500e - 01
2	0.000000, -0.000000	1.155558e - 33

$$f(\mathbf{x}) = \mathbf{x}^T \mathbf{A} \mathbf{x}$$

$$\mathbf{A} = \left[\begin{array}{rrr} 5 & 3 & 1 \\ 3 & 4 & 2 \\ 1 & 2 & 3 \end{array} \right]$$

$$\mathbf{x}_0 = [1, \ 2, \ 3]^T$$

 $\mathbf{v}_{41} = [3.5486e - 06, -7.4471e - 06, 4.6180e - 06]^T$

$$f(\mathbf{x}) = 8.5429e - 11$$

n	$\mathbf{x} = [x_1, x_2, x_3]$	$f(\mathbf{x})$
0	1.000000, 2.000000, 3.000000	4.500000e + 01
1	-0.734716, -0.106441, 1.265284	2.809225e + 00
2	0.123437, -0.209498, 0.136074	3.584736e - 02
3	-0.000000, 0.000000, 0.000000	3.949119e - 31

 $x_1 = 0.5, \ x_2 = 0, \ x_3 = -0.5$

\overline{n}	$\mathbf{x} = [x_1, x_2, x_3]$	$o(\mathbf{x})$
0	0.0000, 0.0000, 0.0000	1.032500e + 02
1	0.0113, 0.0050, -0.5001	3.160163e + 00
2	0.0188, -0.0021, -0.5004	3.095894e + 00
3	0.5009, -0.0018, -0.5004	7.268252e - 05
4	0.5009, -0.0017, -0.5000	1.051537e - 05
5	0.5008, -0.0012, -0.5000	6.511151e - 06
6	0.5001, -0.0005, -0.5000	6.365321e - 07
7	0.5001, -0.0005, -0.5000	5.667357e - 07
8	0.5002, -0.0004, -0.5000	2.675128e - 07
9	0.5001, -0.0003, -0.5000	1.344218e - 07
10	0.5001, -0.0002, -0.5000	1.241196e - 07
11	0.5000, -0.0001, -0.5000	2.120969e - 08
12	0.5000, -0.0001, -0.5000	1.541814e - 08
13	0.5000, -0.0001, -0.5000	7.282025e - 09
14	0.5000, -0.0001, -0.5000	4.801781e - 09
15	0.5000, -0.0000, -0.5000	4.463926e - 09

$$f(\mathbf{x}) = \mathbf{x}^T \mathbf{A} \mathbf{x} / 2 - \mathbf{b}^T \mathbf{x} + c$$

$$df(\mathbf{x})/d\mathbf{x} = \mathbf{A}\mathbf{x} - \mathbf{b} = \mathbf{0}$$

$$\mathbf{d}_i, \ (i=1,\cdots,N)$$

$$\mathbf{d}_i^T \mathbf{A} \mathbf{d}_j = 0 \ (i \neq j)$$

$$\mathbf{x} = \sum_{i=1}^{N} c_i \mathbf{d}_i$$

$$\mathbf{b} = \mathbf{A}\mathbf{x} = \mathbf{A}\left[\sum_{i=1}^{N} c_i \mathbf{d}_i\right] = \sum_{i=1}^{N} c_i \mathbf{A} \mathbf{d}_i$$

$$\mathbf{d}_j^T \mathbf{b} = \sum_{i=1}^N c_i \mathbf{d}_j^T \mathbf{A} \mathbf{d}_i = c_j \mathbf{d}_j^T \mathbf{A} \mathbf{d}_j$$

$$c_j = \frac{\mathbf{d}_j^T \mathbf{b}}{\mathbf{d}_j^T \mathbf{A} \mathbf{d}_j}$$

$$\mathbf{x} = \sum_{i=1}^{N} c_i \mathbf{d}_i = \sum_{i=1}^{N} \left(\frac{\mathbf{d}_i^T \mathbf{b}}{\mathbf{d}_i^T \mathbf{A} \mathbf{d}_i} \right) \mathbf{d}_i$$

$$\mathbf{p}_{\mathbf{d}_i}(\mathbf{x}) = \left(\frac{\mathbf{d}_i^T \mathbf{A} \mathbf{x}}{\mathbf{d}_i^T \mathbf{A} \mathbf{d}_i}\right) \mathbf{d}_i$$

R =M

$$\mathbf{A}^T \mathbf{A} \mathbf{x} = \mathbf{A}^T \mathbf{b}$$

$$\Delta E = f(\mathbf{x}_{n+1}) - f(\mathbf{x}_n)_0$$

$$P(\Delta E, T) = e^{-\Delta E/T} > Th$$

$$\mathbf{x} = [x_1, \cdots, x_N]^T$$

optimize
$$f(\mathbf{x}) = f(x_1, \dots, x_N)$$
 subject to:
$$\begin{cases} h_i(\mathbf{x}) = 0, & (i = 1, \dots m) \\ g_j(\mathbf{x}) \leq 0, & (j = 1, \dots n) \end{cases}$$

$$g_i(\mathbf{x})$$

$$h_j(\mathbf{x})$$

maximize/minimize
$$f(\mathbf{x}) = f(x_1, \dots, x_N)$$

subject to: $h_i(\mathbf{x}) = h_i(x_1, \dots, x_N) = 0, \quad (i = 1, \dots, m)$

$$h_i(\mathbf{x}) = 0$$

$$h(\mathbf{x}) = 0$$

$$f(\mathbf{x}) = d$$

$$\nabla_{\mathbf{x}} f(\mathbf{x}^*) = \lambda^* \ \nabla_{\mathbf{x}} \ h(\mathbf{x}^*)$$

$$L(\mathbf{x}, \lambda) = f(\mathbf{x}) - \lambda h(\mathbf{x})$$

$$\nabla_{\mathbf{x},\lambda} L(\mathbf{x},\lambda) = \nabla_{\mathbf{x},\lambda} [f(\mathbf{x}) - \lambda h(\mathbf{x})] = \mathbf{0}$$

$$\begin{cases} \nabla_{\mathbf{x}} f(\mathbf{x}) = \lambda \nabla_{\mathbf{x}} h(\mathbf{x}) \\ \nabla_{\lambda} L(\mathbf{x}, \lambda) = \partial L(\mathbf{x}, \lambda) / \partial \lambda = -h(\mathbf{x}) = 0 \end{cases}$$

$$N + m = 2 + 1 = 3$$

$$\frac{\partial f(\mathbf{x})}{\partial x_i} = \lambda \frac{\partial h(\mathbf{x})}{\partial x_i}$$
 $(i = 1, 2), \quad h(\mathbf{x}) = 0$

$$\mathbf{x}^* = [x_1^*, x_2^*]^T$$

$$\nabla_{\mathbf{x}} f(\mathbf{x}^*) = \lambda^* \nabla_{\mathbf{x}} h(\mathbf{x}^*)$$

$$h(\mathbf{x}^*) = 0$$

$$\lambda^* = 0$$

$$\nabla_{\mathbf{x}} f(\mathbf{x}^*) = \lambda^* \nabla_{\mathbf{x}} h(\mathbf{x}^*) = 0$$

$$\lambda^* \neq 0$$

$$\nabla_{\mathbf{x}} f(\mathbf{x}^*) \neq 0$$

$$\nabla_{\mathbf{x}} h(\mathbf{x}^*) \neq \mathbf{0}$$

$$h_i(\mathbf{x}) = 0 \ (i = 1, \cdots, m)$$

$$h_1(\mathbf{x}^*) = \dots = h_m(\mathbf{x}^*) = 0$$

$$f(\mathbf{x}^*) = d$$

$$\nabla f(\mathbf{x}^*)$$

$$\nabla h_i(\mathbf{x}^*)$$

$$\nabla_{\mathbf{x}} f(\mathbf{x}^*) = \sum_{i=1}^m \lambda_i^* \nabla_{\mathbf{x}} h_i(\mathbf{x}^*)$$

$$L(\mathbf{x}, \lambda) = f(\mathbf{x}) - \sum_{i=1}^{m} \lambda_i h_i(\mathbf{x})$$

$$\lambda = [\lambda_1, \cdots, \lambda_m]^T$$

$$\nabla_{\mathbf{x},\lambda} L(\mathbf{x},\lambda) = \nabla_{\mathbf{x},\lambda} \left[f(\mathbf{x}) - \sum_{i=1}^{m} \lambda_i h_i(\mathbf{x}) \right] = \mathbf{0}$$

$$\nabla_{\mathbf{x}} f(\mathbf{x}) = \sum_{i=1}^{m} \lambda_i \nabla_{\mathbf{x}} h_i(\mathbf{x})$$

$$\frac{\partial f(\mathbf{x})}{\partial x_j} = \sum_{i=1}^m \lambda_i \frac{\partial h_i(\mathbf{x})}{\partial x_j} \quad (j = 1, \dots, N),$$

$$\frac{\partial L(\mathbf{x}, \lambda)}{\partial \lambda_i} = h_i(\mathbf{x}) = 0 \quad (i = 1, \dots, m)$$

$$\nabla_{\mathbf{x}} f(\mathbf{x}^*)$$

$$\nabla_{\mathbf{x}} h_i(\mathbf{x}^*)$$

 λ_1 , $, \wedge_m$

 $i=1,\cdots,m$

$$\nabla_{\mathbf{x}} f(\mathbf{x}^*) = \sum_{i=1}^m \lambda_i \nabla_{\mathbf{x}} h_i(\mathbf{x}) = \mathbf{0}$$

maximize/minimize
$$f(\mathbf{x}) = f(x_1, \dots, x_N)$$

subject to: $g_j(\mathbf{x}) = g_j(x_1, \dots, x_N) \le \text{ or } \ge 0, \qquad (j = 1, \dots, n)$

$$g(\mathbf{x}) > 0$$

$$\mathbf{x}^* = [x_1^*, x_2^*]^T$$

$$g(x_1^*, x_2^*) = 0$$

$$\nabla_{\mathbf{x}} f(\mathbf{x}) = 0$$

$$g(\mathbf{x}^*) = 0$$

$$\nabla_{\mathbf{x}} f(\mathbf{x}^*) \neq \mathbf{0}$$

$$\nabla_{\mathbf{x}} f(\mathbf{x}^*) = \mu^* \ \nabla_{\mathbf{x}} g(\mathbf{x}^*)$$

$$\nabla_{\mathbf{x}} f(\mathbf{x}^*) - \mu^* \ \nabla_{\mathbf{x}} g(\mathbf{x}^*) = 0$$

$$g(\mathbf{x}) \ge 0$$

$$g(\mathbf{x}) \le 0$$

	$g(\mathbf{x}) \ge 0$	$g(\mathbf{x}) \le 0$
$\max f(\mathbf{x})$		$\nabla f(\mathbf{x}_3) = \mu \nabla g(\mathbf{x}_3), \mu > 0$
$\min f(\mathbf{x})$	$\nabla f(\mathbf{x}_2) = \mu \nabla g(\mathbf{x}_2), \ \mu > 0$	$\nabla f(\mathbf{x}_4) = \mu \nabla g(\mathbf{x}_4), \ \mu < 0$

$$\mu g(\mathbf{x}) \begin{cases} \leq 0 & \text{for maximization} \\ \geq 0 & \text{for minimization} \end{cases}$$

$$\nabla_{\mathbf{x}} f(\mathbf{x}^*) = \mathbf{0}$$

$$g(\mathbf{x}^*) \neq 0$$

$$\nabla_{\mathbf{x}} f(\mathbf{x}) = \mu \ \nabla_{\mathbf{x}} g(\mathbf{x}) = \mathbf{0}$$

$$\mu^* \neq 0$$

$$\mu^* = 0$$

$$\mu^* g(\mathbf{x}^*) = 0$$

$$g_i(\mathbf{x}) = 0 \ (i = 1, \cdots, n)$$

$$L(\mathbf{x}, \mu) = f(\mathbf{x}) - \sum_{i=1}^{n} \mu_i g_i(\mathbf{x})$$

$$\nabla_{\mathbf{x},\mu} L(\mathbf{x},\mu) = \nabla_{\mathbf{x},\mu} \left[f(\mathbf{x}) - \sum_{i=1}^{n} \mu_i g_i(\mathbf{x}) \right] = \mathbf{0}$$

$$\nabla_{\mathbf{x}} f(\mathbf{x}) = \sum_{i=1}^{n} \mu_i \, \nabla_{\mathbf{x}} \, g_i(\mathbf{x})$$

$$\frac{\partial L(\mathbf{x}, \mu)}{\partial \mu_i} = g_i(\mathbf{x}) = 0 \quad (i = 1, \dots, n)$$

$$g(\mathbf{x}^*)$$

maximize/minimize
$$f(\mathbf{x}) = f(x_1, \dots, x_N)$$

subject to:
$$\begin{cases} h_i(\mathbf{x}) = 0, & (i = 1, \dots m) \\ g_j(\mathbf{x}) \le 0 \text{ or } g_j(\mathbf{x}) \ge 0, & (j = 1, \dots n) \end{cases}$$

maximize/minimize	$f(\mathbf{x})$
subject to:	$\left\{ egin{array}{l} \mathbf{h}(\mathbf{x}) = 0 \\ \mathbf{g}(\mathbf{x}) \leq 0 \ \ \mathrm{or} \ \ \mathbf{g}(\mathbf{x}) \geq 0 \end{array} ight.$

$$\mathbf{h}(\mathbf{x}) = [h_1(\mathbf{x}), \cdots, h_m(\mathbf{x})]^T$$

$$\mathbf{g}(\mathbf{x}) = [g_1(\mathbf{x}), \cdots, g_n(\mathbf{x})]^T$$

$$L(\mathbf{x}, \lambda, \mu) = f(\mathbf{x}) - \sum_{i=1}^{m} \lambda_i h_i(\mathbf{x}) - \sum_{j=1}^{n} \mu_j g_j(\mathbf{x}) = f(\mathbf{x}) - \lambda^T \mathbf{h}(\mathbf{x}) - \mu^T \mathbf{g}(\mathbf{x})$$

$$\mu = [\mu_1, \cdots, \mu_n]^T$$

$$g_j(\mathbf{x}) \le 0$$

$$g_j(\mathbf{x}) \ge 0$$

$$\mu_j g_j(\mathbf{x}^*) = 0, \qquad (j = 1, \dots, n)$$

$$f(x_1, x_2) = x_1^2 + x_2^2$$

$$x_1 + x_2 = 1$$

$$x_1 + x_2 \le 1$$

$$x_1 + x_2 \ge 1$$

$$h(x_1, x_2) = x_1 + x_2 - 1 = 0$$

minimize/maximize: $f(x_1, x_2) = x_1^2 + x_2^2$ subject to: $h(x_1, x_2) = x_1 + x_2 - 1 = 0$

$$L(x_1, x_2, \lambda) = f(x_1, x_2) - \lambda g(x_1, x_2) = x_1^2 + x_2^2 - \lambda (x_1 + x_2 - 1)$$

$$\frac{\partial L}{\partial x_1} = 2x_1 - \lambda = 0, \quad \frac{\partial L}{\partial x_2} = 2x_2 - \lambda = 0, \quad \frac{\partial L}{\partial \lambda} = x_1 + x_2 - 1 = 0$$

$$x_1^* = x_2^* = 0.5$$

$$x_1^* + x_2^* = 1$$

$$f(0.5, 0.5) = 0.5$$

$$f(x_1,x_2)$$

$$g(x_1,x_2)$$

$$\nabla f(0.5, 0.5) = \nabla h(0.5, 0.5) = [1, 1]^T$$

minimize:
$$f(x_1, x_2) = x_1^2 + x_2^2$$

subject to: $g(x_1, x_2) = x_1 + x_2 - 1 \ge 0$

maximize:
$$f(x_1, x_2) = x_1^2 + x_2^2$$

subject to: $g(x_1, x_2) = x_1 + x_2 - 1 \le 0$

$$L(x_1, x_2, \mu) = f(x_1, x_2) - \mu g(x_1, x_2) = x_1^2 + x_2^2 - \mu(x_1 + x_2 - 1)$$

$$\frac{\partial L}{\partial x_1} = 2x_1 - \mu = 0, \quad \frac{\partial L}{\partial x_2} = 2x_2 - \mu = 0, \quad \frac{\partial L}{\partial \mu} = x_1 + x_2 - 1 = 0$$

$$\mu^* = 1 > 0$$

$$x_1^* = x_2^* = \mu^*/2 = 0.5$$

minimize:
$$f(x_1, x_2) = x_1^2 + x_2^2$$

subject to: $g(x_1, x_2) = x_1 + x_2 - 1 \le 0$

maximize:
$$f(x_1, x_2) = x_1^2 + x_2^2$$

subject to: $g(x_1, x_2) = x_1 + x_2 - 1 \ge 0$

$$\frac{\partial L}{\partial x_1} = 2x_1 = 0, \quad \frac{\partial L}{\partial x_2} = 2x_2 = 0$$

$$x_1^* = x_2^* = 0$$

$$\begin{array}{ll} \text{minimize} & f_p(\mathbf{x}) \\ \text{subject to:} & \left\{ \begin{array}{ll} \mathbf{h}(\mathbf{x}) = \mathbf{0} \\ \mathbf{g}(\mathbf{x}) \leq \mathbf{0} \ \text{or} \ \mathbf{g}(\mathbf{x}) \geq \mathbf{0} \end{array} \right. \end{array}$$

$$L(\mathbf{x}, \lambda, \mu) = f_p(\mathbf{x}) - \lambda^T \mathbf{h}(\mathbf{x}) - \mu^T \mathbf{g}(\mathbf{x})$$

$$\mu^T g(\mathbf{x}) \ge 0$$

$$f_d(\lambda, \mu) = \inf_{\mathbf{x}} L(\mathbf{x}, \lambda, \mu) = \inf_{\mathbf{x}} \left[f_p(\mathbf{x}) - \lambda^T \mathbf{h}(\mathbf{x}) - \mu^T \mathbf{g}(\mathbf{x}) \right]$$

$$\nabla_{\mathbf{x}} L(\mathbf{x}, \lambda, \mu) = \mathbf{0}$$

$$f_d(\lambda,\mu)$$

maximize $f_d(\lambda, \mu) = \inf_{\mathbf{x}} L(\mathbf{x}, \lambda, \mu)$ subject to: $\mu \leq \mathbf{0} \text{ or } \mu \geq \mathbf{0}$

$$\mu^T \mathbf{g}(\mathbf{x}) \ge 0$$

$$\mathbf{h}(\mathbf{x}) = \mathbf{0}$$

$$g(x) \ge 0$$

$$f_d(\lambda, \mu) \le L(\mathbf{x}, \lambda, \mu) = f_p(\mathbf{x}) - \lambda^T \mathbf{h}(\mathbf{x}) - \mu^T \mathbf{g}(\mathbf{x}) \le f_p(\mathbf{x}) - \mu^T \mathbf{g}(\mathbf{x}) \le f_p(\mathbf{x})$$

$$d^* = f_d(\lambda^*, \mu^*)$$

$$f_p(\mathbf{x}^*) = p^*$$

$$d^* = f_d(\lambda^*, \mu^*) \le L(\mathbf{x}, \lambda, \mu) \le f_p(\mathbf{x}^*) = p^*$$

$$p^* = f(\mathbf{x}^*) \ge f(\mathbf{x})$$

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$$d^* = f_d(\lambda^*, \ \mu^*) \ge f_p(\mathbf{x}^*) = p^*$$

$$p^* - d^* \ge 0$$

$$d^* - p^* \ge 0$$

$$d^* \neq p^*$$

$$f_d(\lambda^*, \mu^*) = L(\mathbf{x}^*, \lambda^*, \mu^*) = f_p(\mathbf{x}^*) - \lambda^{*T} \mathbf{h}(\mathbf{x}^*) - \mu^{*T} \mathbf{g}(\mathbf{x}^*)$$

$$f_p(\mathbf{x}^*) - \mu^{*T}\mathbf{g}(\mathbf{x}^*) = f_p(\mathbf{x}^*) = p^*$$

$$\begin{cases} \nabla_{\mathbf{x}} L(\mathbf{x}^*, \lambda, \mu) = \nabla_{\mathbf{x}} f(\mathbf{x}^*) - \sum_{i=1}^m \lambda_i \nabla_{\mathbf{x}} h_i(\mathbf{x}^*) - \sum_{j=1}^n \mu_j \nabla_{\mathbf{x}} g_j(\mathbf{x}^*) = \mathbf{0} & \text{(stationarity)} \\ h_i(\mathbf{x}^*) = 0, & (i = 1, \dots, m) & \text{(primal feasibility)} \\ g_j(\mathbf{x}^*) \le 0 & \text{or } g_j(\mathbf{x}^*) \ge 0, & (j = 1, \dots, n) \\ \mu_j^* \le 0, & \text{or } \mu_j^* \ge 0, & (j = 1, \dots, n) \\ \mu_j^* g_j(\mathbf{x}^*) = 0, & (j = 1, \dots, n) & \text{(dual feasibility)} \end{cases}$$

$$\text{(complementarity)}$$

$$g_j(\mathbf{x}^*)$$

$$g_j(\mathbf{x}^*) \neq 0$$

$$\mu_j^* = 0$$

$$\mu_j^* \neq 0$$

$$g_j(\mathbf{x}^*) = 0$$

$$f(x_1, \cdots, x_N) = \sum_{i=1}^{N} c_i x_i$$

 x_1, \cdots $,x_N$

$$f'(x_1, \dots, x_N) = -\sum_{i=1}^{N} c_i x_i$$

maximize
$$f(x_1, \dots, x_N) = \sum_{i=1}^{N} c_i x_i$$
 subject to:
$$\begin{cases} \sum_{i=1}^{n} a_{1i} x_i \leq b_1 \\ \dots \\ \sum_{i=1}^{N} a_{Mi} x_i \leq b_M \\ x_1 \geq 0, \dots, x_N \geq 0 \end{cases}$$

 c_1, \cdots, c_N

$$f(x_1,\cdots,x_N)$$

 b_1 , $,b_{M}$

maximize
$$f(x, y, z) = 2x - y + 3z$$

subject to:
$$\begin{cases} x - 2y + z = 3\\ 3x - y + 4z = 10\\ y \ge 0, \ z \ge 0 \end{cases}$$

$$x = 3 + 2y - z$$

$$2x - y + z = 3y + z + 6$$

$$\begin{array}{ll} \text{maximize} & f(y,z) = 3y + z + 6 \\ \text{subject to:} & \begin{cases} 5y + z = 1 \\ y \geq 0, \ z \geq 0 \end{cases} \end{array}$$

maximize
$$\mathbf{c}^T \mathbf{x}$$
 subject to
$$\begin{cases} \mathbf{A} \mathbf{x} - \mathbf{b} \leq \mathbf{0} \\ \mathbf{x} \geq \mathbf{0} \end{cases}$$

$$\mathbf{x} = \left[egin{array}{c} x_1 \ dots \ x_N \end{array}
ight] \qquad \mathbf{c} = \left[egin{array}{c} c_1 \ dots \ c_N \end{array}
ight] \qquad \mathbf{b} = \left[egin{array}{c} b_1 \ dots \ b_M \end{array}
ight] \qquad \mathbf{A} = \left[egin{array}{c} a_{11} & \cdots & a_{1N} \ dots & \ddots & dots \ a_{M1} & \cdots & a_{MN} \end{array}
ight]$$

$$L(\mathbf{x}) = \mathbf{c}^T \mathbf{x} - \mathbf{y}^T (\mathbf{A} \mathbf{x} - \mathbf{b}) = (\mathbf{c} - \mathbf{A}^T \mathbf{y})^T \mathbf{x} + \mathbf{y}^T \mathbf{b}$$

$$\mathbf{y} = [y_1, \cdots, y_M]^T$$

$$Ax - b \le 0$$

$$\mathbf{y} \geq \mathbf{0}$$

$$f_d(\mathbf{y}) = \max_{\mathbf{x}} L(\mathbf{x}) = \max_{\mathbf{x}} [\mathbf{c}^T \mathbf{x} - \mathbf{y}^T (\mathbf{A} \mathbf{x} - \mathbf{b})] = \max_{\mathbf{x}} [(\mathbf{c} - \mathbf{A}^T \mathbf{y})^T \mathbf{x} + \mathbf{y}^T \mathbf{b}]$$

$$\nabla_{\mathbf{x}} L(\mathbf{x}) = \nabla_{\mathbf{x}} [(\mathbf{c} - \mathbf{A}^T \mathbf{y})^T \mathbf{x} + \mathbf{y}^T \mathbf{b}] = \mathbf{c} - \mathbf{A}^T \mathbf{y} = \mathbf{0}$$

$$L_d(\mathbf{y}) = \max_{\mathbf{x}} L(\mathbf{x})$$

$$f_d(\mathbf{y}) = \mathbf{y}^T \mathbf{b}$$

$$\mathbf{c} - \mathbf{A}^T \mathbf{y} \le 0$$

$$(\mathbf{c} - \mathbf{A}^T \mathbf{y})^T \mathbf{x} \le 0$$

$$f_d(\mathbf{y})$$

$$f_d(\mathbf{y}) = \mathbf{b}^T \mathbf{y} \ge L(\mathbf{x}) = \mathbf{c}^T \mathbf{x} - \mathbf{y}^T (\mathbf{A} \mathbf{x} - \mathbf{b}) \ge \mathbf{c}^T \mathbf{x} = f_p(\mathbf{x})$$

minimize
$$f_d(\mathbf{y}) = \mathbf{b}^T \mathbf{y}$$

subject to
$$\begin{cases} \mathbf{A}^T \mathbf{y} - \mathbf{c} \ge \mathbf{0} \\ \mathbf{y} \ge \mathbf{0} \end{cases}$$

$$\begin{cases} \max : & \mathbf{c}^T \mathbf{x} \\ \text{s.t.} & \mathbf{A} \mathbf{x} - \mathbf{b} \le \mathbf{0}, \ \mathbf{x} \ge \mathbf{0} \end{cases} \iff \begin{cases} \min & \mathbf{b}^T \mathbf{y} \\ \text{s.t.} & \mathbf{A} \mathbf{y} - \mathbf{c} \ge \mathbf{0}, \ \mathbf{y} \ge \mathbf{0} \end{cases}$$

$$\begin{cases} \min: & \mathbf{c}^T \mathbf{x} \\ \text{s.t.} & \mathbf{A} \mathbf{x} - \mathbf{b} \leq \mathbf{0}, \ \mathbf{x} \geq \mathbf{0} \end{cases} \iff \begin{cases} \max & \mathbf{b}^T \mathbf{y} \\ \text{s.t.} & \mathbf{A} \mathbf{y} - \mathbf{c} \leq \mathbf{0}, \ \mathbf{y} \leq \mathbf{0} \end{cases} \iff \begin{cases} \max & -\mathbf{b}^T \mathbf{y} \\ \text{s.t.} & \mathbf{A} \mathbf{y} + \mathbf{c} \geq \mathbf{0}, \ \mathbf{y} \geq \mathbf{0} \end{cases}$$

$$\mathbf{s} \geq \mathbf{0}$$

minimize
$$\mathbf{b}^T \mathbf{y}$$

subject to
$$\begin{cases} \mathbf{A}^T \mathbf{y} - \mathbf{s} = \mathbf{c} \\ \mathbf{s} \ge \mathbf{0} \end{cases}$$

$$\mathbf{b} \geq \mathbf{A}\mathbf{x}$$

$$\mathbf{c}^T \mathbf{x} \le (\mathbf{A}^T \mathbf{y})^T \mathbf{x} = \mathbf{y}^T \mathbf{A} \mathbf{x} = \mathbf{x}^T \mathbf{A}^T \mathbf{y} = (\mathbf{A} \mathbf{x})^T \mathbf{y} \le \mathbf{b}^T \mathbf{y}$$

$$\mathbf{b}^T \mathbf{y} \ge \mathbf{c}^T \mathbf{x}$$

$$\sum_{i=1}^{N} a_i x_i \le b \implies \sum_{i=1}^{N} a_i x_i + s = b, \quad (s \ge 0)$$

maximize
$$z = f(\mathbf{x}) = \mathbf{c}^T \mathbf{x} = \sum_{j=1}^N c_j x_j$$

subject to:
$$\begin{cases} \sum_{i=1}^N a_{1i} x_i & +s_1 & = b_1 \\ \sum_{i=1}^N a_{2i} x_i & +s_2 & = b_2 \\ \cdots & \cdots & \cdots & \cdots \\ \sum_{i=1}^N a_{Mi} x_i & +s_M & = b_M \\ x_1 \ge 0, & \cdots, & x_N \ge 0, & s_1 \ge 0, & \cdots & s_M \ge 0 \end{cases}$$

$$\{s_1,\cdots,s_M\}$$

$$\{x_1,\cdots,x_N\}$$

$$\mathbf{x} = [x_1, x_2, \cdots, x_N, s_1, s_2, \cdots, s_M]^T$$

$$M \times (N+M)$$

$$\mathbf{A} = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1N} & 1 & 0 & \cdots & 0 \\ a_{21} & a_{22} & \cdots & a_{1N} & 0 & 1 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \ddots & \vdots \\ a_{M1} & a_{M2} & \cdots & a_{MN} & 0 & \cdots & 0 & 1 \end{bmatrix}_{M \times (N+M)} = \begin{bmatrix} \mathbf{A}_{M \times N} \mid \mathbf{I}_{M \times M} \end{bmatrix}$$

 $\mathbf{A}_{M \times N}$

maximize	$\mathbf{c}^T \mathbf{x}$		maximize	$\mathbf{c}^T \mathbf{x}$
subject to	$\left\{ egin{array}{l} \mathbf{A}\mathbf{x}+\mathbf{s}=\mathbf{b} \ \mathbf{x}\geq0,\ \mathbf{s}\geq0 \end{array} ight.$	or	subject to	$\left\{ egin{array}{l} \mathbf{A}\mathbf{x} = \mathbf{b} \ \mathbf{x} \geq 0 \end{array} ight.$

$$f(\mathbf{x}) = \mathbf{c}^T \mathbf{x}$$

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$$\mathbf{a}_j = [a_{i1}, \cdots, a_{IN}]^T$$

$$\sum_{j=1}^{N} a_{ij} x_j = \mathbf{a}_j^T \mathbf{x} = b_i, \qquad (i = 1, \dots, M)$$

$$\mathbf{x} \geq \mathbf{0}$$



$$C_{M+N}^{N} = \frac{(M+N)!}{N! \ M!}$$

$$C_{M+N}^N$$

$$\mathbf{x}^* \in P$$

$$\mathbf{x}' = \mathbf{x}^* + \epsilon |\mathbf{c}/||\mathbf{c}|| \in P$$

$$f(\mathbf{x}') = \mathbf{c}^T \mathbf{x}' = \mathbf{c}^T (\mathbf{x}^* + \epsilon \mathbf{c}/||\mathbf{c}||) = \mathbf{c}^T \mathbf{x}^* + \epsilon \mathbf{c}^T \mathbf{c}/||\mathbf{c}|| = \mathbf{c}^T \mathbf{x}^* + \epsilon ||\mathbf{c}|| > \mathbf{c}^T \mathbf{x}^* = f(\mathbf{x}^*)$$

$$C_{N+M}^N$$

maximize
$$f_p(x_1, x_2) = 2x_1 + 3x_2$$

subject to:
$$\begin{cases} 2x_1 + x_2 \le 18 & \text{maximize} & f_p(\mathbf{x}) = \mathbf{c}^T \mathbf{x} \\ 6x_1 + 5x_2 \le 60 & \text{or} \\ 2x_1 + 5x_2 \le 40 \\ x_1 \ge 0, & x_2 \ge 0 \end{cases}$$
 subject to:
$$\begin{cases} \mathbf{A} \mathbf{x} \le \mathbf{b} \\ \mathbf{x} \ge \mathbf{0} \end{cases}$$

$$\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}, \quad \mathbf{c} = \begin{bmatrix} 2 \\ 3 \end{bmatrix}, \quad \mathbf{b} = \begin{bmatrix} 18 \\ 60 \\ 40 \end{bmatrix}, \quad \mathbf{A} = \begin{bmatrix} 2 & 1 \\ 6 & 5 \\ 2 & 5 \end{bmatrix}$$

$$C_{N+M}^N = C_5^2 = 10$$

$$f_p(x_1, x_2) = \mathbf{c}^T \mathbf{x} = c_1 x_1 + c_2 x_2$$

$$\mathbf{x} = [x_1, \ x_2]^T$$

$$f_p(x_1, x_2) = 2x_1 + 3x_2 = 2 \times 5 + 3 \times 6 = 28$$

$$\mathbf{c}^T \mathbf{x}^* / ||\mathbf{c}|| = 28/3.61 = 7.77$$

	(x_1, x_2)		objective function (normalized)
1	5.00, 6.00	feasible	28 (7.77)
2	6.25, 5.50	infeasible	29 (8.04)
3	7.50, 3.00	feasible	24 (6.66)
4	20.00, 0.00	infeasible	40 (11.1)
5	10.00, 0.00	infeasible	20(5.55)
6	9.00, 0.00	feasible	18 (4.10)
7	0.00, 8.00	feasible	24 (6.66)
8	0.00, 12.00	infeasible	36 (9.98)
9	0.00, 18.00	infeasible	54 (14.99)
10	0.00, 0.00	feasible	0 (0.00)

maximize
$$f_p(x_1, x_2) = 2x_1 + 3x_2$$
 maximize $f_p(x_1, x_2) = 2x_1 + 3x_2$
subject to:
$$\begin{cases} 2x_1 + x_2 \le 18 \\ 6x_1 + 5x_2 \le 60 \\ 2x_1 + 5x_2 \le 40 \\ x_1 \ge 0, \ x_2 \ge 0 \end{cases} \implies \text{subject to:} \begin{cases} 2x_1 + x_2 + s_1 = 18 \\ 6x_1 + 5x_2 + s_2 = 60 \\ 2x_1 + 5x_2 + s_3 = 40 \\ x_1 \ge 0, \ x_2 \ge 0, \ s_1 \ge 0, \ s_2 \ge 0, \ s_3 \ge 0 \end{cases}$$

$$\mathbf{c}^T \mathbf{x} = z$$

minimize
$$f_d(y_1, y_2, y_3) = 18y_1 + 60y_2 + 40y_3$$

subject to:
$$\begin{cases} 2y_1 + 6y_2 + 2y_3 \ge 2 \\ y_1 + 5y_2 + 5y_3 \ge 3 \\ y_1 \ge 0, \ y_2 \ge 0, \ y_3 \ge 0 \end{cases}$$
 or minimize $f_d(\mathbf{y}) = \mathbf{b}^T \mathbf{y}$
subject to:
$$\begin{cases} \mathbf{A}^T \mathbf{y} \le \mathbf{b} \\ \mathbf{y} \ge \mathbf{0} \end{cases}$$

	(y_1, y_2, y_3)	objective function	
1	(1,0,0)	infeasible	18
2	(3,0,0)	feasible	54
3	(0, 1/3, 0)	infeasible	20
4	(0, 3/5, 0)	feasible	36
5	(0, 0, 1)	feasible	40
6	(0,0,3/5)	infeasible	24
7	(0, 1/5, 2/5)	feasible	28
8	(1/2, 0, 1/2)	feasible	29
9	(-2, 1, 0)	infeasible	24
10	(0,0,0)	infeasible	0

(0, 1/5, 2/5)

$$f_d(y_1, y_2, y_3) = 18y_1 + 60y_2 + 40y_3$$

$$f_p(x_1, x_2) = 2x_1 + 3x_2$$

$$\mathbf{A}_{M\times N}\mathbf{x}_{N\times 1}\leq \mathbf{b}_{M\times 1}$$

$$\mathbf{A}\mathbf{x} = [\mathbf{A}_{M \times N} \mid \mathbf{I}_{M \times M}]\mathbf{x} = \mathbf{b}$$

$$rank(\mathbf{A}) = M$$

$$\mathbf{A}\mathbf{x} = [\mathbf{A}_n \ \mathbf{I}] \left[egin{array}{c} \mathbf{x}_n \ \mathbf{x}_b \end{array}
ight] = \mathbf{A}_n \mathbf{x}_n + \mathbf{I} \ \mathbf{x}_b = \mathbf{b}$$

$$\mathbf{x}_n = \mathbf{0}$$

$$\mathbf{x}_b = \mathbf{b}$$

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$$\{\mathbf{e}_1,\cdots,\mathbf{e}_M\}$$

$$C_{N+M}^{M}$$

$$\sum_{j=1}^{N} a_{ij} x_j + s_i = b_i$$

$$i=1,\cdots,M$$

$$(M+1)st$$

$$f(\mathbf{x}) = \sum_{j=1}^{N} c_j x_j$$

$$x_1 = \dots = x_N = 0$$

$$z - c_1 x_1 - \dots - c_N x_N = 0$$

$$(N+M+1)st$$

$$\mathbf{b} = [b_1, \cdots, b_M]^T$$

$$z = f(\mathbf{x})$$

basic variables	x_1	x_2		x_N	s_1	s_2		s_M	basic solution
s_1	a_{11}	a_{12}		a_{1N}	1	0		0	b_1
s_2	a_{21}	a_{22}	• • •	a_{2N}	0	1	• • •	0	b_2
:	:	:	٠.	:	:	:	٠	:	<u>:</u>
s_M	a_{M1}	a_{M2}		a_{MN}	0	0		1	b_M
z	$-c_1$	$-c_2$		$-c_M$	0	0		0	0

$$\mathbf{x}_n = [x_1, \cdots, x_N]^T$$

$$\mathbf{x}_b = [s_1, \cdots, s_M]^T$$

$$\mathbf{x}_b = [s_1, \cdots, s_M]^T = \mathbf{b}, \quad \mathbf{x}_n = [x_1, \cdots, x_N]^T = \mathbf{0}$$

$$\mathbf{A}\mathbf{x} = \mathbf{A}_n\mathbf{x}_n + \mathbf{I}\mathbf{x}_b = \mathbf{b}$$

$$c_j = \max\{c_1, \cdots, c_N\}$$

$$z = \sum_{j=1}^{N} c_j x_j$$

$$x_k \ (k \neq j)$$

$$\sum_{j=1}^{N} a_{kj} x_j \le b_k$$

$$x_j \le b_k / a_{kj}$$

$$b_i/a_{ij} \le b_k/a_{kj}$$

$$k=1,\cdots,M$$



$$\mathbf{r}_i \leftarrow \mathbf{r}_i/a_{ij}$$

$$\mathbf{r}_k \leftarrow \mathbf{r}_k - a_{kj}\mathbf{r}_i$$

$$C_{M+N}^M = C_{10}^3 = 10$$

	x_1	x_2	s_1	s_2	s_3		objective function (normalized)
1	5	6	2	0	0	feasible	28 (7.77)
2	6.25	5.5	0	-5	0	infeasible	29 (8.04)
3	7.5	3	0	0	10	feasible	24 (6.66)
4	20	0	-2	-60	20	infeasible	40 (11.1)
5	10	0	-2	0	20	infeasible	20 (5.55)
6	9	0	0	6	22	feasible	18 (4.10)
7	0	8	10	20	0	feasible	24 (6.66)
8	0	12	6	0	-20	infeasible	36 (9.98)
9	0	18	0	-30	-50	infeasible	54 (14.99)
10	0	0	18	60	40	feasible	0 (0.00)

maximize
$$f(\mathbf{x}) = 2x_1 + 3x_2$$
 subject to:
$$\begin{cases} 2x_1 + x_2 + s_1 = 18 \\ 6x_1 + 5x_2 + s_2 = 60 \\ 2x_1 + 5x_2 + s_3 = 40 \\ x_1 \ge 0, \ x_2 \ge 0, \ s_1 \ge 0, \ s_2 \ge 0, \ s_3 \ge 0 \end{cases}$$

	x_1	x_2	s_1	s_2	s_3	b
s_1	2	1	1	0	0	18
s_2	6	5	0	1	0	60
s_3	2	5	0	0	1	40
z	-2	-3	0	0	0	0

$$x_1 = x_2 = 0$$

$$\mathbf{r}_i = \mathbf{r}_3 = [2, 5, 0, 0, 1, 40]$$

$$a_{ij} = a_{32} = 5$$

$$\mathbf{r}_3 = [0.4, 1, 0, 0, 0.2, 8]$$

 $a_{ki} = a_{k2}$

$$a_{kj} = a_{k2} = 0, (k = 1, 2)$$

 $a_{4j}\mathbf{r}_i = a_{42}\mathbf{r}_3$

	x_1	x_2	s_1	s_2	s_3	b
s_1	1.6	0	1	0	-0.2	10
s_2	4	0	0	1	-1	20
x_2	0.4	1.0	0	0	0.2	8
z	-0.8	0	0	0	0.6	24

 $10/1.6 = 6.25, \ 20/4 = 5, \ 8/0.4 = 20$

$$\mathbf{r}_i = \mathbf{r}_2 = [4, 0, 0, 1, -1, 20]$$

$$\mathbf{r}_2 = [1, 0, 0, 0.25, -0.25, 5]$$

$$a_{k1} = 0, (k = 1, 3)$$

 $a_{4j}\mathbf{r}_i = a_{41}\mathbf{r}_4$

	x_1	x_2	s_1	s_2	s_3	b
s_1	0	0	1	-0.4	0.2	2
x_1	1	0	0	0.25	-0.25	5
x_2	0	1	0	-0, 1	0.3	6
z	0	0	0	0.2	0.4	28

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$$x_1 = b_2 = 5$$

$$y_2 = b_3 = 6$$

$$s_2 = s_3 = 0$$

$$C_{N+M}^M = C_5^3 = 10$$

$$x_1 = s_2 = 0$$

minimize
$$f(\mathbf{x}) = \frac{1}{2}[\mathbf{x} - \mathbf{m}]^T \mathbf{Q}[\mathbf{x} - \mathbf{m}] = \frac{1}{2}\mathbf{x}^T \mathbf{Q}\mathbf{x} + \mathbf{c}^T \mathbf{x} + c$$

subject to: $\mathbf{A}\mathbf{x} \leq \mathbf{b}$

$$\mathbf{Q} = \mathbf{Q}^T$$

$$\mathbf{c} = -\mathbf{Q}\mathbf{m}, \ c = \mathbf{m}^T \mathbf{Q} \mathbf{m} / 2$$

$$L(\mathbf{x}, \lambda) = f(\mathbf{x}) + \lambda^T (\mathbf{A}\mathbf{x} - \mathbf{b}) = \frac{1}{2}\mathbf{x}^T \mathbf{Q}\mathbf{x} + \mathbf{c}^T \mathbf{x} + \lambda^T (\mathbf{A}\mathbf{x} - \mathbf{b})$$

$$\nabla_{\mathbf{x}} L(\mathbf{x}, \lambda)$$

$$\nabla_{\mathbf{x}} f(\mathbf{x}) + \nabla_{\mathbf{x}} \lambda^T (\mathbf{A} \mathbf{x} - \mathbf{b}) = \mathbf{Q} \mathbf{x} + \mathbf{c} + \mathbf{A}^T \lambda = \mathbf{0}$$

$$\nabla_{\lambda}L(\mathbf{x},\lambda)$$

 $\begin{bmatrix} \mathbf{Q} & \mathbf{A}^T \end{bmatrix} \begin{bmatrix} \mathbf{x} \end{bmatrix}$

A

 $-\mathbf{c}$

$$f(x_1, x_2) = x_1^2 + x_2^2 = \frac{1}{2} [x_1, x_2] \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \frac{1}{2} \mathbf{x}^T \mathbf{Q} \mathbf{x}$$

$$\mathbf{A}\mathbf{x} = [1, \ 1] \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = x_1 + x_2 = \mathbf{b} = 1$$

$$\mathbf{Q} = \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}, \quad \mathbf{A} = [1, 1], \quad \mathbf{c} = \begin{bmatrix} 0 \\ 0 \end{bmatrix},$$

 $\mathbf{b} = 1;$

$$\begin{bmatrix} 2 & 0 & 1 \\ 0 & 2 & 1 \\ 1 & 1 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \lambda \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

$$x_1^* = x_2^* = 0.5$$

$$\lambda^* = -1$$

$$f(x_1^*, x_2^*) = 0.5$$

minimize
$$f(\mathbf{x}) = \frac{1}{2}\mathbf{x}^T\mathbf{Q}\mathbf{x} + \mathbf{c}^T\mathbf{x}$$
 subject to:
$$\mathbf{A}\mathbf{x} \leq \mathbf{b}$$

minimize
$$f(\mathbf{x})$$

subject to:
$$\begin{cases} \mathbf{h}(\mathbf{x}) = \mathbf{0} \\ \mathbf{g}(\mathbf{x}) \leq \mathbf{0} \text{ or } \mathbf{g}(\mathbf{x}) \geq \mathbf{0} \\ \mathbf{x} \geq \mathbf{0} \end{cases}$$

$$\mathbf{g}(\mathbf{x}) \leq \mathbf{0} \implies \mathbf{g}(\mathbf{x}) + \mathbf{s} = \mathbf{0}, \qquad \mathbf{g}(\mathbf{x}) \geq \mathbf{0} \implies \mathbf{g}(\mathbf{x}) - \mathbf{s} = \mathbf{0}$$

minimize
$$f(\mathbf{x})$$

subject to:
$$\begin{cases} \mathbf{h}(\mathbf{x}) = \mathbf{0} \\ \mathbf{x} \ge \mathbf{0} \end{cases}$$

$$L(\mathbf{x}, \lambda, \mu) = f(\mathbf{x}) + \lambda^T \mathbf{h}(\mathbf{x}) - \mu^T \mathbf{x}$$

$$\begin{cases} \nabla_{\mathbf{x}} L(\mathbf{x}, \lambda, \mu) = \mathbf{g}_{f}(\mathbf{x}) + \mathbf{J}_{\mathbf{h}}^{T}(\mathbf{x})\lambda - \mu = \mathbf{0} & \text{(stationarity)} \\ \mathbf{h}(\mathbf{x}) = \mathbf{0}, \quad \mathbf{x} \geq \mathbf{0} & \text{(dual feasibility)} \\ \mu \geq \mathbf{0} & \text{(dual feasibility)} \\ \mu_{j} x_{j} = 0, \quad (j = 1, \dots, N) & \text{or } \mathbf{XM1} = \mathbf{0} & \text{(complementarity)} \end{cases}$$

$$\mathbf{X} = diag(x_1, \cdots, x_N)$$

$$\mathbf{M} = diag(\mu_1, \cdots, \mu_N)$$

$$\begin{bmatrix} x_1 & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & x_n \end{bmatrix}_{N \times N}, \quad \mathbf{M} = \begin{bmatrix} \mu_1 & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & \mu_N \end{bmatrix}_{N \times N}, \quad \mathbf{J_h}(\mathbf{x}) = \begin{bmatrix} \frac{\partial h_1}{\partial x_1} & \cdots & \frac{\partial h_1}{\partial x_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial h_m}{\partial x_1} & \cdots & \frac{\partial h_m}{\partial x_n} \end{bmatrix}_{M \times N}$$

$$I(x) = \begin{cases} 0 & x \ge 0\\ \infty & x < 0 \end{cases}$$

minimize
$$f(\mathbf{x}) + \sum_{i=1}^{N} I(x_i)$$

subject to $\mathbf{h}(\mathbf{x}) = \mathbf{0}$

$$\ln(x) \stackrel{t \to \infty}{\Longrightarrow} I(x)$$

minimize
$$f(\mathbf{x}) - \frac{1}{t} \sum_{i=1}^{N} \ln x_i$$

subject to $\mathbf{h}(\mathbf{x}) = \mathbf{0}$

$$L(\mathbf{x}, \lambda, \mu) = f(\mathbf{x}) + \lambda^T \mathbf{h}(\mathbf{x}) - \frac{1}{t} \sum_{j=1}^{N} \ln x_j$$

$$\nabla_{\mathbf{x}} L(\mathbf{x}, \lambda, \mu) = \mathbf{g}_f(\mathbf{x}) + \mathbf{J}_{\mathbf{h}}^T(\mathbf{x})\lambda - \frac{1}{t}\mathbf{X}^{-1}\mathbf{1}$$

$$\mu = \frac{\mathbf{x}}{t} = \frac{1}{t} \mathbf{X}^{-1} \mathbf{1},$$

$$\mathbf{X}\mu = \mathbf{XM1} = \frac{\mathbf{1}}{t}$$

$$\begin{cases} \mathbf{g}_f(\mathbf{x}) + \mathbf{J}_{\mathbf{h}}^T(\mathbf{x})\lambda - \mu = \mathbf{0} & \text{(stationarity)} \\ \mathbf{h}(\mathbf{x}) = \mathbf{0} & \text{(primal feasibility)} \\ \mathbf{XM1} - \mathbf{1}/t = \mathbf{0} & \text{(complementarity)} \end{cases}$$

$$\mu \geq \mathbf{0}$$

$$\mathbf{F}(\mathbf{x}, \lambda, \mu) = \mathbf{0}$$

 $\mathbf{F}(\mathbf{x}, \lambda, \mu)$

$$\mathbf{F}(\mathbf{x}, \lambda, \mu) = \begin{bmatrix} \mathbf{g}_f(\mathbf{x}) + \mathbf{J}_{\mathbf{h}}^T(\mathbf{x})\lambda - \mu \\ \mathbf{h}(\mathbf{x}) \\ \mathbf{XM1} - \mathbf{1}/t \end{bmatrix}, \quad \mathbf{J}_{\mathbf{F}} = \begin{bmatrix} \mathbf{W}(\mathbf{x}) & \mathbf{J}_{\mathbf{h}}^T(\mathbf{x}) & -\mathbf{I} \\ \mathbf{J}_{\mathbf{h}}(\mathbf{x}) & \mathbf{0} & \mathbf{0} \\ \mathbf{M} & \mathbf{0} & \mathbf{X} \end{bmatrix}$$

$$\nabla_{\mathbf{x}} \left[\mathbf{g}_f(\mathbf{x}) + \mathbf{J}_{\mathbf{h}}^T(\mathbf{x}) \lambda - \mu \right]$$

$$\nabla_{\mathbf{x}} \mathbf{g}_f(\mathbf{x}) - \nabla_{\mathbf{x}} \left[\sum_{i=1}^M \lambda_i \frac{\partial h_i}{\partial x_1}, \cdots, \sum_{i=1}^M \lambda_i \frac{\partial h_i}{\partial x_N} \right]^T = \mathbf{H}_f(\mathbf{x}) - \sum_{i=1}^M \lambda_i \mathbf{H}_{h_i}(\mathbf{x})$$

$$\mathbf{H}_{h_i}, \ (i=1,\cdots,M)$$

$$\begin{bmatrix} \mathbf{x}_{n+1} \\ \lambda_{n+1} \\ \mu_{n+1} \end{bmatrix} = \begin{bmatrix} \mathbf{x}_n \\ \lambda_n \\ \mu_n \end{bmatrix} + \alpha \begin{bmatrix} \delta \mathbf{x}_n \\ \delta \lambda_n \\ \delta \mu_n \end{bmatrix}$$

$$[\delta \mathbf{x}_n, \delta \lambda_n, \delta \mu_n]^T$$

$$\mathbf{J}_{\mathbf{F}}(\mathbf{x}_n, \lambda_n, \mu_n) \begin{bmatrix} \delta \mathbf{x}_n \\ \delta \lambda_n \\ \delta \mu_n \end{bmatrix} = \begin{bmatrix} \mathbf{W}(\mathbf{x}_n) & \mathbf{J}_{\mathbf{h}}^T(\mathbf{x}_n) & -\mathbf{I} \\ \mathbf{J}_{\mathbf{h}}(\mathbf{x}_n) & \mathbf{0} & \mathbf{0} \\ \mathbf{M}_n & \mathbf{0} & \mathbf{X}_n \end{bmatrix} \begin{bmatrix} \delta \mathbf{x}_n \\ \delta \lambda_n \\ \delta \mu_n \end{bmatrix}$$

$$-\mathbf{F}(\mathbf{x}_n, \lambda_n, \mu_n) = -\begin{bmatrix} \mathbf{g}_f(\mathbf{x}_n) + \mathbf{J}_{\mathbf{h}}^T(\mathbf{x}_n)\lambda_n - \mu_n \\ \mathbf{h}(\mathbf{x}_n) \\ \mathbf{X_n}\mathbf{M_n}\mathbf{1} - \mathbf{1}/t \end{bmatrix}$$

$$\mu_0 = \mathbf{x}_0/t$$

$$\mathbf{M}\delta\mathbf{x} + \mathbf{X}\delta\mu = -\mathbf{X}\mathbf{M}\mathbf{1} + \mathbf{1}/t$$

$$\mathbf{X}^{-1}\mathbf{M}\delta\mathbf{x} + \delta\mu = -\mathbf{M}\mathbf{1} + \mathbf{X}^{-1}\mathbf{1}/t = -\mu + \mathbf{X}^{-1}\mathbf{1}/t$$

$$\mathbf{W}\delta\mathbf{x} + \mathbf{J}_{\mathbf{h}}^{T}(\mathbf{x})\delta\lambda - \delta\mu = -\mathbf{g}_{f}(\mathbf{x}) - \mathbf{J}_{\mathbf{h}}^{T}(\mathbf{x})\lambda + \mu$$

$$(\mathbf{W} + \mathbf{X}^{-1}\mathbf{M})\delta\mathbf{x} + \mathbf{J}_{\mathbf{h}}^{T}(\mathbf{x})\delta\lambda = -\mathbf{g}_{f}(\mathbf{x}) - \mathbf{J}_{\mathbf{h}}^{T}(\mathbf{x})\lambda + \mathbf{X}^{-1}\mathbf{1}/t = - \nabla_{\mathbf{x}} L(\mathbf{x}, \lambda, \mu)$$

$$\left[\begin{array}{cc} \mathbf{W} + \mathbf{X}^{-1}\mathbf{M} & \mathbf{J}_{\mathbf{h}}^{T}(\mathbf{x}) \\ \mathbf{J}_{\mathbf{h}}(\mathbf{x}) & \mathbf{0} \end{array}\right] \left[\begin{array}{c} \delta \mathbf{x} \\ \delta \lambda \end{array}\right] = - \left[\begin{array}{c} \nabla_{\mathbf{x}} L(\mathbf{x}, \lambda, \mu) \\ \mathbf{h}(\mathbf{x}) \end{array}\right]$$

$$\mathbf{M}\delta\mathbf{x} + \mathbf{X}\delta\mu = -\mathbf{X}\mathbf{M}\mathbf{1} - \mathbf{1}/t$$

$$\delta \mu = \mathbf{X}^{-1} \mathbf{1} / t - \mathbf{X}^{-1} \mathbf{M} \delta \mathbf{x} - \mu$$

minimize
$$f(\mathbf{x}) = \mathbf{c}^T \mathbf{x}$$

subject to: $\mathbf{h}(\mathbf{x}) = \mathbf{A}\mathbf{x} - \mathbf{b} = \mathbf{0}, \quad \mathbf{x} \ge \mathbf{0}$

$$\mathbf{g}_f(\mathbf{x}) = \nabla_{\mathbf{x}} f(\mathbf{x}) = \mathbf{c}$$

$$\mathbf{J_h}(\mathbf{x}) = \nabla_{\mathbf{x}}(\mathbf{A}\mathbf{x} - \mathbf{b}) = \mathbf{A}$$

$$\mathbf{W}(\mathbf{x}) = \nabla_{\mathbf{x}}^2 L(\mathbf{x}, \lambda, \mu) = \mathbf{0}$$

$$L(\mathbf{x}, \lambda, \mu) = \mathbf{c}^T \mathbf{x} + \lambda^T (\mathbf{A} \mathbf{x} - \mathbf{b}) - \mu^T \mathbf{x}$$

$$\begin{cases} \nabla_{\mathbf{x}} L(\mathbf{x}, \lambda, \mu) = \mathbf{g}_f(\mathbf{x}) + \mathbf{J}_{\mathbf{h}}^T(\mathbf{x})\lambda - \mu = \mathbf{c} + \mathbf{A}^T \lambda - \mu = \mathbf{0} \\ \mathbf{h}(\mathbf{x}) = \mathbf{A}\mathbf{x} - \mathbf{b} = \mathbf{0} \\ \mathbf{X}\mathbf{M}\mathbf{1} - \mathbf{1}/t = \mathbf{0} \end{cases}$$

$$\left[egin{array}{ccc} \mathbf{0} & \mathbf{A}^T & -\mathbf{I} \ \mathbf{A} & \mathbf{0} & \mathbf{0} \ \mathbf{M} & \mathbf{0} & \mathbf{X} \end{array}
ight] \left[egin{array}{c} \delta \mathbf{x} \ \delta \lambda \ \delta \mu \end{array}
ight] = - \left[egin{array}{c} \mathbf{c} + \mathbf{A}^T \lambda - \mu \ \mathbf{A} \mathbf{x} - \mathbf{b} \ \mathbf{X} \mathbf{M} \mathbf{1} - \mathbf{1}/t \end{array}
ight]$$

$$\mathbf{c} + \mathbf{A}^T \lambda - \mu = \mathbf{0}$$

maximize
$$f(x_1, x_2) = 2x_1 + 3x_2$$
 minimize $f(x_1, x_2, x_3, x_4, x_5) = -2x_1 - 3x_2$
subject to:
$$\begin{cases} 2x_1 + x_2 \le 18 \\ 6x_1 + 5x_2 \le 60 \\ 2x_1 + 5x_2 \le 40 \\ x_1 \ge 0, \quad x_2 \ge 0 \end{cases} \implies \text{subject to:} \begin{cases} 2x_1 + x_2 + x_3 = 18 \\ 6x_1 + 5x_2 + x_4 = 60 \\ 2x_1 + 5x_2 + x_5 = 40 \\ x_i \ge 0, \quad (i = 1, \dots, 5) \end{cases}$$

minimize
$$f(\mathbf{x}) = \mathbf{c}^T \mathbf{x}$$

subject to $\mathbf{h}(\mathbf{x}) = \mathbf{A}\mathbf{x} - \mathbf{b} = \mathbf{0}, \ \mathbf{x} \ge \mathbf{0}$

$$\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{bmatrix}, \quad \mathbf{c} = \begin{bmatrix} -2 \\ -3 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \quad \mathbf{A} = \begin{bmatrix} 2 & 1 & 1 & 0 & 0 \\ 6 & 5 & 0 & 1 & 0 \\ 2 & 5 & 0 & 0 & 1 \end{bmatrix}, \quad \mathbf{b} = \begin{bmatrix} 18 \\ 60 \\ 40 \end{bmatrix}$$

 x_3, x_4, x_5

$$\mathbf{x}_0 = [1, 2, 1, 1, 1]^T$$

$$\mathbf{x}^* = [5, 6]^T$$

$$f(\mathbf{x}^*) = 2x_1 + 3x_2 = 28$$

	(x_1)	$x_2)$	$f(\mathbf{x})$	error
1	(1.000000e + 00)	2.000000e + 00	-8.000000	52.413951
2	(4.654514e + 00)	6.346354e + 00	-28.348090	3.080204
3	(5.040828e + 00)	5.946973e + 00	-27.922575	0.213509
4	(4.997282e + 00)	6.001764e + 00	-27.999856	0.004262
5	(4.999906e + 00)	5.999962e + 00	-27.999697	0.000303
6	(4.999989e + 00)	5.999996e + 00	-27.999966	0.000034
7	(4.9999999e + 00)	6.000000e + 00	-27.999996	0.000004
8	(5.000000e + 00)	6.000000e + 00)	-28.000000	0.000000

$$x_3 = 2, \ x_4 = x_5 = 0$$

minimize
$$f(\mathbf{x}) = \frac{1}{2}\mathbf{x}^T\mathbf{Q}\mathbf{x} + \mathbf{c}^T\mathbf{x}$$

subject to: $\mathbf{h}(\mathbf{x}) = \mathbf{A}\mathbf{x} - \mathbf{b} = \mathbf{0}, \quad \mathbf{x} \ge \mathbf{0}$

$$\mathbf{g}_f(\mathbf{x}) = \nabla_{\mathbf{x}} f(\mathbf{x}) = \mathbf{Q} \mathbf{x} + \mathbf{c}$$

$$\mathbf{W}(\mathbf{x}) = \nabla_{\mathbf{x}}^2 L(\mathbf{x}, \lambda, \mu) = \mathbf{Q}$$

$$L(\mathbf{x}, \lambda, \mu) = \frac{1}{2} \mathbf{x}^T \mathbf{Q} \mathbf{x} + \mathbf{c}^T \mathbf{x} + \lambda^T (\mathbf{A} \mathbf{x} - \mathbf{b}) - \mu^T \mathbf{x}$$

$$\begin{bmatrix} \mathbf{Q} & \mathbf{A}^T & -\mathbf{I} \\ \mathbf{A} & \mathbf{0} & \mathbf{0} \\ \mathbf{M} & \mathbf{0} & \mathbf{X} \end{bmatrix} \begin{bmatrix} \delta \mathbf{x} \\ \delta \lambda \\ \delta \mu \end{bmatrix} = - \begin{bmatrix} \mathbf{Q} \mathbf{x} + \mathbf{c} + \mathbf{A}^T \lambda - \mu \\ \mathbf{A} \mathbf{x} - \mathbf{b} \\ \mathbf{X} \mathbf{M} \mathbf{1} - \mathbf{1}/t \end{bmatrix}$$

minimize
$$f(\mathbf{x}) = [\mathbf{x} - \mathbf{m}]^T \mathbf{Q} [\mathbf{x} - \mathbf{m}] = \mathbf{x}^T \mathbf{Q} \mathbf{x} + \mathbf{c}^T \mathbf{x} + d$$

subject to $\mathbf{h}(\mathbf{x}) = \mathbf{A} \mathbf{x} - \mathbf{b} = \mathbf{0}, \ \mathbf{x} \ge \mathbf{0}$

$$\begin{bmatrix} 4 & -1 \\ -1 & 4 \end{bmatrix}, \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}, \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}, \begin{bmatrix} 3 & -1 \\ -1 & 3 \end{bmatrix}$$

 $\mathbf{Q} = |$

$$\left[\begin{array}{c} 9 \\ 5 \end{array}\right], \quad \left[\begin{array}{c} 7 \\ 2 \end{array}\right]$$

 \mathbf{m}

$$\mathbf{x}_0 = [2, \ 1]^T$$