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1. (a) 设 $f(x) = \frac{1}{2}x^T Gx + b^T x + c$
 $g(x) = Gx + b$
 对于最优点 x^* 有 $Gx^* + b = 0$
 $g(x^{(0)}) = G(x^* + \mu s) + b$
 $= Gx^* + b + G\mu s$
 $= G\mu s$
 由特征向量的性质得 $Gs = \lambda s$
 $\therefore g^{(0)} = \mu \lambda s$
 $x^{(1)} = x^{(0)} - \frac{1}{\lambda} g^{(0)} = x^{(0)} - \mu s = x^*$
 \therefore 一次迭代后终止

(b) 设 $G = aI$, $x^{(0)} = x^* + p$
 $g^{(0)} = G(x^* + p) + b$
 $= Gp$
 $Gp = ap$
 $x^{(1)} = x^{(0)} - \frac{1}{a} g^{(0)} = x^* - p = x^*$
 \therefore 一次迭代后终止

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2. (a)
$$\begin{cases} \frac{\partial q}{\partial x_1} = 10x_1 - x_2 - 11 = 0 \\ \frac{\partial q}{\partial x_2} = 10x_2 - x_1 + 11 = 0 \end{cases}$$

$\therefore x^* = (1, -1)^T$

(b) $G = \begin{bmatrix} 10 & -1 \\ -1 & 10 \end{bmatrix}$
 $\therefore G$ 正定
 $\therefore q(x)$ 是凸函数

(c) G 的特征值 λ 满足 $(\lambda - 10)^2 + 1 = 0$
 $\therefore \lambda = 9, 11$

$$\frac{f(x_{k+1}) - f(x^*)}{f(x_k) - f(x^*)} \leq \left(\frac{11-9}{11+9} \right)^2 = 0.01$$

 \therefore 收敛因子最大不超过 0.01

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3. 经共轭梯度法求得 $x^* = (-0.4472, -1.8944, -3.3416, -2.7889)^T$
 $f(x^*) = -6.2361$

$$g^{(0)} = Gx^{(0)} + b = (-1, 0, 2, \sqrt{5})^T$$

$$Gg^{(0)} = (-2, -1, 4 - \sqrt{5}, -2 + 2\sqrt{5})^T$$

$$G^2g^{(0)} = \begin{pmatrix} 5 & -4 & 1 & 0 \\ -4 & 6 & -4 & 1 \\ 1 & -4 & 6 & -4 \\ 0 & 1 & -4 & 5 \end{pmatrix} (-1, 0, 2, \sqrt{5})^T = (-3, -4 + \sqrt{5}, 11 - 4\sqrt{5}, -8 + 5\sqrt{5})^T$$

$g^{(0)}$ 和 $Gg^{(0)}$ 显然线性无关

$$G^2g^{(0)} = \lambda_0 g^{(0)} + \lambda_1 Gg^{(0)} \text{ 求得 } \lambda_0 = -5 + 2\sqrt{5}, \lambda_1 = 4 - \sqrt{5}$$

$$G^3g^{(0)} = G \cdot G^2g^{(0)} = \lambda_0 Gg^{(0)} + \lambda_1 G^2g^{(0)} = \lambda_0 \lambda_1 g^{(0)} + (\lambda_0 + \lambda_1^2) Gg^{(0)}$$

\therefore 向量组 $g^{(0)}, Gg^{(0)}, G^2g^{(0)}, G^3g^{(0)}$ 中只有两个独立向量

(b) 迭代次数与 $g^{(0)}, Gg^{(0)}, G^2g^{(0)}, G^3g^{(0)}$ 中独立向量的个数相等

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$$4. (a) H^{(0)} = I, g = \begin{bmatrix} 20x_1 \\ 2x_2 \end{bmatrix}, G = \begin{bmatrix} 20 & 0 \\ 0 & 2 \end{bmatrix}, x^{(0)} = (0.1, 1)^T$$

$$g^{(0)} = \begin{bmatrix} 2 \\ 2 \end{bmatrix}$$

$$p^{(0)} = -H_0 g_0 = \begin{bmatrix} -2 \\ -2 \end{bmatrix}$$

$$\alpha^{(0)} = \frac{g^{(0)T} p^{(0)}}{p^{(0)T} G p^{(0)}} = \frac{8}{88} = \frac{1}{11}$$

$$x^{(1)} = x^{(0)} + \alpha^{(0)} p^{(0)} = (-\frac{9}{110}, \frac{9}{11})^T$$

$$g^{(1)} = (-\frac{18}{11}, \frac{18}{11})^T$$

$$s^{(0)} = x^{(1)} - x^{(0)} = (-\frac{2}{11}, -\frac{2}{11})^T$$

$$y^{(0)} = g^{(1)} - g^{(0)} = (-\frac{40}{11}, -\frac{4}{11})^T$$

$$H^{(1)} = H^{(0)} + \frac{s^{(0)} s^{(0)T}}{s^{(0)T} y^{(0)}} - \frac{H^{(0)} y^{(0)} y^{(0)T} H^{(0)}}{y^{(0)T} H^{(0)} y^{(0)}} = I + \begin{pmatrix} \frac{1}{22}, \frac{1}{22} \\ \frac{1}{22}, \frac{1}{22} \end{pmatrix} + \begin{pmatrix} \frac{100}{101}, \frac{10}{101} \\ \frac{10}{101}, \frac{1}{101} \end{pmatrix}$$

$$= \frac{1}{2222} \begin{bmatrix} 123 & -119 \\ -119 & 2301 \end{bmatrix}$$

$$p^{(1)} = -H^{(1)} g^{(1)} = \frac{1}{101} (18, -180)^T$$

$$\alpha^{(1)} = \frac{g^{(1)T} p^{(1)}}{p^{(1)T} G p^{(1)}} = \frac{101}{220}$$

$$x^{(2)} = x^{(1)} + \alpha^{(1)} p^{(1)} = (0, 0)^T$$

$$g^{(2)} = (0, 0)^T$$

$$s^{(1)} = x^{(2)} - x^{(1)} = (\frac{9}{110}, -\frac{9}{11})^T$$

$$y^{(1)} = g^{(2)} - g^{(1)} = (\frac{18}{11}, -\frac{18}{11})^T$$

$$H^{(2)} = \begin{bmatrix} \frac{1}{20} & 0 \\ 0 & \frac{1}{2} \end{bmatrix}$$

\therefore 2次一维搜索后达到最优解, 且 $H^{(2)} = G^{-1}$

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$$b) p^{(0)} = -g^{(0)} = (-2, -2)^T$$

$$t = - \frac{p^{(0)T} g^{(0)}}{p^{(0)T} p^{(0)}} = \frac{1}{11}$$

$$x^{(1)} = x^{(0)} + t p^{(0)} = (-\frac{9}{11}, \frac{9}{11})^T$$

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$$g^{(1)} = (-\frac{18}{11}, \frac{18}{11})^T$$

$$\alpha = \frac{g^{(1)T} p^{(0)}}{p^{(0)T} p^{(0)}} = \frac{81}{121}$$

$$p^{(1)} = -g^{(1)} + \alpha p^{(0)} = (\frac{36}{121}, -\frac{360}{121})^T$$

$$t = - \frac{p^{(1)T} g^{(1)}}{p^{(1)T} p^{(1)}} = \frac{11}{40}$$

$$x^{(2)} = x^{(1)} + t p^{(1)} = (0, 0)^T$$

$$g^{(2)} = (0, 0)^T$$

∴ 该方法与共轭梯度法等价