Continuous	Optimization:	Assignment	2
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Exercise 1

(i)

The gradient of $\varphi(x)$ is given by

$$\nabla \varphi(x) = \frac{1}{2} (A + A^{\top}) x - b$$

$$= \frac{1}{2} (2A) x - b$$

$$= Ax - b$$
A is symetric

Similarly, the Hessian of $\varphi(x)$ is given by

$$\nabla^2 \varphi(x) = (D(\nabla \varphi(x)))^{\top}$$
$$= (A^{\top})^{\top} = A$$

Suppose $\nabla \varphi(x^*) = 0$ and $\nabla^2 \varphi(x^*)$ is positive definite, then Theorem 6.9 shows that x^* is a local minimum of $\varphi(x)$. To meet the condition, we only need to show that $\nabla \varphi(x^*) = 0$ because A is positive definite which meets the second part of the condition. Thus we can find the minimizer by solving Ax = b.

(ii)

The steepest descent direction is given by

$$d^{(k)} = -\nabla \varphi(x^{(k)})$$
$$= -Ax^{(k)} + b = r^{(k)}$$

(iii)

Let $g_k(\tau) = x^{(k)} + \tau r^{(k)}$ be the function that gives $x^{(k+1)}$ given τ at time step k. We can rewrite the objective function as $\min_{\tau > 0} \varphi(g_k(\tau))$ and to find the minimizer, we can solve $\frac{\partial \varphi}{\partial \tau} = 0$.

$$\begin{split} \frac{\partial \varphi}{\partial \tau} &= \frac{\partial g_k}{\partial \tau} \cdot \frac{\partial \varphi}{\partial g_k} \\ &= (b - Ax^{(k)})^\top (Ag_k - b) \\ &= -b^\top b + (Ax^{(k)})^\top b + b^\top Ag_k - (Ax^{(k)})^\top Ag_k \\ &= -b^\top b + (Ax^{(k)})^\top b + b^\top A(x^{(k)} + \tau(b - Ax^{(k)})) - (Ax^{(k)})^\top A(x^{(k)} + \tau(b - Ax^{(k)})) \\ &= -b^\top b + 2(Ax^{(k)})^\top b - (Ax^{(k)})^\top Ax^{(k)} + \tau(b - Ax^{(k)})^\top A(b - Ax^{(k)}) \\ &= -(b - Ax^{(k)})^\top (b - Ax^{(k)}) + \tau(b - Ax^{(k)})^\top A(b - Ax^{(k)}) = 0 \\ &\Rightarrow \tau = \frac{(b - Ax^{(k)})^\top (b - Ax^{(k)})}{(b - Ax^{(k)})^\top A(b - Ax^{(k)})} \end{split}$$

Exercise 3

(a)

To show that f has a global minimizer at (a, a^2) we need to show that $f(a, a^2) \leq f(x_1, x_2)$ for all (x_1, x_2) .

$$f(a, a^2) - f(x_1, x_2) = 0 - (a - x_1)^2 - b(x_2 - x_1^2)^2 \le 0$$

(b)