Suggested Solutions to Assignment 3

- 1. (a) f^{**} is $\frac{1}{\sigma}$ -smooth, and $f^{**} = f$ since f is proper closed convex.
 - (b) Notice that $f^* = (f^* \frac{\sigma}{2} \|\cdot\|^2) + \frac{\sigma}{2} \|\cdot\|^2$ and $f^* \frac{\sigma}{2} \|\cdot\|^2$ is proper convex, then by Theorem 4.17,

$$f = f^{**} = \left(\left(f^* - \frac{\sigma}{2} \left\| \cdot \right\|^2 \right) + \frac{\sigma}{2} \left\| \cdot \right\|^2 \right)^*$$
$$= \left(f^* - \frac{\sigma}{2} \left\| \cdot \right\|^2 \right)^* \square \left(\frac{\sigma}{2} \left\| \cdot \right\|^2 \right)^*$$
$$= \left(f^* - \frac{\sigma}{2} \left\| \cdot \right\|^2 \right)^* \square \left(\frac{1}{2\sigma} \left\| \cdot \right\|^2 \right)$$

Finally, $\left(f^* - \frac{\sigma}{2} \|\cdot\|^2\right)^*$ is convex since $f^* - \frac{\sigma}{2} \|\cdot\|^2$ is proper convex.

- (c) Let $h_1(x) = x^3 x^2$ if $x \ge 0$, while $h(x) = \infty$ otherwise. Let $h_2(x) = x^2$. Define $f = h_1 \square h_2$.
- 2. (a) By the second prox theorem, for any x,

$$\begin{split} x - \mathrm{prox}_f(x) &\in \partial f \left(\mathrm{prox}_f(x) \right) \\ \Leftrightarrow & x \in \mathrm{prox}_f(x) + \partial f \left(\mathrm{prox}_f(x) \right) = \left(I + \partial f \right) \left(\mathrm{prox}_f(x) \right) \\ \Leftrightarrow & \mathrm{prox}_f(x) = \left(I + \partial f \right)^{-1}(x). \end{split}$$

(b)

$$\nabla M_f^{\mu}(x) = \frac{1}{\mu} \left(x - \operatorname{prox}_{\mu f}(x) \right) = \frac{1}{\mu} \left(I - \operatorname{prox}_{\mu f} \right) (x)$$
$$= \frac{1}{\mu} \left(I - \left(I + \partial (\mu f)^{-1} \right) \right) (x) = \frac{1}{\mu} \left(I - \left(I + \mu \partial f \right)^{-1} \right) (x).$$

3.

- 4. Hint:
 - (b) Characterize the positive definiteness of $\lambda A + I$. Hence we need $\lambda \geq 0$ if A is positive semidefinite and $0 \leq \lambda < \frac{-1}{\lambda_{\min}(A)}$ otherwise. In this case, $\operatorname{prox}_{\lambda f} = (\lambda A + I)^{-1}(x + \lambda b)$.
- 5. Recall Theorem 6.13 for

$$f(x) = \left(g - \frac{\sigma}{2} \|\cdot\|^2\right)(x) + \frac{\sigma - c}{2} \|x\|^2 + \langle e, x \rangle,$$

Then

$$\operatorname{prox}_f(x) = \operatorname{prox}_{\frac{1}{\sigma - c + 1} \left(g - \frac{\sigma}{2} \| \cdot \|^2\right)} \left(\frac{x - e}{\sigma - c + 1}\right).$$

- 6. (a) For any x, $M_f \mu(x) \ge \inf_{u \in \mathbb{R}^n} f(u) > -\infty$ and choose $u_x \in \text{dom}(f)$, then $M_f^{\mu} \le f(u_x) + \frac{1}{2\mu} \|x u_x\|^2 < \infty$.
 - (b) $M_f^{\mu}(x) \le f(x) + \frac{1}{2\mu} \|x x\|^2 = f(x)$.

- 7. (i). $(a) \Rightarrow (b)$.
 - (ii). (b) \Rightarrow (c). We know $\exists \alpha, \beta, \gamma \in \mathbb{R}$, $a \in \mathbb{R}^n$ such that $q(u) = \alpha \|u\|^2 + \beta a^T u + \gamma$. Then choose $\mu = \frac{1}{4|\alpha|}$.

$$f(u) + \frac{1}{2\mu} \|u\|^2 \ge \alpha \|u\|^2 + \beta a^T u + \gamma + \frac{1}{2\mu} \|u\|^2$$

$$\ge |\alpha| \|u\|^2 + \beta a^T u + \gamma.$$

(iii). $(c) \Rightarrow (d)$. $\exists c$ constant such that $f(u) + \frac{1}{2\mu} \|u\|^2 > c$. Then

$$\liminf_{\|x\| \to \infty} \frac{f(x)}{\|x\|^2} + \frac{1}{2\mu} \ge \liminf_{\|x\| \to \infty} \frac{c}{\|x\|^2} = 0.$$

(iv). $(d) \Rightarrow (a)$. Define

$$\liminf_{\|x\| \to \infty} \frac{f(x)}{\|x\|^2} = \alpha$$

and $\mu = 1/6|\alpha|$. Notice that there exists M > 0 such that $f(u)/\|u\|^2 \ge -2|\alpha|$ for which $u \notin \mathbb{B}_M(0)$. Then

$$\begin{split} M_f^{\mu}(x) &= \inf_{u \in \mathbb{R}^n} \left\{ f(u) + \frac{1}{2\mu} \left\| u - x \right\|^2 \right\} \\ &= \min \left\{ \inf_{u \in B_M(0)} \left\{ f(u) + \frac{1}{2\mu} \left\| u - x \right\|^2 \right\}, \inf_{u \notin B_M(0)} \left\{ f(u) + \frac{1}{2\mu} \left\| u - x \right\|^2 \right\} \right\} \\ &> - \infty, \end{split}$$

since

$$\inf_{u \in B_{M}(0)} \left\{ f(u) + \frac{1}{2\mu} \left\| u - x \right\|^{2} \right\} > -\infty \text{ by the Weierstrass theorem,}$$

and

$$\inf_{u \notin B_{M}(0)} \left\{ f(u) + \frac{1}{2\mu} \|u - x\|^{2} \right\}$$

$$\geq \inf_{u \notin B_{M}(0)} \left\{ -2|\alpha| \|u\|^{2} + 3|\alpha| \|u - x\|^{2} \right\} > -\infty.$$

8. Fix $\bar{\mu} > 0, \bar{x} \in \mathbb{R}^n$. Let $\bar{y} = \varphi(\bar{\mu}, \bar{x})$.

For any $\{x^k\} \to \bar{x}$, $\{\mu_k\} \to \bar{\mu}$, define $y^k = \varphi(\mu_k, x^k)$. Just show $y^k \to \bar{y}$. Notice that

$$f(y^k) + \frac{1}{2\mu_k} \|y^k - x^k\|^2 \le f(\bar{y}) + \frac{1}{2\mu_k} \|\bar{y} - x^k\|^2$$
.

Take the upper limit,

$$\limsup_{k\to\infty}f\left(y^{k}\right)+\frac{1}{2\bar{\mu}}\left\|y^{k}-\bar{x}\right\|^{2}\leq f\left(\bar{y}\right)+\frac{1}{2\bar{\mu}}\left\|\bar{y}-\bar{\mu}\right\|^{2}=M_{f}^{\bar{\mu}}\left(\bar{x}\right)<\infty.\tag{\star}$$

If $||y^k|| \to \infty$, then

$$\limsup_{k\to\infty}f\left(y^{k}\right)+\frac{1}{4\bar{\mu}}\left\|y^{k}-\bar{x}\right\|^{2}\leq\limsup_{k\to\infty}f\left(y^{k}\right)+\frac{1}{2\bar{\mu}}\left\|y^{k}-\bar{x}\right\|^{2}-\lim_{k\to\infty}\frac{1}{4\bar{\mu}}\left\|y^{k}-\bar{x}\right\|^{2}=-\infty.$$

While

$$f(y^k) + \frac{1}{4\bar{\mu}} \|y^k - \bar{x}\|^2 > M_f^{2\bar{\mu}}(\bar{x}),$$

i.e.,

$$\left\{ f\left(y^{k}\right) + \frac{1}{4\bar{\mu}} \left\| y^{k} - \bar{x} \right\|^{2} \right\}$$

has a finite lower bound. It's a contradiction with (\star) . Hence $\{y^k\}$ is bounded. Let y^* be a limit point of $\{y^k\}$, then by (\star) and the lower-semicontinuity of f, we have

$$M_{f}^{\bar{\mu}}(\bar{x}) \geq \limsup_{k \to \infty} f\left(y^{k}\right) + \frac{1}{2\bar{\mu}}\left\|y^{k} - \bar{x}\right\|^{2} \geq \liminf_{k \to \infty} f\left(y^{k}\right) + \frac{1}{2\bar{\mu}}\left\|y^{*} - \bar{x}\right\|^{2} \geq f\left(y^{*}\right) + \frac{1}{2\bar{\mu}}\left\|y^{*} - \bar{x}\right\|^{2} \geq M_{f}^{\bar{\mu}}(\bar{x}),$$

Then $y^* = \bar{y}$.

9.

- 10. (a) Apply the second projection theorem and the definition of normal cone.
 - (b) Notice that

$$\operatorname{dist}(\bar{x} + \alpha v, C) = \alpha \|v\| = \|(\bar{x} + \alpha v) - \bar{x}\|$$

implies that $P_C(\bar{x} + \alpha v) = \bar{x}$. Hence

$$N_C(\bar{x}) = N_C^{\text{prox}}(\bar{x}).$$

- 11. (a) fundamental prox-grad inequality.
 - (b) Plugging $y = \text{proj}(x; X^*)$ into the fundamental prox-grad inequality, we have

$$\langle G_L(x), x - \operatorname{proj}(x; X^*) \rangle \ge F(T_L(x)) - F(\operatorname{proj}(x; X^*)) + \frac{1}{2L} \|G_L(x)\|^2$$

$$\ge F(T_L(x)) - F_{\operatorname{opt}} + \frac{\alpha}{2L} \operatorname{dist}^2(x, X^*)$$

$$\ge \frac{\alpha}{2L} \operatorname{dist}^2(x, X^*).$$

12. Define $F_{\mu} = \mu h + (1 - \mu)f$, it's easy to see that F_{μ} is convex and L-smooth. The fundamental prox-grad inequality implies that

$$F_{\mu}(x) - F_{\mu}\left(y - \frac{1}{L}\nabla F_{\mu}(y)\right) \ge \frac{L}{2} \left\| x - \left(y - \frac{1}{L}\nabla F_{\mu}(y)\right) \right\|^{2} - \frac{L}{2} \left\| x - y \right\|^{2}.$$

For any $n \ge 0$, substituting $\mu = \mu_k$, $x = x^*$ and $y = x^n$ in the above inequality, we obtain

$$F_{\mu_n}(x^*) - F_{\mu_n}(x^{n+1}) \ge \frac{L}{2} \|x^* - x^{n+1}\|^2 - \frac{L}{2} \|x^* - x^n\|^2.$$

Summing the above inequality over $n = 0, 1, \dots, k-1$ we obtain

$$\begin{split} &\frac{L}{2} \left\| x^* - x^k \right\|^2 - \frac{L}{2} \left\| x^* - x^0 \right\|^2 \\ &\leq \sum_{n=0}^{k-1} \left(F_{\mu_n}(x^*) - F_{\mu_n}(x^{n+1}) \right) \\ &= \sum_{n=0}^{k-1} \left(\mu_n \left(h(x^*) - h(x^{n+1}) \right) + (1 - \mu_n) \left(f(x^*) - f(x^{n+1}) \right) \right) \\ &\leq \left[h(x^*) - \min_{n=0,1,\cdots,k-1} h(x^n) \right] \left(\sum_{n=0}^{k-1} \mu_n \right) + \left[f(x^*) - \min_{n=0,1,\cdots,k-1} f(x^n) \right] \left(k - \sum_{n=0}^{k-1} \mu_n \right). \end{split}$$

Thus

$$\left[\min_{n=0,1,\cdots,k-1} h(x^n) - h(x^*)\right] \left(\sum_{n=0}^{k-1} \mu_n\right) + \left[\min_{n=0,1,\cdots,k-1} f(x^n) - f(x^*)\right] \left(k - \sum_{n=0}^{k-1} \mu_n\right) \le \frac{L}{2} \|x^* - x^0\|^2.$$

Consequently,

$$\min_{n=0,1,\cdots,k-1} f(x^n) - f(x^*) \le \frac{L \|x^* - x^0\|^2}{2\left(k - \sum_{n=0}^{k-1} \mu_n\right)} \le \frac{L \|x^* - x^0\|^2}{2\left(k - \frac{(k-1)^{\alpha}}{\alpha} - \frac{1}{\alpha}\right)} = O\left(\frac{1}{k}\right),$$

and

$$\min_{n=0,1,\dots,k-1} h(x^n) - h(x^*) \le \frac{L \|x^* - x^0\|^2}{2 \sum_{n=0}^{k-1} \mu_n} \le \frac{\alpha L \|x^* - x^0\|^2}{2(k-1)^{\alpha}} = O\left(\frac{1}{k^{\alpha}}\right),$$