Appendix

Lemma 1 *(a) If random vector $\mathbf{X}$ is symmetric about 0, then $A\mathbf{X} + \mu$ is symmetric about μ.*
(b) If $\mathbf{X}$, $\mathbf{Y}$ are independent and both symmetric about 0, $\mathbf{Z} = \mathbf{X} + \mathbf{Y}$ is also symmetric about 0.

Proof: (a). Since $\mathbf{X}$ is symmetric about 0, for all y, $\Pr(A\mathbf{X} + \mu \geq \mu + y) = \Pr(A\mathbf{X} \geq y) = \int \mathbf{1}_{Ax \geq y} \mathrm{d}u_x = \int \mathbf{1}_{-Ax \geq y} \mathrm{d}u_x = \Pr(A\mathbf{X} \leq -y) = \Pr(A\mathbf{X} + \mu \leq \mu - y)$. From the definition of symmetry, we know $A\mathbf{X} + \mu$ is symmetric about μ.

(b). Since $\mathbf{X}$ and $\mathbf{Y}$ are independent and both symmetric about 0, for all z, $\Pr(\mathbf{Z} \geq z) = \iint \mathbf{1}_{x+y \geq z} \mathrm{d}u_x \mathrm{d}u_y = \int \mathrm{d}u_x \int \mathbf{1}_{y \geq z-x} \mathrm{d}u_y = \int \mathrm{d}u_x \int \mathbf{1}_{y \leq x-z} \mathrm{d}u_y = \int \mathrm{d}u_y \int \mathbf{1}_{x \geq y+z} \mathrm{d}u_x = \int \mathrm{d}u_y \int \mathbf{1}_{x \leq -y-z} \mathrm{d}u_x = \iint \mathbf{1}_{x+y \leq -z} \mathrm{d}u_x \mathrm{d}u_y = \Pr(\mathbf{Z} \leq -z)$. From the definition of symmetry, we know $\mathbf{Z}$ is symmetric about 0. ∎

Equation (26) in Proposition 3

$$\max_{\mathbf{x} \sim (\mu, \sigma^2)_{SU}} \mathbb{E}\big[(\mathbf{x} - t)_+\big] = \begin{cases} \frac{(\sqrt{3}\sigma - t + \mu)^2}{4\sqrt{3}\sigma}, & \text{if } \mu - \frac{\sigma}{\sqrt{3}} \leq t \leq \mu + \frac{\sigma}{\sqrt{3}} \\ \frac{\sigma^2}{9(t-\mu)}, & \text{if } t > \mu + \frac{\sigma}{\sqrt{3}} \\ -\frac{\sigma^2 + 9(t-\mu)^2}{9(t-\mu)}, & \text{if } t < \mu - \frac{\sigma}{\sqrt{3}} \end{cases}$$

Proof: Without loss of generality, we assume $\mu = 0$, otherwise we can explicitly achieve this by translation $x' = x - \mu, t' = t - \mu$. We use the technique established in [13] to prove (26).

Using the strong duality property between moment problems and linear programming [13], $\max_{\mathbf{x} \sim (0,\sigma^2)_{SU}} \mathbb{E}\big[(x - t)_+\big]$ is equivalent to:

$$\min_{y_0, y_1} \quad y_0 + y_1 \sigma^2 \tag{27}$$

$$\text{s.t.} \quad y_0 p + \frac{y_1 p^3}{3} \geq \frac{1}{2} \int_{-p}^{p} [x - t]_+ \mathrm{d}x, \quad \forall \, p \geq 0 \tag{28}$$

1). Consider $t \geq 0$.
Constraint (28) now can be simplified as:

$$\begin{cases} y_0 p + \frac{y_1 p^3}{3} \geq 0, & 0 \leq p \leq t \\ y_0 p + \frac{y_1 p^3}{3} \geq \frac{(p-t)^2}{4}, & p \geq t \end{cases}$$

The first condition implies $y_0 \geq 0$ and the second implies $y_1 \geq 0$. Therefore problem (27)-(28) can be rewritten as:

$$\min_{y_0 \geq 0, y_1 \geq 0} \quad y_0 + y_1 \sigma^2 \tag{29}$$

$$\text{s.t.} \quad y_0 p + \frac{y_1 p^3}{3} - \frac{(p-t)^2}{4} \geq 0, \quad \forall \, p \geq t \tag{30}$$

Denote

$$h(p) = \frac{y_1 p^3}{3} - \frac{p^2}{4} + (y_0 + \frac{t}{2})p - \frac{t^2}{4}$$

$$h'(p) = y_1 p^2 - \frac{p}{2} + (y_0 + \frac{t}{2})$$

Obviously, $h(t) \geq 0, h'(t) \geq 0$. The constraint (30) can be satisfied either

- $y_0 + t/2 - 1/(16y_1) \geq 0$, which means $\forall \, p$, $h'(p) \geq 0$, or
- $y_0 + t/2 - 1/(16y_1) \leq 0$ but the minimum of $h(p)$ is no less than 0.

The first case amounts to:

$$\min_{y_0 \geq 0, y_1 \geq 0} \quad y_0 + y_1 \sigma^2$$

$$\text{s.t.} \quad y_0 + t/2 - 1/(16y_1) \geq 0$$

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and the minimum is (exactly the same as that of symmetric distributions)

$$
\begin{cases}
(\sigma - t)/2, & 0 \le t \le \sigma/2 \\
\frac{\sigma^2}{8t}, & t \ge \sigma/2
\end{cases}
\tag{31}
$$

For the second case, let $p_+ = \frac{1+\sqrt{1-8ty_1-16y_0y_1}}{4y_1}$ be the larger minimizer of $h(p)$, problem (29)-(20) now becomes:

$$
\min_{y_1 \ge 0} \quad y_0 + y_1\sigma^2
$$

$$
\text{s.t.} \quad 0 \le y_0 \le \frac{1}{16y_1} - \frac{t}{2}
$$

$$
h(p_+) \ge 0 \Rightarrow \frac{t^3 - 30t^2 y_0 - 96ty_0^2 - 64y_0^3 + (t^2 + 16ty_0 + 16y_0^2)^{3/2}}{18t^4} \le y_1 \le \frac{1}{8t}
$$

The minimum is (thanks to Mathematica!)

$$
\begin{cases}
-t/2 + \frac{3\sigma^2 + t^2}{4\sqrt{3}\sigma}, & 0 \le t \le \sigma/\sqrt{3} \\
\frac{\sigma^2}{9t}, & t \ge \sigma/\sqrt{3}
\end{cases}
\tag{32}
$$

Comparing (31) and (32), we know the latter gives the minimum of problem (29)-(30).

2). Consider $t \le 0$.
Similarly, problem (27)-(28) can be rewritten as:

$$
\min_{y_0 \ge -t, y_1 \ge 0} \quad y_0 + y_1\sigma^2
$$

$$
\text{s.t.} \quad y_0 p + \frac{y_1 p^3}{3} - \frac{(p-t)^2}{4} \ge 0, \quad \forall\, p \ge -t
$$

Using similar procedures as when $t \ge 0$, we know the minimum is:

$$
\begin{cases}
-t/2 + \frac{3\sigma^2 + t^2}{4\sqrt{3}\sigma}, & -\sigma/\sqrt{3} \le t \le 0 \\
-t - \frac{\sigma^2}{9t}, & t \le -\sigma/\sqrt{3}
\end{cases}
$$

Combining the above two cases finishes the proof. ∎

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