



A NOTE ON THE UPPER BOUNDS FOR THE DISPERSION

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ABSTRACT. In this note we provide upper bounds for the standard deviation ($\sigma(X)$), for the quantity $\sigma^2(X) + (x - \mathbb{E}(X))^2$ and for the L^p absolute deviation of a random variable. These improve and extend current results in the literature.

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1. INTRODUCTION

Let $f : [a, b] \subset \mathbb{R} \rightarrow \mathbb{R}_+$ be the probability density function (p.d.f.) of the random variable X and $\mathbb{E}(X)$ and $\sigma(X)$ the mean and standard deviation respectively.

In [2], the authors gave upper bounds for the dispersion $\sigma(X)$ and for $\sigma^2(X) + (x - \mathbb{E}(X))^2$ in terms of the p.d.f. for a continuous random variable. Recently, Agbeko [1] also obtained some upper bounds for the same two quantities, namely,

$$(1.1) \quad \sigma(X) \leq \min\{\max\{|a|, |b|\}, (b - a)\},$$

and

$$(1.2) \quad \sqrt{\sigma^2(X) + (x - \mathbb{E}(X))^2} \leq 2 \min\{\max\{|a|, |b|\}, (b - a)\},$$

for all $x \in [a, b]$.

Both the work of Barnett et al. [2] and Agbeko [1] exclude the discrete case. In this present note, we remove this exclusion and improve the bounds of (1.1) and (1.2) in Section 2. In Section 3, we consider the L^p absolute deviation. The symbolism 1_A denotes the indicator function on the set A .

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2. STANDARD DEVIATION

Let \mathbb{P} be the distribution of the random variable X , then the expectation of X can be denoted by

$$\mathbb{E}(X) = \int_a^b x d\mathbb{P}(x)$$

and the dispersion or standard deviation of X by

$$\sigma(X) = \sqrt{\int_a^b x^2 d\mathbb{P}(x) - \left(\int_a^b x d\mathbb{P}(x)\right)^2}.$$

Theorem 2.1. *Let X be a random variable with $a \leq X \leq b$. Then we have*

$$(2.1) \quad \sigma(X) \leq \min\{\max\{|a|, |b|\}, M(a, b, \mathbb{E}(X))(b - a)\},$$

where

$$M(a, b, \mathbb{E}(X)) = \frac{1}{\sqrt{2}} \sqrt{2 - \exp\left\{-\frac{2(b - \mathbb{E}(X))^2}{(b - a)^2}\right\} - \exp\left\{-\frac{2(\mathbb{E}(X) - a)^2}{(b - a)^2}\right\}}.$$

Proof. Since $\sigma^2(X) = \mathbb{E}(X - \mathbb{E}(X))^2$, by the Fubini's theorem, we have

$$(2.2) \quad \begin{aligned} \mathbb{E}(X - \mathbb{E}(X))^2 &= \mathbb{E} \int_0^{(X - \mathbb{E}(X))^2} ds \\ &= \mathbb{E} \int_0^\infty 1_{\{(X - \mathbb{E}(X))^2 \geq s\}} ds \\ &= \int_0^\infty \mathbb{P}((X - \mathbb{E}(X))^2 \geq s) ds \\ &= \int_0^{(b - \mathbb{E}(X))^2} \mathbb{P}(X - \mathbb{E}(X) \geq \sqrt{s}) ds \\ &\quad + \int_0^{(\mathbb{E}(X) - a)^2} \mathbb{P}(-(X - \mathbb{E}(X)) \geq \sqrt{s}) ds. \end{aligned}$$

For any $\lambda > 0$, we have

$$\begin{aligned} \mathbb{E}e^{\lambda(X - \mathbb{E}(X))} &\leq e^{-\lambda\mathbb{E}(X)} \left(\frac{b - \mathbb{E}(X)}{b - a} e^{\lambda a} + \frac{\mathbb{E}(X) - a}{b - a} e^{\lambda b} \right) \\ &= e^{\lambda(a - \mathbb{E}(X))} \left(1 - \frac{\mathbb{E}(X) - a}{b - a} + \frac{\mathbb{E}(X) - a}{b - a} e^{\lambda(b - a)} \right) \\ &\triangleq e^{L(\lambda)}, \end{aligned}$$

where

$$L(\lambda) = \lambda(a - \mathbb{E}(X)) + \log \left(1 - \frac{\mathbb{E}(X) - a}{b - a} + \frac{\mathbb{E}(X) - a}{b - a} e^{\lambda(b - a)} \right).$$

The first two derivatives of $L(\lambda)$ are

$$\begin{aligned} L'(\lambda) &= a - \mathbb{E}(X) + \frac{(\mathbb{E}(X) - a)e^{\lambda(b - a)}}{\left(1 - \frac{\mathbb{E}(X) - a}{b - a} + \frac{\mathbb{E}(X) - a}{b - a} e^{\lambda(b - a)}\right)} \\ &= a - \mathbb{E}(X) + \frac{\mathbb{E}(X) - a}{\left(\left(1 - \frac{\mathbb{E}(X) - a}{b - a}\right)e^{-\lambda(b - a)} + \frac{\mathbb{E}(X) - a}{b - a}\right)}, \end{aligned}$$

$$L''(\lambda) = \frac{(b - \mathbb{E}(X))(\mathbb{E}(X) - a)e^{-\lambda(b-a)}}{\left(\left(1 - \frac{\mathbb{E}(X)-a}{b-a}\right)e^{-\lambda(b-a)} + \frac{\mathbb{E}(X)-a}{b-a} \right)^2}.$$

Noting that

$$L''(\lambda) \leq \frac{(b - a)^2}{4},$$

then by Taylor's formula, we have

$$L(\lambda) \leq L(0) + L'(0)\lambda + \frac{(b - a)^2}{8}\lambda^2 = \frac{(b - a)^2}{8}\lambda^2.$$

Therefore from Markov's inequality, we have

$$\begin{aligned} (2.3) \quad \mathbb{P}(X - \mathbb{E}(X) \geq \sqrt{s}) &\leq \inf_{\lambda > 0} e^{-\sqrt{s}\lambda} \mathbb{E}e^{\lambda(X - \mathbb{E}(X))} \\ &\leq \inf_{\lambda > 0} \exp \left\{ -\sqrt{s}\lambda + \frac{(b - a)^2}{8}\lambda^2 \right\} \\ &= \exp \left\{ -\frac{2s}{(b - a)^2} \right\}. \end{aligned}$$

Similarly,

$$\mathbb{P}(-(X - \mathbb{E}(X)) \geq \sqrt{s}) \leq \exp \left\{ -\frac{2s}{(b - a)^2} \right\}.$$

From (2.2), it follows that

$$\begin{aligned} (2.4) \quad \mathbb{E}(X - \mathbb{E}(X))^2 &= \int_0^{(b - \mathbb{E}(X))^2} \mathbb{P}((X - \mathbb{E}(X)) \geq \sqrt{s}) ds + \int_0^{(\mathbb{E}(X) - a)^2} \mathbb{P}(-(X - \mathbb{E}(X)) \geq \sqrt{s}) ds \\ &\leq \int_0^{(b - \mathbb{E}(X))^2} \exp \left\{ -\frac{2s}{(b - a)^2} \right\} ds + \int_0^{(\mathbb{E}(X) - a)^2} \exp \left\{ -\frac{2s}{(b - a)^2} \right\} ds \\ &= \frac{(b - a)^2}{2} \left(2 - \exp \left\{ -\frac{2(b - \mathbb{E}(X))^2}{(b - a)^2} \right\} - \exp \left\{ -\frac{2(\mathbb{E}(X) - a)^2}{(b - a)^2} \right\} \right). \end{aligned}$$

Furthermore,

$$\mathbb{E}(X - \mathbb{E}(X))^2 = \mathbb{E}(X^2) - (\mathbb{E}(X))^2 \leq \mathbb{E}(X^2) \leq \max\{|a|^2, |b|^2\},$$

which, by (2.4), implies the result. □

Remark 2.2. In fact, by,

$$\begin{aligned} M(a, b, \mathbb{E}(X)) &= \frac{1}{\sqrt{2}} \sqrt{2 - \exp \left\{ -\frac{2(b - \mathbb{E}(X))^2}{(b - a)^2} \right\} - \exp \left\{ -\frac{2(\mathbb{E}(X) - a)^2}{(b - a)^2} \right\}} \\ &\leq \frac{1}{\sqrt{2}} \sqrt{2 - e^{-2}} < 1, \end{aligned}$$

$$\min\{\max\{|a|, |b|\}, M(a, b, \mathbb{E}(X))(b - a)\} \leq \min\{\max\{|a|, |b|\}, (b - a)\},$$

and is, therefore, a tighter bound than (1.1). If $\mathbb{E}(X) = (b + a)/2$, then

$$\begin{aligned} M(a, b, \mathbb{E}(X)) &= \frac{1}{\sqrt{2}} \sqrt{2 - \exp \left\{ -\frac{2(b - \mathbb{E}(X))^2}{(b - a)^2} \right\} - \exp \left\{ -\frac{2(\mathbb{E}(X) - a)^2}{(b - a)^2} \right\}} \\ &= \sqrt{1 - e^{-1/2}} < \frac{1}{\sqrt{2}}, \end{aligned}$$

which means that under some assumptions, this bound is better than the result of [2] for the case $f \in L_1([a, b])$.

Corollary 2.3. *Let X be a random variable with $a \leq X \leq b$, then for any $x \in [a, b]$,*

$$(2.5) \quad \sqrt{\sigma^2(X) + (x - \mathbb{E}(X))^2} \leq \min\{2 \max\{|a|, |b|\}, N(a, b, \mathbb{E}(X))(b - a)\},$$

where

$$N^2(a, b, \mathbb{E}(X)) = 2 - \frac{1}{2} \left(\exp \left\{ -\frac{2(b - \mathbb{E}(X))^2}{(b - a)^2} \right\} + \exp \left\{ -\frac{2(\mathbb{E}(X) - a)^2}{(b - a)^2} \right\} \right).$$

Proof. It is clear that $(x - \mathbb{E}(X))^2 \leq (b - a)^2$ and from the proof of Theorem 2.1, we know that

$$\sigma^2(X) + (x - \mathbb{E}(X))^2 \leq \left(2 - \frac{1}{2} \left(\exp \left\{ -\frac{2(b - \mathbb{E}(X))^2}{(b - a)^2} \right\} + \exp \left\{ -\frac{2(\mathbb{E}(X) - a)^2}{(b - a)^2} \right\} \right) \right) (b - a)^2.$$

Furthermore,

$$\sigma^2(X) + (x - \mathbb{E}(X))^2 = \mathbb{E}(X - x)^2 \leq \max\{(x - y)^2, x, y \in [a, b]\}.$$

The remainder of the proof follows as Theorem 1.2 in Agbeko [1], giving

$$\max\{|x - y|, x, y \in [a, b]\} \leq 2 \min\{\max\{|a|, |b|\}, (b - a)\}.$$

□

3. L^p ABSOLUTE DEVIATION

In fact by the method in Section 2, we could extend the case of L^p absolute deviation, i.e., the following quantity,

$$\sigma_p(X) = \mathbb{E}(|X - \mathbb{E}(X)|^p)^{1/p}.$$

We have the following

Theorem 3.1. *Let X be a random variable and $p \geq 1$. Assume that $\mathbb{E}|X|^p < \infty$, then its L^p absolute deviation has the following estimation for $p > 2$,*

$$(\sigma_p(X))^p \leq \min\{|b - a|^p, M_1(a, b, p, X)\},$$

where

$$(3.1) \quad M_1(a, b, p, X) = \frac{(b - a)^2}{2} \left(1 - \exp \left\{ -\frac{2(|b - \mathbb{E}(X)|^p \wedge 1)}{(b - a)^2} \right\} \right) \\ + (|b - \mathbb{E}(X)|^p - |b - \mathbb{E}(X)|^p \wedge 1) \\ \times \exp \left\{ -\frac{2(|b - \mathbb{E}(X)|^p \wedge 1)^{2/p}}{(b - a)^2} \right\} \\ + \frac{(b - a)^2}{2} \left(1 - \exp \left\{ -\frac{2(|a - \mathbb{E}(X)|^p \wedge 1)}{(b - a)^2} \right\} \right) \\ + (|a - \mathbb{E}(X)|^p - |a - \mathbb{E}(X)|^p \wedge 1) \\ \times \exp \left\{ -\frac{2(|a - \mathbb{E}(X)|^p \wedge 1)^{2/p}}{(b - a)^2} \right\}.$$

If $1 \leq p < 2$, then we have

$$(\sigma_p(X))^p \leq \min\{|b - a|^p, M_2(a, b, p, X)\},$$

where

$$(3.2) \quad M_2(a, b, p, X) \\ = |b - \mathbb{E}(X)|^p \wedge 1 + \frac{(b - a)^2}{2} \left(\exp \left\{ -\frac{2|b - \mathbb{E}(X)|^p \wedge 1}{(b - a)^2} \right\} - \exp \left\{ -\frac{2|b - \mathbb{E}(X)|^p}{(b - a)^2} \right\} \right) \\ + |a - \mathbb{E}(X)|^p \wedge 1 + \frac{(b - a)^2}{2} \left(\exp \left\{ -\frac{2|a - \mathbb{E}(X)|^p \wedge 1}{(b - a)^2} \right\} - \exp \left\{ -\frac{2|a - \mathbb{E}(X)|^p}{(b - a)^2} \right\} \right).$$

Proof. On one hand, by the Fubini's theorem, we have

$$(3.3) \quad \mathbb{E}|X - \mathbb{E}(X)|^p = \mathbb{E} \int_0^{|X - \mathbb{E}(X)|^p} ds \\ = \mathbb{E} \int_0^\infty 1_{\{|X - \mathbb{E}(X)|^p \geq s\}} ds \\ = \int_0^\infty \mathbb{P}(|X - \mathbb{E}(X)|^p \geq s) ds \\ = \int_0^{|b - \mathbb{E}(X)|^p} \mathbb{P}((X - \mathbb{E}(X)) \geq s^{1/p}) ds \\ + \int_0^{|\mathbb{E}(X) - a|^p} \mathbb{P}(-(X - \mathbb{E}(X)) \geq s^{1/p}) ds.$$

From the estimation (2.3), we have

$$\mathbb{E}|X - \mathbb{E}(X)|^p = \int_0^{|b - \mathbb{E}(X)|^p} \mathbb{P}(X - \mathbb{E}(X) \geq s^{1/p}) ds + \int_0^{|\mathbb{E}(X) - a|^p} \mathbb{P}(-(X - \mathbb{E}(X)) \geq s^{1/p}) ds \\ \leq \int_0^{|b - \mathbb{E}(X)|^p} \exp \left\{ -\frac{2s^{2/p}}{(b - a)^2} \right\} ds + \int_0^{|\mathbb{E}(X) - a|^p} \exp \left\{ -\frac{2s^{2/p}}{(b - a)^2} \right\} ds.$$

(Case: $p > 2$.) By simple calculating, we have

$$\int_0^{|b - \mathbb{E}(X)|^p} \exp \left\{ -\frac{2s^{2/p}}{(b - a)^2} \right\} ds \\ \leq \int_0^{|b - \mathbb{E}(X)|^p \wedge 1} \exp \left\{ -\frac{2s}{(b - a)^2} \right\} ds + \int_{|b - \mathbb{E}(X)|^p \wedge 1}^{|b - \mathbb{E}(X)|^p} \exp \left\{ -\frac{2s^{2/p}}{(b - a)^2} \right\} ds \\ \leq \frac{(b - a)^2}{2} \left(1 - \exp \left\{ -\frac{2(|b - \mathbb{E}(X)|^p \wedge 1)}{(b - a)^2} \right\} \right) \\ + (|b - \mathbb{E}(X)|^p - |b - \mathbb{E}(X)|^p \wedge 1) \exp \left\{ -\frac{2(|b - \mathbb{E}(X)|^p \wedge 1)^{2/p}}{(b - a)^2} \right\}$$

and with the same reason

$$\int_0^{|\mathbb{E}(X) - a|^p} \exp \left\{ -\frac{2s^{2/p}}{(b - a)^2} \right\} ds \leq \frac{(b - a)^2}{2} \left(1 - \exp \left\{ -\frac{2(|a - \mathbb{E}(X)|^p \wedge 1)}{(b - a)^2} \right\} \right) \\ + (|a - \mathbb{E}(X)|^p - |a - \mathbb{E}(X)|^p \wedge 1) \exp \left\{ -\frac{2(|a - \mathbb{E}(X)|^p \wedge 1)^{2/p}}{(b - a)^2} \right\}.$$

(Case $1 \leq p < 2$) In this case, we have

$$\begin{aligned} & \int_0^{|b-\mathbb{E}(X)|^p} \exp\left\{-\frac{2s^{2/p}}{(b-a)^2}\right\} ds \\ & \leq \int_0^{|b-\mathbb{E}(X)|^p \wedge 1} \exp\left\{-\frac{2s^{2/p}}{(b-a)^2}\right\} ds + \int_{|b-\mathbb{E}(X)|^p \wedge 1}^{|b-\mathbb{E}(X)|^p} \exp\left\{-\frac{2s}{(b-a)^2}\right\} ds \\ & \leq |b-\mathbb{E}(X)|^p \wedge 1 + \frac{(b-a)^2}{2} \left(\exp\left\{-\frac{2|b-\mathbb{E}(X)|^p \wedge 1}{(b-a)^2}\right\} - \exp\left\{-\frac{2|b-\mathbb{E}(X)|^p}{(b-a)^2}\right\} \right) \end{aligned}$$

and with the same reason

$$\begin{aligned} & \int_0^{|\mathbb{E}(X)-a|^p} \exp\left\{-\frac{2s^{2/p}}{(b-a)^2}\right\} ds \leq |a-\mathbb{E}(X)|^p \wedge 1 \\ & \quad + \frac{(b-a)^2}{2} \left(\exp\left\{-\frac{2|a-\mathbb{E}(X)|^p \wedge 1}{(b-a)^2}\right\} - \exp\left\{-\frac{2|a-\mathbb{E}(X)|^p}{(b-a)^2}\right\} \right). \end{aligned}$$

On the other hand, for any $p \geq 1$, it is clear that

$$\mathbb{E}|X - \mathbb{E}(X)|^p \leq (b-a)^p.$$

From the above discussion, the desired results are obtained. \square

Remark 3.2. $M_1(a, b, p, X)$ and $M_2(a, b, p, X)$ are sometimes better than $(b-a)^p$, e.g., if $p > 2$, taking $\mathbb{E}(X) = (a+b)/2$ and letting $1/2 < (b-a)/2 < 1$, then

$$\begin{aligned} M_1(a, b, p, X) &= (b-a)^2 \left(1 - \exp\{-2^{1-p}(b-a)^{p-2}\}\right) \\ &< (b-a)^p \left(1 - \exp\{-2^{1-p}(b-a)^{p-2}\}\right) \\ &< (b-a)^p. \end{aligned}$$

Further, if $1 \leq p < 2$, taking $\mathbb{E}(X) = (a+b)/2$ and letting $(b-a)/2 < 1$, then

$$M_2(a, b, p, X) = 2 \left(\frac{(b-a)^p}{2^p} \right) = 2^{1-p}(b-a)^p \leq (b-a)^p.$$

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