

proof of Bernstein inequalities

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Entry type Proof Classification msc 60E15 1) By http://planetmath.org/ChernoffCramerBoundChernoff-Cramèr bound, we have:

$$\Pr\left\{\sum_{i=1}^{n} \left(X_{i} - E[X_{i}]\right) > \varepsilon\right\} \leq \exp\left[-\sup_{t>0} \left(t\varepsilon - \psi(t)\right)\right]$$

where

$$\psi(t) = \sum_{i=1}^{n} \left(\ln E \left[e^{tX_i} \right] - tE \left[X_i \right] \right)$$

Since $\ln x \le x - 1 \ \forall x \ge 0$,

$$\psi(t) = \sum_{i=1}^{n} \left(\ln E \left[e^{tX_i} \right] - tE \left[X_i \right] \right)
\leq \sum_{i=1}^{n} E \left[e^{tX_i} \right] - tE \left[X_i \right] - 1
= \sum_{i=1}^{n} E \left[1 + tX_i + \frac{1}{2} t^2 X_i^2 + \sum_{k=3}^{+\infty} \frac{t^k X_i^k}{k!} \right] - tE \left[X_i \right] - 1
= \sum_{i=1}^{n} \left(\frac{1}{2} t^2 E \left[X_i^2 \right] + \sum_{k=3}^{+\infty} \frac{t^k E \left[X_i^k \right]}{k!} \right)
= \frac{1}{2} t^2 \sum_{i=1}^{n} E \left[X_i^2 \right] + \sum_{k=3}^{+\infty} \frac{t^k \sum_{i=1}^{n} E \left[X_i^k \right]}{k!}
\leq \frac{1}{2} t^2 \sum_{i=1}^{n} E \left[X_i^2 \right] + \sum_{k=3}^{+\infty} \frac{t^k \sum_{i=1}^{n} E \left[|X_i|^k \right]}{k!},$$

and, keeping in mind hypotheses a) and b),

$$\psi(t) \leq \frac{1}{2}t^2v^2 + \sum_{k=3}^{+\infty} \frac{t^k}{2}v^2c^{k-2} = \frac{1}{2}t^2v^2 + \frac{1}{2}t^3v^2c\sum_{k=0}^{+\infty} (tc)^k$$

Now, if tc < 1, we obtain

$$\psi(t) \le \frac{1}{2}t^2v^2\left(1 + \frac{tc}{1 - tc}\right) = \frac{v^2t^2}{2(1 - tc)}$$

whence

$$\sup_{t>0} (t\varepsilon - \psi(t)) \ge \sup_{0 < t < \frac{1}{c}} \left(t\varepsilon - \frac{v^2 t^2}{2(1 - tc)} \right)$$

By elementary calculus, we obtain the value of t that maximizes the expression in brackets (out of the two roots of the second degree polynomial equation, we choose the one which is $<\frac{1}{c}$):

$$t_{opt} = \frac{v^2 + 2c\varepsilon - v^2\sqrt{1 + \frac{2c\varepsilon}{v^2}}}{c\left(v^2 + 2c\varepsilon\right)} = \frac{1}{c}\left(1 - \frac{1}{\sqrt{1 + \frac{2c\varepsilon}{v^2}}}\right)$$

which, once plugged into the bounds, yields

$$\Pr\left\{\sum_{i=1}^{n} \left(X_i - E[X_i]\right) > \varepsilon\right\} \le \exp\left[-\frac{v^2}{c^2} \left(1 + \frac{c\varepsilon}{v^2} - \sqrt{1 + 2\frac{c\varepsilon}{v^2}}\right)\right]$$

Observing that $\sqrt{1+x} \le 1 + \frac{1}{2}x$, one gets:

$$t_{opt} = \frac{1}{c} \left(1 - \frac{1}{\sqrt{1 + \frac{2c\varepsilon}{v^2}}} \right) \le \frac{1}{c} \left(1 - \frac{1}{1 + \frac{c\varepsilon}{v^2}} \right) = \frac{\varepsilon}{v^2 + c\varepsilon} = t' < \frac{1}{c}$$

Plugging t' in the bound expression, the sub-optimal yet more easily manageable formula is obtained:

$$\Pr\left\{\sum_{i=1}^{n} \left(X_{i} - E[X_{i}]\right) > \varepsilon\right\} \leq \exp\left(-\frac{\varepsilon^{2}}{2\left(v^{2} + c\varepsilon\right)}\right)$$

which is obviously a worse bound than the preceding one, since $t' \neq t_{opt}$. One can also verify the consistency of this inequality directly proving that, for any $x \geq 0$,

$$1 + x - \sqrt{1 + 2x} \ge \frac{x^2}{2(1+x)}$$

(see http://planetmath.org/ASimpleMethodForComparingRealFunctionshere for an easy way, which can be used with $x_0 = 0$)

2) To prove this more specialized statement let's recall that the condition

$$\Pr\{|X_i| \le M\} = 1 \ \forall i$$

implies that, for all i,

$$E[|X_i|^k] \le M^k \ \forall k \ge 0$$

 $(See \ http://planetmath.org/RelationBetween \verb|AlmostSurely| Absolutely Bounded Random Variable for a proof.)$

Now, it's enough to verify that the condition

$$E[|X_i|^k] \le M^k$$

imply both conditions a) and b) in part 1).

Indeed, part a) is obvious, while for part b) one happens to have:

$$E[|X_i|^k] \le E\left[X_i^2\right] M^{k-2}$$

(see http://planetmath.org/AbsoluteMomentsBoundingNecessaryAndSufficientConditionhe for a proof).

So

$$\sum_{i=1}^{n} E[|X_i|^k] \le \sum_{i=1}^{n} E[X_i^2] M^{k-2} = v^2 M^{k-2}$$

Let's find a value for c such that $v^2M^{k-2} \leq \frac{k!}{2}v^2c^{k-2}$, thus satisfying part b) of the hypotheses.

After simplifying, we have to study the inequality

$$k!c^{k-2} > 2 \cdot M^{k-2}$$

for any $k \geq 3$. Let's proceed by induction. For k = 3, we have

which suggests $c = \frac{M}{3}$. Let's now verify if this position is consistent with the inductive hypothesis:

$$(k+1)! = (k+1)k! \ge (k+1) \cdot 2 \cdot 3^{k-2} \ge 3 \cdot 2 \cdot 3^{k-2} = 2 \cdot 3^{(k+1)-2}$$

which confirms the validity of the choice $c = \frac{M}{3}$, which has to be plugged into the former bound to obtain the new one.

[to be continued...]