plain definition remark

Useful Inequalities in Probability and Statistics

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1 Tails of the normal distribution

Theorem 1.1 Let $X \sim \mathcal{N}(0,1)$. Define $\overline{\Phi}(x) := P(X > x), x \ge 0$. Then

$$\frac{1}{2}e^{-x\sqrt{2/\pi}-x^2/2} \le \overline{\Phi}(x) \le \min\left(\frac{1}{2}e^{-x^2/2}, \frac{1}{x}\phi(x)\right), \ x \ge 0$$

We have, by continuity and Corollary 1.4:

$$2\overline{\Phi}(x) = \lim_{y \downarrow 0} \frac{\overline{\Phi}(x+y)}{\overline{\Phi}(y)} \left\{ \leq \lim_{y \downarrow 0} e^{-xy - x^2/2} = e^{-x^2/2} \geq \lim_{y \downarrow 0} e^{-\rho(y)x - x^2/2} = e^{-x\sqrt{2/\pi} - x^2/2} \right\}$$

We conclude for the upper bound with Lemma 1.2.

Remark 1 Note that Proposition 1.1 yields a stronger version of the infimum Chernoff bound which is found with

$$\overline{\Phi}(x) \le \inf_{\lambda > 0} e^{-\lambda x} e^{\lambda^2/2} = e^{-x^2/2}, \ x \ge 0$$

where the second equality follows from the quadratic minimization condition $\frac{d}{d\lambda}(-\lambda x + \lambda^2/2) = 0$ which yields $\lambda = x$.

1.1 Lemmas for the proof of Theorem 1.1 - mostly from [?]; Appendix D

Lemma 1.2 Let $\phi(x) := (2\pi)^{-1/2}e^{-x^2/2}$. Then:

$$\left(\frac{1}{x} - \frac{1}{x^3}\right)\phi(x) < \overline{\Phi}(x) < \frac{1}{x}\phi(x), x > 0 \tag{1}$$

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We have $\overline{\Phi}(x)=\int_x^\infty\phi(t)dt$. Note that for all t>0 the functions $1-3/t^4<1$ and $1+1/t^2>1$ so $(1-3/t^4)\phi(t)<\phi(t)<(1+1/t^2)\phi(t)$. We thus integrate from x>0 to infinity and obtain the claimed bounds. It follows immediately that for the function $\rho(x):=\phi(x)/\overline{\Phi}(x), \ x>0$ we have the bounds

$$x < \rho(x) < \frac{x^3}{x^2 - 1}, x > 1$$

Theorem 1.3 ρ is increasing, $\rho(-\infty)=0$ and $\rho(0)=\sqrt{2/\pi}$. The function $\rho(x)-x$ decreases to zero as x tends to infinity. The function $\ln \rho(x)$ is concave and $\ln \rho(x+\delta) < \ln \rho(x) + (\rho(x)-x)\delta$ for $x \in R$ and $\delta > 0$.

We have

$$\frac{1}{\rho(x)} = \frac{\overline{\Phi}(x)}{\phi(x)} = \int_x^\infty \frac{\phi(t)}{\phi(x)} dt \stackrel{z=t-x}{=} \int_0^\infty \frac{\phi(z+x)}{\phi(x)} dz = \int_0^\infty e^{-zx} e^{-z^2/2} dz$$

That is, the Laplace transform of the finite measure $\mu(dz):=e^{-z^2/2}dz, z>0$. From this it follows that $\rho(0)=\sqrt{2/\pi}$ and that $\rho(-\infty)=0$, since by monotone convergence $\lim_{x\to -\infty}\frac{1}{\rho(x)}=\infty$. For $x\geq 0$ we have $ze^{-zx}\leq z, z>0$ and $z\in L^1(d\mu)$. So by differentiation-under-integral lemma we have

$$\rho(x) \left(\frac{d}{dx} \frac{1}{\rho(x)}\right) = \rho(x) \left(-\int_0^\infty z e^{-zx} \mu(dz)\right) < 0 \\ \rho(x) \left(\frac{d^2}{dx^2} \frac{1}{\rho(x)}\right) = \rho(x) \int_0^\infty z^2 e^{-zx} \mu(dz) > 0$$

The function $1/\rho$ is decreasing because $\frac{d}{dx}\frac{1}{\rho(x)}<0$, so that ρ is increasing. Now, we have:

$$\ln \rho(x) = \ln(\phi(x)) - \ln(\overline{\Phi}(x)) = -\ln(\sqrt{2\pi}) - \frac{x^2}{2} - \ln(\overline{\Phi}(x))$$

Therefore

$$\frac{d}{dx}\ln\rho(x) = -x - \frac{1}{\overline{\Phi}(x)}\frac{d}{dx}\overline{\Phi}(x) = \frac{\phi(x)}{\overline{\Phi}(x)} - x = \rho(x) - x$$

But we also have $\frac{d}{dx}(-\ln\rho(x)) = \frac{d}{dx}\ln(1/\rho(x))$ so

$$\frac{d}{dx}(\rho(x)-x) = -\frac{d^2}{dx^2}\ln\frac{1}{\rho(x)} = -\frac{d}{dx}\bigg(\rho(x)\bigg(\frac{d}{dx}\frac{1}{\rho(x)}\bigg)\bigg) = -\bigg(\rho(x)\bigg(\frac{d^2}{dx^2}\frac{1}{\rho(x)}\bigg) + \bigg(\bigg(\frac{d}{dx}\rho(x)\bigg)\bigg\bigg(\frac{d}{dx}\frac{1}{\rho(x)}\bigg)\bigg)\bigg)$$

and

$$\frac{d}{dx}\frac{1}{\rho(x)} = -\frac{d}{dx}\rho(x)\frac{1}{\rho(x)^2}\frac{d}{dx}\rho(x) = -\rho(x)^2\left(\frac{d}{dx}\frac{1}{\rho(x)}\right)$$

Therefore

$$\frac{d}{dx}(\rho(x) - x) = -\left(\rho(x)\left(\frac{d^2}{dx^2}\frac{1}{\rho(x)}\right) - \rho(x)^2\left(\frac{d}{dx}\frac{1}{\rho(x)}\right)^2\right)$$

The term in the parenthesis in the variance of the measure $\nu_x(dz) := \rho(x)e^{-zx}\mu(dz)$, which is strictly positive. In fact for any $x \geq 0$, $\nu_x(dz)$ is a probability measure:

$$\int_0^\infty \nu_x(dz) = \rho(x) \int_0^\infty e^{-zx-z^2/2} dz = \rho(x) \int_x^\infty \frac{\phi(t)}{\phi(x)} dt = 1$$

Thus, $\rho(x)-x$ is decreasing because its first derivative is strictly negative. Now since

$$\frac{d^2}{dx^2}\ln\rho(x) = \frac{d}{dx}(\rho(x) - x) < 0$$

then the function $\ln \rho(x)$ is thus concave. To see that $\rho(x)-x$ vanishes as $x\to\infty$:

$$0 < \rho(x) - x < \frac{x}{x^2 - 1} \stackrel{x \to \infty}{\to} 0$$

It remains to show the last claim. For $\delta>0$ and some $x^*\in(x,x+\delta)$, since $\rho(x)-x$ is decreasing:

$$\ln \frac{\rho(x+\delta)}{\rho(x)} = \delta \frac{\ln \rho(x+\delta) - \ln \rho(x)}{\delta} \stackrel{\text{MVT}}{=} \delta(\rho(x^*) - x^*) < \delta(\rho(x) - x)$$

and we conclude.

Corollary 1.4 For $x \ge 0$ and $\delta > 0$:

$$e^{-\rho(x)\delta-\delta^2/2} \le \frac{\overline{\Phi}(x+\delta)}{\overline{\Phi}(x)} \le e^{-x\delta-\delta^2/2}$$

We have

$$\overline{\Phi}(x+\delta) = \int_{x}^{\infty} \phi(z+\delta)dz = \frac{e^{-\delta^2/2}}{\sqrt{2\pi}} \int_{x}^{\infty} e^{-z^2/2 - z\delta}dz \overset{e^{-z\delta} \le e^{-x\delta}, \forall z \ge x}{\le} \frac{e^{-\delta^2/2 - x\delta}}{\sqrt{2\pi}} \int_{x}^{\infty} e^{-z^2/2}dz = e^{-\delta^2/2 - x\delta} \overline{\Phi}(x)$$

By Theorem 1.3 we have $\ln rac{
ho(x+\delta)}{
ho(x)} < (
ho(x)-x)\delta.$ We also have

$$\frac{\overline{\Phi}(x+\delta)}{\overline{\Phi}(x)} = \frac{\overline{\Phi}(x+\delta)}{\phi(x+\delta)} \frac{\phi(x)}{\overline{\Phi}(x)} \frac{\phi(x+\delta)}{\phi(x)} = \frac{\rho(x+\delta)}{\rho(x)} \frac{\phi(x+\delta)}{\phi(x)} = e^{-\ln\frac{\rho(x+\delta)}{\rho(x)}} e^{-\delta^2/2 - x\delta} \ge e^{-\rho(x)\delta - \delta^2/2}$$

and we conclude.

1.2 Law of iterated logarithm - mostly from [?], Chapters 11.1 and 11.2

Lemma 1.5 Let $S_n = X_1 + ... + X_n$ for $(X_n)_{n \le N}$ IID $\sim \mathcal{N}(0,1)$. Then for $x \ge 0$

$$P(\max_{n \le N} S_n \ge x) \le 2P(S_N \ge x)$$

We have $P(S_N-S_n\geq 0)=1/2$ for any n< N by the fact that $S_N-S_n\sim \mathcal{N}(0,N-n)$. Define $\tau=\inf\{n\leq N:S_n\geq x\}\wedge N$. The events $\{\tau=n\},n\in\{1,...,N\}$ are disjoint. So we have

$$P(\max_{n \leq N} S_n \geq x) = P(\cup_{n \leq N} \{\tau = n\}) = P(\cup_{n \leq N} (\{\tau = n\} \cap \{S_n \geq x\})) = \sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n < N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau = n\} \cap \{S_n \geq x\}) = 2\sum_{n \leq N} P(\{\tau =$$

where equality (*) follows from the fact that S_N-S_n is independent of S_n , while the inequality (**) follows from the fact that $\{S_n \geq x\} \cap \{S_N-S_n \geq 0\} \subseteq \{S_N \geq x\}$ and $P(\{\tau=N\} \cap \{S_N \geq x\}) \leq 2P(\{\tau=N\} \cap \{S_N \geq x\})$.

Theorem 1.6 Let $S_n = X_1 + ... + X_n$ where $(X_n)_{n \in \mathbb{N}}$ are IID $\sim \mathcal{N}(0,1)$. Then:

- 1. $\limsup_{n\to\infty} \frac{S_n}{\sqrt{2n\ln\ln n}} = 1$ a.s.;
- 2. $\liminf_{n\to\infty} \frac{S_n}{\sqrt{2n\ln\ln n}} = -1$ a.s.;
- 3. $\frac{S_n}{\sqrt{2n\ln \ln n}} \in B$ i.o. a.s. for any open $B \subseteq [-1,1]$.
- (i). Let B be a block of consecutive positive integers (i.e. $B=\{n+1,...,n+k\}$ for some n,k). Let $\gamma>1$. Then by Lemma 1.5 and Theorem 1.2

$$P(\cup_{\ell \in B} \{ S_{\ell} \ge \gamma \sqrt{2\ell \ln \ln \ell} \}) \le P(\max_{\ell \in B} S_{\ell} \ge \gamma \sqrt{2n \ln \ln n}) \le 2P(S_{n+k} \ge \gamma \sqrt{2n \ln \ln n}) = 2P\left(\frac{S_{n+k}}{\sqrt{n+k}} \ge \gamma \frac{\sqrt{2n \ln \ln n}}{\sqrt{n+k}}\right)$$

Now consider a sequence of consecutive blocks $B_k = \{n: n_k < n \le n_{k+1}\}$ so that $n_k/\rho^k \to 1$ for some constant $\rho \in (1,\gamma)$. By the above then we then have the bounds

$$P(\cup_{\ell \in B_k} \{S_\ell \geq \gamma \sqrt{2\ell \ln \ln \ell}\}) \leq \exp\left(-\frac{1}{2}\rho^2 \frac{2n_k \ln \ln n_k}{n_{k+1}}\right)^k \underset{\boldsymbol{\approx}}{\text{large enough}} \exp\left(-\frac{1}{2}\rho^2 \frac{2\rho^k \ln \ln \rho^k}{\rho^{k+1}}\right) = \exp\left(-\rho(\ln n_k)\right)^k \frac{1}{n_{k+1}} \exp\left(-\frac{1}{2}\rho^2 \frac{2\rho^k \ln \ln \rho^k}{\rho^{k+1}}\right) = \exp\left(-\frac{1}{2}\rho^2 \frac{2\rho^k \ln \ln \rho^k}{\rho^{k+1}}\right)$$

which implies, by Borel-Cantelli I, that $P(S_n \geq \gamma \sqrt{2n \ln \ln n} \text{ i.o.}) = 0$. Therefore $P(\limsup_{n \to \infty} \frac{S_n}{\sqrt{2n \ln \ln n}} \leq \gamma) = 1$ for any $\gamma > 1$. We conclude that, considering a sequence $\gamma_\ell \downarrow 1$,

$$P\bigg(\limsup_{n\to\infty}\frac{S_n}{\sqrt{2n\ln\ln n}}\le 1\bigg)=1$$

We now need to prove that $\limsup_{n\to\infty} \frac{S_n}{\sqrt{2n\ln\ln n}} \geq 1$

References

[1] Pollard, D. (2001). A User's Guide to Measure Theoretic Probability. Cambridge University Press.