

Supplementary Material of “Stochastic Approximations to the Pitman–Yor Process”

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Appendix A: Random generation of $T_{\alpha,\theta}$

Let Y be a standard $\text{Exp}(1)$ random variable. Note that

$$\begin{aligned} P((Y/T_\alpha)^\alpha > x) &= \int_0^\infty P(Y > x^{1/\alpha}t) f_\alpha(t) dt = \int_0^\infty \exp[-x^{1/\alpha}t] f_\alpha(t) dt \\ &= E[e^{-x^{1/\alpha}T_\alpha}] = e^{-x} = P(Y > x) \end{aligned}$$

so we have $Y =_d (Y/T_\alpha)^\alpha$. For $r < \alpha$, $E(Y^{-r/\alpha}) < \infty$, so we find that $E(Y^{-r/\alpha}) = E(T_\alpha^r)E(Y^{-r})$ and

$$E(T_\alpha^r) = \frac{E(Y^{-r/\alpha})}{E(Y^{-r})} = \frac{\Gamma(1 - r/\alpha)}{\Gamma(1 - r)}. \quad (1)$$

The normalizing constant in $f_{\alpha,\theta}(t)$ is $\int_0^\infty t^{-\theta} f_\alpha(t) dt = E(T_\alpha^{-\theta})$, so set $r = -\theta$ and note that $-\theta < \alpha$. Let G_a be a gamma random variable with shape $a > 0$ and unit rate. Simple moment comparisons using (1) yield the distributional equality $G_{1+\theta/\alpha} \stackrel{d}{=} (G_{1+\theta}/T_{\alpha,\theta})^\alpha$, which, however, does not provide a way to generate from $T_{\alpha,\theta}$. For this we resort to Devroye (2009). First we recall how to generate a Zolotarev random variable $Z_{\alpha,b}$ for $\alpha \in (0, 1)$ and $b = \theta/\alpha > -1$. Let

$$C = \frac{\Gamma(1 + b\alpha)\Gamma(1 + b(1 - \alpha))}{\pi\Gamma(1 + b)}$$

and

$$B(u) = A(u)^{-(1-\alpha)} = \frac{\sin(u)}{\sin(\alpha u)^\alpha \sin((1 - \alpha)u)^{1-\alpha}}.$$

A simple asymptotic argument yields the value $B(0) = \alpha^{-\alpha}(1 - \alpha)^{-(1-\alpha)}$. Then $f(x) = CB(x)^b$, $0 \leq x \leq \pi$. The following bound holds

$$f(x) \leq CB(0)^b e^{-\frac{x^2}{2\sigma^2}}, \quad \text{with } \sigma^2 = \frac{1}{b\alpha(1 - \alpha)}.$$

This Gaussian upper bound suggests a simple rejection sampler for sampling Zolotarev random variates. Following Devroye (2009), it is most efficient to adapt the sampler to

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the value of σ . If $\sigma \geq \sqrt{2\pi}$, rejection from a uniform random variate is best. Otherwise, use a normal dominating curve as suggested in the bound above. The details are given below.

ALGORITHM 3 (Sampler of $T_{\alpha,\theta}$)

1. set $b = \theta/\alpha$ and $\sigma = \sqrt{b\alpha(1-\alpha)}$
2. if $\sigma \geq \sqrt{2\pi}$:
 - then repeat: generate $U \sim \text{Unif}(0, \pi)$ and $V \sim \text{Unif}(0, 1)$.
 set $X \leftarrow U$, $W \leftarrow B(X)$.
 until $V \leq (W/B(0))^b$
 - else repeat: generate $N \sim \text{N}(0, 1)$ and $V \sim \text{Unif}(, 1)$.
 set $X \leftarrow \sigma|N|$, $W \leftarrow B(X)$.
 until $X \leq \pi$ and $Ve^{-N^2/2} \leq (W/B(0))^b$
3. generate $G \stackrel{d}{=} G_{1+b(1-\alpha)/\alpha}$
4. set $T \leftarrow 1/(WG^{1-\alpha})^{1/\alpha}$
5. return T

References

- Devroye, L. (2009). “Random variate generation for exponentially and polynomially tilted stable distributions.” *ACM Transactions on Modeling and Computer Simulation (TOMACS)*, 19(4): Article No. 18. 1