

Pathological Properties of Deep Bayesian Hierarchies

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Overview

- In hierarchical models, we need a distribution over the latent parameters at each node
- Common solution: recursively draw from a distribution such as a Dirichlet process, beta process, Pitman-Yor process, etc.
- We show that for DP, BP, and GammaP, **this won't work for deep hierarchies**
- But!...Pitman-Yor is okay

Convergence of Martingale Sequences

- Consider the following sequences (thought of as parameters on a path down a hierarchy):

$$\begin{aligned} \theta_{n+1} | \theta_n &\sim \text{DP}(c\theta_n), & \theta_{n+1} | \theta_n &\sim \text{BP}(c\theta_n), \\ \theta_{n+1} | \theta_n &\sim \text{GammaP}(c\theta_n), & \theta_{n+1} | \theta_n &\sim \text{PYP}(c\theta_n) \end{aligned}$$

- All have the property that $E[\theta_{n+1} | \theta_n] = \theta_n$.
 - Called the *martingale property*
 - Philosophically desirable because it means that information is preserved as we move down the hierarchy
- **Theorem (Doob):** All non-negative martingale sequences have a limit with probability 1.

Computing the Limit

- The limiting variance of the distributions in a martingale must be 0, which implies:
 - θ converges to a single atom (DP and PYP)
 - All masses converge to 0 or 1 (beta process)
 - θ converges to 0 (gamma process)
- **DP, BP, and GammaP all involve draws from a gamma random variable, so we will necessarily run into the pathology described in Lemma 1!**
- See Example 3 for a martingale that can converge to an arbitrary value in $[0,1]$ (also used in Solution 2)

Solution 1: Pitman-Yor Processes

- Pitman-Yor processes have the following consistency property: if $G_1 | G_0 \sim \text{PYP}(\alpha, d_0, G_0)$, and $G_2 | G_1 \sim \text{PYP}(\alpha d_1, d_1, G_1)$, then $G_2 | G_0 \sim \text{PYP}(\alpha d_1 d_0, d_1 d_0, G_0)$.
- In general, $G_n | G_0 \sim \text{PYP}(\alpha d_1 \dots d_n, d_0 \dots d_n, G_0)$. If $G_0(\{p\}) = \varepsilon$, then $G_n(\{p\})$ is approximately

$$\left(\frac{\alpha \varepsilon}{d_0 + \alpha \varepsilon} \right)^{\frac{1}{d_0 \dots d_n}}$$

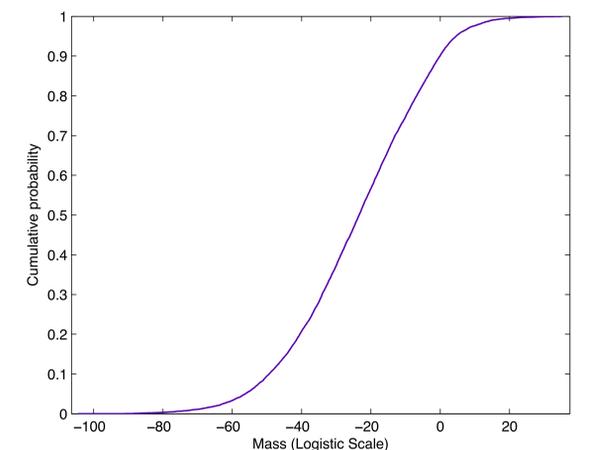


Figure 3: The cdf of the mass placed on an atom of base mass 0.1, for a draw from PYP(0.1, 0.05).

Solution 2: Adding Inertia

- Instead of $x_{n+1} \sim \text{Gamma}(c_n x_n, c_n)$, have, e.g., $d_n \sim \text{Gamma}(c_n x_n, c_n)$, and $x_{n+1} = (1 - a_n)x_n + a_n d_n$.
- Still a martingale, even for Dirichlet
- Rate of decay controlled by the sequence a_n

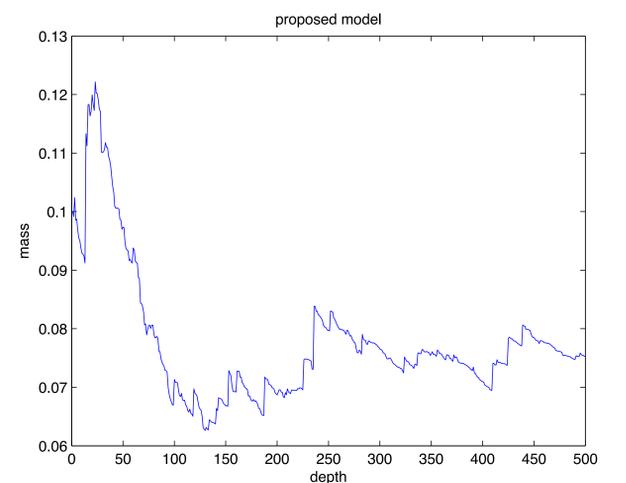


Figure 4: the mass of an atom on an inertia-added hierarchical Beta process. The sequence θ_n is generated as:

$$\begin{aligned} \alpha_{n+1} &= \alpha_n + \text{Gamma}(\alpha_n / \theta_n, 5) \\ \beta_{n+1} &= \beta_n + \text{Gamma}(\beta_n / \theta_n, 5) \\ \theta_{n+1} &= \alpha_{n+1} / (\alpha_{n+1} + \beta_{n+1}) \end{aligned}$$

Pathologies of the Gamma Distribution for Small Parameters

- For small settings of the parameters, samples from a gamma distribution can end up very close to zero.
- Lemma 1: If $y \sim \text{Gamma}(cx, c)$, and $cx \leq 1$, then

$$\mathbb{P} \left[y \leq \frac{1}{c} 2^{-\frac{1}{cx}} \right] \geq \frac{1}{2e}$$

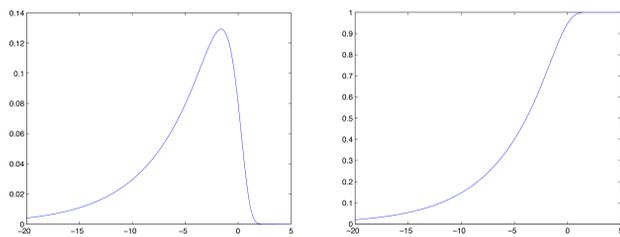


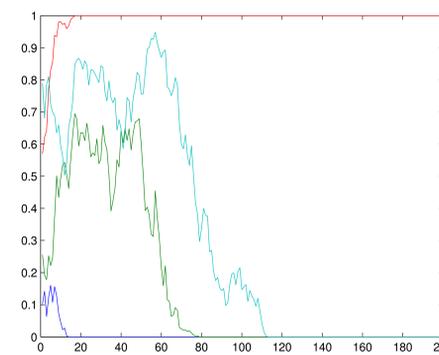
Figure 1: The pdf (left) and cdf (right) of $\log(Y)$, where $Y \sim \text{Gamma}(0.2, 1.0)$. Note the relatively large amount of probability mass placed on values as small as $\exp(-20)$.

- So, we should avoid choosing such small parameters. But **for deep hierarchies, this turns out to be unavoidable!**
- Gamma, beta, and Dirichlet sequences all decay towards 0 or 1 at a rate governed by a **tower of exponentials**: $1/e^{(e^{(e^{(e^{(\dots)})})})})}$.

Why Call This Behavior Pathological?

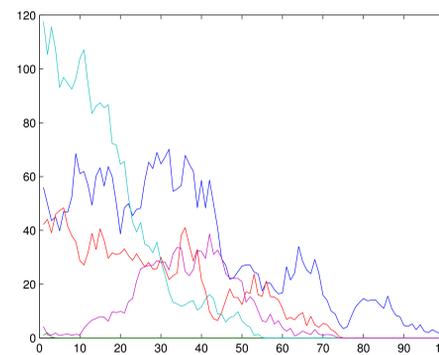
- **Practically:** if the parameters converge extremely rapidly, then posterior inference is extremely sensitive to parameter values deep in the tree, which are too small to represent accurately on a computer
 - The difference between a parameter value of 0, $10^{-1000000000}$, and 10^{-100} matters significantly to the conditional distribution of a parameter 3 levels up
- **Philosophically:** as Bayesians, we would never report confidences as high as $\exp(\exp(\dots(1)))$, so our models should not, either.

Examples of Martingale Sequences



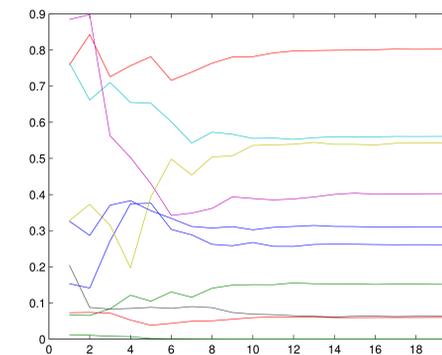
Example 1: parameters of a hierarchical Beta process.

$$\theta_{n+1} | \theta_n \sim \text{Beta}(50\theta_n, 50(1-\theta_n))$$



Example 2: parameters of a hierarchical Gamma process.

$$x_{n+1} | x_n \sim \text{Gamma}(x_n, 1)$$



Example 3: a martingale given by $\theta_n = \alpha_n / (\alpha_n + \beta_n)$, where:

$$\begin{aligned} \alpha_{n+1} | \alpha_n &\sim \alpha_n + \text{Gamma}(\alpha_n, 1), \\ \beta_{n+1} | \beta_n &\sim \beta_n + \text{Gamma}(\beta_n, 1). \end{aligned}$$

This construction can be used to rectify the problems with HBPs and HDPs.

Proving That the Decay Rate is a Tower of Exponentials

- **Theorem:** If $x_{n+1} \sim \text{Gamma}(c_n x_n, c_n)$, where $\{c_n\}$ is bounded, then $x_k \leq (\exp)^M(1)$ with probability $1 - \varepsilon$, where $k = bM$ and b depends only on ε .
 - Note: $(\exp)^M$ means exponentiation composed M times
- Proof sketch: $x_{n+1} \ll x_n$ with non-negligible probability by Lemma 1, but the martingale property together with Markov's inequality bounds the probability that x_{n+1} is ever more than a constant greater than x_n .
- **Similar convergence properties (tower of exponentials) for Beta and Dirichlet.**

Naive Solution: Mixing with Noise

- Break martingale property and take, e.g., $\theta_{n+1} \sim \text{DP}(c[(1-\varepsilon)\theta_n + \varepsilon\mu_0])$, where μ_0 is some global base measure
- Issue: with N atoms, μ_0 places mass $1/N$ on some atom, so DP has at least one parameter as small as $c\varepsilon/N$
 - Even more trouble with infinitely many atoms
- Forgets information after $1/\varepsilon$ steps

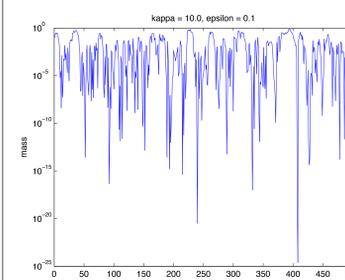


Figure 2: The mass assigned to an atom for a hierarchical Dirichlet process with noise mixed in. Here we have parameters $c = 10.0$, $\varepsilon = 0.1$, and μ_0 a uniform distribution over 10 atoms.