

Discussion of

# On the Problem of Evaluating Formal Posterior Distributions

by Morris Eaton

and

# Reference Priors Under Partial Invariance

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# Evaluating default Bayesian methodology

Model selection	Estimation
Consistency	Admissibility
Mean squared error	Frequentist coverage

- Evaluate formal Bayesian procedures by means of frequentist calculations
- Proper priors result in admissible procedures
- Complete class theorems: admissible are formal Bayes
- Strong admissibility: is it really what we want from a “non-informative” procedure?
- “boundary of admissibility” — Berger and Strawderman (96)

## Admissibility of formal Bayes procedures

- Brown (71) — recurrent diffusion on the sample space  $\mathcal{X}$
- Eaton (92) — recurrent discrete time Markov chain on the parameter space  $\Theta$
- Cast admissibility as a minimization problem using Stein (55) and Blyth (51)

Eaton (92) transition kernel of the MC:

$$K(\theta' | \theta) = \int_{\mathcal{X}} \frac{f(x | \theta') \pi^N(\theta')}{m(x)} f(x | \theta) dx$$

$$\pi^N(\theta) K(\theta' | \theta) = \pi^N(\theta') K(\theta | \theta')$$

$\therefore \pi^N(\theta)$  is the invariant measure (unique up to a multiplicative constant) for  $K$  — but  $\pi^N$  is only  $\sigma$ -finite...

## Verifiable conditions

- Brown (71)
  - conditions on  $m(x)$
- Eaton (08)
  - $\pi^N(\theta) = \int \pi(\theta | \beta) \pi^N(\beta) d\beta$
  - sufficient conditions for recurrence on moments of increments of the chain

## Invariance and invariant priors

- Typically simplify calculations
- Known general results?

# Semi-invariance

- Semi-invariance  $\sim$  partial information
- Sun and Berger (98) “Reference priors with partial information” — for  $\pi(\theta_1, \theta_2)$ 
  - know  $\pi(\theta_1)$ , want  $\pi(\theta_2 | \theta_1)$
  - know  $\pi(\theta_1 | \theta_2)$ , want  $\pi(\theta_2)$
  - know that *a priori*  $\theta_1$  and  $\theta_2$  are independent, want joint
- One can argue that since  $f(x | \theta, \xi)$  has the same invariance structure regardless of  $\theta$ ,  $\pi(\xi | \theta) = \pi^r(\xi)$
- One can argue that one wants to take advantage of invariance, so that  $\pi(\xi) = \pi^r(\xi)$
- One can even argue for independence...

## Applications to testing?

$$H_0 : X \sim \frac{1}{\sigma} g_1 \left( \frac{x - \mu}{\sigma} \mid \theta_1 \right) \text{ vs } H_1 : X \sim \frac{1}{\sigma} g_2 \left( \frac{x - \mu}{\sigma} \mid \theta_2 \right)$$

- reparameterize if necessary to guarantee that the group actions are the same under  $H_0$  and  $H_1$ , leaving  $\theta_1$  and  $\theta_2$  unaffected
- $\pi^R(\theta_1)$  and  $\pi^R(\theta_2)$  proper
- $m_1(x \mid \mu, \sigma)$  and  $m_2(x \mid \mu, \sigma)$  retain the invariance structure (Dass and Berger, 03)
- $B_{12} = f_1(T(x))/f_2(T(x))$  (Berger, Pericchi and Varshavsky, 98)