

Käenmäki example

This example provides a g -measure μ such that the ratio

$$\frac{\mu([x_1, x_2, \dots, x_m, y_1, y_2, \dots, y_m])}{\mu([x_1, x_2, \dots, x_m]) \cdot \mu([y_1, y_2, \dots, y_m])}$$

is uniformly bounded from above but not from below. A restatement says that the likelihood ratios between any conditioning of the measure and measure itself, i.e.

$$\frac{\mu([x_1, x_2, \dots, x_m] | y)}{\mu([x_1, x_2, \dots, x_m])} = \frac{\mu([x_1, x_2, \dots, x_m] | y)}{\int \mu([x_1, x_2, \dots, x_m | z]) d\mu(z)},$$

are bounded from above but not from below.

Let $X = \{0, 1\}^S$, $S = \{0, 1, 2, \dots\}$, and let $\mu \in \mathcal{M}(X)$ be the unique Hofbauer-type g -measure with

$$g(0x) = e^{-s_{W(x)}}, \quad g(1x) = 1 - e^{-s_{W(x)}}$$

where $W(x)$ is the number of initial zeros in $x \in X$ and

$$s_k = 2 - \frac{1}{1+k}.$$

Note that μ is sandwiched (in stochastic dominance ordering) between two iid Bernoulli sequences with $p = 1 - e^{-1}$ and $p = 1 - e^{-2}$.

For the Hofbauer-type measure a likelihood ratio only depends on the span of the zero-sequence, i.e., if we set $V = W(y)$ then

$$\frac{\mu([x_1, x_2, \dots, x_m] | y)}{\mu([x_1, x_2, \dots, x_m])} = \frac{\mu([0_k] | [0_V, 1])}{\mu([0]_k)}$$

where 0_k denotes a sequence of k zeros and where k is given by the condition that either $k = m$ or else

$$x_{m-k} = 1, \quad x_{m-k+1} = x_{m-k+2} = \dots = x_m = 0.$$

Hence, it is enough to consider the case $(x_1, \dots, x_m) = 0_m$.

Let $V = W(y)$ be the waiting time for the sequence y . Note that the ratio

$$\begin{aligned} \frac{\mu(0_m)}{\mu(0_m | y)} &= \int \exp \{ (s_V - s_W) + (s_{V+1} - s_{W+1}) + \dots + (s_{V+m-1} - s_{W+m-1}) \} d\mu(z) \\ &= \int \exp \left\{ \sum_{k=W+1}^{W+m} \frac{1}{k} - \sum_{k=V+1}^{V+m} \frac{1}{k} \right\} d\mu(z), \end{aligned}$$

where $W = W(z)$.

Since $\sum_{k=1}^N \frac{1}{k} = \ln N + \gamma + o(1)$ as $N \rightarrow \infty$, we can deduce that

$$\begin{aligned} \frac{\mu([0_m])}{\mu([0_m] \mid y)} &= \int \exp \left\{ \sum_{k=W+1}^{W+m} \frac{1}{k} - \sum_{k=V+1}^{V+m} \frac{1}{k} \right\} d\mu(z) \\ &\simeq \frac{V+1}{V/m+1} \cdot \int \frac{W/m+1}{W+1} d\mu(z). \end{aligned}$$

where $f \simeq g$ here means that $|f/g|$ is bounded away from 0 and ∞ as for all m .

Since $W(z)$ is sandwiched between two geometric distributions with $p = 1 - e^{-2}$ and $p = 1 - e^{-1}$, it is clear that the integral

$$I = \int \frac{W/m+1}{W+1} d\mu(z) = \frac{1}{m} + \left(1 - \frac{1}{m}\right) \cdot \mathbb{E} \left(\frac{1}{W+1} \right)$$

satisfies

$$(1 - e^{-1}) < I < 1.$$

Thus, we obtain the following expression for the likelihood ratios

$$\frac{\mu([0_m] \mid y)}{\mu([0_m])} \simeq \frac{1}{I} \cdot \frac{V/m+1}{V+1}.$$

The right hand side is less than $1/I$. Hence the likelihood ratios are bounded from above. On the other hand, if we let $V \rightarrow \infty$ then

$$\frac{\mu([0_m] \mid y)}{\mu([0_m])} \rightarrow \frac{1}{I} \cdot \Theta\left(\frac{1}{m}\right)$$

and the likelihood-ratios are thus not bounded away from zero since m is arbitrary.