

Asymptotic behaviour of the weighted Shannon differential entropy in a Bayesian problem

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Introduction

Let $U \sim \mathbb{U}[0, 1]$.

Given a realization of this RV p , consider a sequence of conditionally independent identically distributed $(\xi_i, i = 1, 2, \dots)$, where $\xi_i = 1$ with probability p and $\xi_i = 0$ with probability $1 - p$. Let x_i , each 0 or 1, be an outcome in trial i .

Denote $S_n = \xi_1 + \dots + \xi_n$ and $x = \sum_{i=1}^n x_i$.

$$\mathbb{P}(\xi_i = 1, \xi_j = 1) = \int_0^1 p^2 dp = 1/3 \text{ if } i \neq j, \text{ but}$$

$$\mathbb{P}(\xi_i = 1)\mathbb{P}(\xi_j = 1) = \left(\int_0^1 p dp\right)^2 = 1/4.$$

The probability that after n trials the exact sequence $(x_i, i = 1, \dots, n)$ will appear equals

$$\mathbb{P}(\xi_1 = x_1, \dots, \xi_n = x_n) = \int_0^1 p^x (1-p)^{n-x} dp = \frac{1}{(n+1) \binom{n}{x}}. \quad (1)$$

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This implies that the posterior probability density function (PDF) of the number of x successes after n trials is uniform

$$\mathbb{P}(S_n = x) = \frac{1}{(n+1)}, x = 0, \dots, n.$$

The posterior PDF given the information that after n trials one observes x successes takes the form

$$f^{(n)}(p|\xi_1 = x_1, \dots, \xi_n = x_n) = f^{(n)}(p|S_n = x) = (n+1) \binom{n}{x} p^x (1-p)^{n-x}, \quad (2)$$

Note that conditional distribution given in (2) is a **Beta-distribution**.

“It is known that Beta-distribution is *asymptotically normal* with its mean and variance as x and $(n-x)$ tend to infinity, but this fact is *lacking a handy reference*”

Introduction

Consider RV $Z^{(n)}$ on $[0; 1]$ with PDF (2). Note that $Z^{(n)}$ has the followings expectation:

$$\mathbb{E}_x[Z^{(n)}] = \frac{x+1}{n+2}, \quad (3)$$

and the following variance:

$$\mathbb{V}_x[Z^{(n)}] = \frac{(x+1)(n-x+1)}{(n+3)(n+2)^2}. \quad (4)$$

Shannon's differential entropy

The goal of our previous work [13] was to study the asymptotic behaviour of the differential entropy (DE) of the following RVs:

- 1 $Z_\alpha^{(n)}$ with PDF $f_\alpha^{(n)}$ given in (2) when $x = x(n) = \lfloor \alpha n \rfloor$ where $0 < \alpha < 1$ and $\lfloor a \rfloor$ is integer part of a .
- 2 $Z_\beta^{(n)}$ with PDF $f_\beta^{(n)}$ given in (2) when $x = x(n) = \lfloor n^\beta \rfloor$ where $0 < \beta < 1$
- 3 $Z_{c_1}^{(n)}$ with PDF $f_{c_1}^{(n)}$ given in (2) when $x = c_1$ and $Z_{n-c_2}^{(n)}$ with PDF $f_{n-c_2}^{(n)}$ given in (2) when $n - x(n) = c_2$ where c_1 and c_2 are some constants.

It is shown that in the **first** and **second** cases limiting distribution is *Gaussian* and the differential entropy of standardized RV converges to differential entropy of *standard Gaussian RV*.

In the **third** case the limiting distribution is *not Gaussian*, but still the asymptotic of differential entropy can be found explicitly.

Recall

Differential entropy (DE) $h(f)$ of a RV Z with the PDF f :

$$h(f) = h_{diff}(f) = - \int_{\mathbb{R}} f(z) \log(f(z)) dz \quad (5)$$

with the convention $0 \log 0 = 0$.

A linear transformation $X = b_1 Z + b_2$,

$$h(g) = h(f) + \log b_1 \quad (6)$$

where g is a PDF of RV X .

Let \bar{Z} be the standard Gaussian RV with PDF φ then the differential entropy of \bar{Z} equals:

$$h(\varphi) = \frac{1}{2} \log(2\pi e).$$

Recall the definition of the Kullback–Leibler divergence of g from f

$$\mathbb{D}(f||g) = \int_{\mathbb{R}} f(x) \log \frac{f(x)}{g(x)} dx. \quad (7)$$

Shannon's differential entropy. Case I

Theorem

Let $\tilde{Z}_\alpha^{(n)} = n^{\frac{1}{2}}(\alpha(1-\alpha))^{-\frac{1}{2}}(Z_\alpha^{(n)} - \alpha)$ be a RV with PDF $\tilde{f}_\alpha^{(n)}$. Let $\bar{Z} \sim \mathcal{N}(0, 1)$ be the standard Gaussian RV, then

(a) $\tilde{Z}_\alpha^{(n)}$ weakly converges to \bar{Z} :

$$\tilde{Z}_\alpha^{(n)} \Rightarrow \bar{Z} \text{ as } n \rightarrow \infty.$$

(b) The differential entropy of $\tilde{Z}_\alpha^{(n)}$ converges to differential entropy of \bar{Z} :

$$\lim_{n \rightarrow \infty} h(\tilde{f}_\alpha^{(n)}) = \frac{1}{2} \log(2\pi e).$$

(c) The Kullback-Leibler divergence of φ from $\tilde{f}_\alpha^{(n)}$ tends to 0 as $n \rightarrow \infty$:

$$\lim_{n \rightarrow \infty} \mathbb{D}(\tilde{f}_\alpha^{(n)} || \varphi) = 0.$$

Shannon's differential entropy. Case I

We obtained the following asymptotic of the differential entropy:

$$\lim_{n \rightarrow \infty} \left[h(f_\alpha^{(n)}) - \frac{1}{2} \log \frac{2\pi e [x(n-x)]}{n^3} \right] = 0. \quad (8)$$

Particularly,

$$\lim_{n \rightarrow \infty} \left[h(f_\alpha^{(n)}) - \frac{1}{2} \log \frac{2\pi e [\alpha(1-\alpha)]}{n} \right] = 0. \quad (9)$$

Due to (6), the differential entropy of RV $\tilde{Z}_\alpha^{(n)}$ has the form:

$$\lim_{n \rightarrow \infty} \left[h(\tilde{f}_\alpha^{(n)}) \right] = \frac{1}{2} \log (2\pi e). \quad (10)$$

Shannon's differential entropy. Case II

Theorem

Let $\tilde{Z}_\beta^{(n)} = n^{1-\beta/2}(Z_\beta^{(n)} - n^{\beta-1})$ be a RV with PDF $\tilde{f}_\beta^{(n)}$ and $\bar{Z} \sim \mathcal{N}(0, 1)$ then

(a) $\tilde{Z}_\beta^{(n)}$ weakly converges to \bar{Z} :

$$\tilde{Z}_\beta^{(n)} \Rightarrow \bar{Z} \text{ as } n \rightarrow \infty.$$

(b) The differential entropy of $\tilde{Z}_\beta^{(n)}$ converges to differential entropy of \bar{Z} :

$$\lim_{n \rightarrow \infty} h(\tilde{f}_\beta^{(n)}) = \frac{1}{2} \log(2\pi e).$$

(c) The Kullback-Leibler divergence of φ from $\tilde{f}_\beta^{(n)}$ tends to 0 as $n \rightarrow \infty$:

$$\lim_{n \rightarrow \infty} \mathbb{D}(\tilde{f}_\beta^{(n)} \parallel \varphi) = 0.$$

Shannon's differential entropy. Case III

Theorem

Let $\tilde{Z}_{c_1}^{(n)} = nZ_{c_1}^{(n)}$ be a RV with PDF $\tilde{f}_{c_1}^{(n)}$ and $\tilde{Z}_{n-c_2}^{(n)} = nZ_{n-c_2}^{(n)}$ be a RV with PDF $\tilde{f}_{n-c_2}^{(n)}$. Denote $H_k = 1 + \frac{1}{2} + \dots + \frac{1}{k}$ the partial sum of harmonic series and γ the Euler-Mascheroni constant, then

$$(a) \lim_{n \rightarrow \infty} h(\tilde{f}_{c_1}^{(n)}) = c_1 + \sum_{i=0}^{c_1-1} \log(c_1 - i) - c_1(H_{c_1} - \gamma) + 1.$$

$$(b) \lim_{n \rightarrow \infty} h(\tilde{f}_{n-c_2}^{(n)}) = c_2 + \sum_{i=0}^{c_2-1} \log(c_2 - i) - c_2(H_{c_2} - \gamma) + 1.$$

Motivation of the weighted differential entropy

Consider the following statistical experiment with **twofold goal**:

- 1 **on the initial stage** an experimenter is mainly concerned whether the coin is approximately **fair** with a high precision.
- 2 **As the size of a sample grows**, he proceeds to estimate **the true value** of the parameter anyway.

We want to *quantify the differential entropy of this experiment* taking into account its two-sided objective.

Quantitative measure of information gain of this experiment is provided by the concept of

the weighted differential entropy.

Introducing the weight function

Let $\phi^{(n)} \equiv \phi^{(n)}(\alpha, \gamma, \mathbf{p})$ be a weight function that underlines the importance of some particular value γ .

Choosing the weight function we adopt the following normalization rule:

$$\int_{\mathbb{R}} \phi^{(n)} f_{\alpha}^{(n)} d\mathbf{p} = 1. \quad (11)$$

Weighted differential entropies

The goal of this work is to study the asymptotic behaviour of weighted Shannon's (12) and Renyi's differential entropies of RV $Z^{(n)}$ with PDF $f^{(n)}$ given in (2) and particular RV $Z_\alpha^{(n)}$ with PDF $f_\alpha^{(n)}$ given in (2) with $x = \lfloor \alpha n \rfloor$ where $0 < \alpha < 1$:

$$h^\phi(f_\alpha^{(n)}) = - \int_{\mathbb{R}} \phi^{(n)} f_\alpha^{(n)} \log f_\alpha^{(n)} dp, \quad (12)$$

$$H_\nu^\phi(f_\alpha^{(n)}) = \frac{1}{1-\nu} \log \int_{\mathbb{R}} \phi^{(n)} \left(f_\alpha^{(n)}\right)^\nu dp \quad (13)$$

where $\nu \geq 0$ and $\nu \neq 1$.

The weight function $\phi^{(n)}$

The following special cases are considered:

- 1 $\phi^{(n)} \equiv 1$
- 2 $\phi^{(n)}$ depends both on n and p

In this paper we consider the weight function of the following form:

$$\phi^{(n)}(p) = \Lambda^{(n)}(\alpha, \gamma) p^{\gamma\sqrt{n}} (1-p)^{(1-\gamma)\sqrt{n}} \quad (14)$$

where $\Lambda^{(n)}(\alpha, \gamma, p)$ is found from the normalizing condition (11).

This is the model example with a **twofold goal**:

- **to emphasize a particular value** γ (for moderate n)
- **asymptotically unbiased** estimate

$$\lim_{n \rightarrow \infty} \int_0^1 p \phi^{(n)} f^{(n)} dp = \alpha.$$

The weighted Shannon differential entropy

Theorem

For the weighted Shannon differential entropy of RV $Z_\alpha^{(n)}$ with PDF $f_\alpha^{(n)}$ and weight function $\phi^{(n)}$ given in (14) the following limit exists

$$\lim_{n \rightarrow \infty} \left(h^\phi(f_\alpha^{(n)}) - \frac{1}{2} \log \left(\frac{2\pi e \alpha (1 - \alpha)}{n} \right) \right) = \frac{(\alpha - \gamma)^2}{2\alpha(1 - \alpha)}. \quad (15)$$

If the $\alpha = \gamma$ then

$$\lim_{n \rightarrow \infty} \left(h^\phi(f_\alpha^{(n)}) - h(f_\alpha^{(n)}) \right) = 0 \quad (16)$$

where $h(f_\alpha^{(n)})$ is the standard ($\phi \equiv 1$) Shannon's differential entropy.

The weighted Shannon differential entropy

The normalizing constant in the weight function (14) is found from the condition (11). We obtain that

$$\Lambda^{(n)}(\gamma) = \frac{\Gamma(x+1)\Gamma(n-x+1)\Gamma(n+2+\sqrt{n})}{\Gamma(x+\gamma\sqrt{n}+1)\Gamma(n-x+1+\sqrt{n}-\gamma\sqrt{n})\Gamma(n+2)} = \frac{\mathbb{B}(x+1, n-x+1)}{\mathbb{B}(x+\gamma\sqrt{n}+1, n-x+\sqrt{n}-\gamma\sqrt{n}+1)} \quad (17)$$

where $\Gamma(x)$ is the Gamma function and $\mathbb{B}(x, y)$ is the Beta function. We denote by $\psi^{(0)}(x) = \psi(x)$ and by $\psi^{(1)}(x)$ the digamma function and its first derivative respectively.

Recall the Stirling formula:

$$n! = \sqrt{2\pi n} \left(\frac{n}{e}\right)^n \left(1 + \frac{1}{12n} + O\left(\frac{1}{n^2}\right)\right) \text{ as } n \rightarrow \infty. \quad (18)$$

The weighted Renyi differential entropy

Theorem

Let $Z^{(n)}$ be a RV with PDF $f^{(n)}$ given in (2), $Z_\alpha^{(n)}$ be a RV with PDF $f_\alpha^{(n)}$ given in (2) with $x = \lfloor \alpha n \rfloor$, $0 < \alpha < 1$ and $H_\nu(f^{(n)})$ be the weighted Renyi differential entropy given in (13).

(a) When $\phi^{(n)} \equiv 1$ and both (x) and $(n - x)$ tend to infinity as $n \rightarrow \infty$ the following limit holds

$$\lim_{n \rightarrow \infty} \left(H_\nu(f^{(n)}) - \frac{1}{2} \log \frac{2\pi x(n-x)}{n^3} \right) = -\frac{\log(\nu)}{2(1-\nu)}, \quad (19)$$

(b) When the weight function $\phi^{(n)}$ is given in (14) the following limit for the Renyi weighted entropy of $f_\alpha^{(n)}$ holds

$$\lim_{n \rightarrow \infty} \left(H_\nu^\phi(f_\alpha^{(n)}) - \frac{1}{2} \log \frac{2\pi\alpha(1-\alpha)}{n} \right) = -\frac{\log(\nu)}{2(1-\nu)} + \frac{(\alpha - \gamma)^2}{2\alpha(1-\alpha)\nu}, \quad (20)$$

The weighted Renyi differential entropy

$$H_\nu(f^{(n)}) = \frac{1}{2} \log \left(\frac{2\pi x(n-x)}{n^3} \right) - \frac{\log(\nu)}{2(1-\nu)} + O\left(\frac{1}{n}\right). \quad (21)$$

Note that the leading terms in (21) looks like Renyi differential entropy of Gaussian RV with variance $\sigma^2 = \frac{x(n-x)}{n^3}$.

Taking the limit when $\nu \rightarrow 1$ and applying L'Hopital's rule we get that

$$H_{\nu \rightarrow 1}(f^{(n)}) = \lim_{\nu \rightarrow 1} H_\nu(f^{(n)}) = \frac{1}{2} \log \left(\frac{2e\pi x(n-x)}{n^3} \right) + O\left(\frac{1}{n}\right). \quad (22)$$

For example, when $x = \lfloor \alpha n \rfloor$, $0 < \alpha < 1$ the Renyi entropy:

$$H_{\nu \rightarrow 1}(f^{(n)}) = \frac{1}{2} \log \frac{2\pi e[\alpha(1-\alpha)]}{n} + O\left(\frac{1}{n}\right)$$

where the leading term is Shannon's differential entropy of Gaussian RV with corresponding variance.

The weighted Renyi differential entropy

Theorem

For any continuous random variable X with PDF f and for any non-negative weight function $\phi(x)$ which satisfies condition (11) and such that

$$\int_{\mathbb{R}} \phi(x) f(x)^\nu |\log(f(x))| dx < \infty,$$

the weighted Renyi differential entropy $H_\nu^\phi(f)$ is a non-increasing function of ν and

$$\frac{\partial}{\partial \nu} H_\nu^\phi(f) = -\frac{1}{(1-\nu)^2} \int_{\mathbb{R}} z(x) \log \frac{z(x)}{\phi(x) f(x)} dx \quad (23)$$

where

$$z(x) = \frac{\phi(x)(f(x))^\nu}{\int_{\mathbb{R}} \phi(x)(f(x))^\nu dx}$$

Further extension

Natural extension of this work is to derive **the weighted analogous of the Fisher Information** and the generalized version of well-known inequalities for the weighted variance

- Cramér-Rao inequality
- Bhattacharyya inequality

and for weighted Kullback distance

- Kullback inequality

Similar models of sensitive estimator appear in many fields of statistics. So, application of the weighted differential entropy approach can be adapted to a large variety of problems.

Thank you for attention!



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