



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

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Abstract: Consider a Bayesian problem of success probability estimation in a series of conditionally independent trials with binary outcomes. We study the asymptotic behaviour of the weighted differential entropy for posterior probability density function conditional on x successes after n conditionally independent trials when $n \rightarrow \infty$. Suppose that one is interested to know whether the coin is approximately fair with a high precision and for large n is interested in the true frequency. In other words, the statistical decision is particularly sensitive in small neighbourhood of the particular value $\gamma = 1/2$. For this aim the concept of weighted differential entropy introduced in [1] is used when the frequency γ is necessary to emphasize. It is shown that when x is a proportion of n after an appropriate normalization the limiting distribution is Gaussian and the standard differential entropy of standardized RV converges to differential entropy of standard Gaussian random variable. Also, we found that the weight in suggested form does not change the asymptotic form of the Shannon and Renyi differential entropies, but changes the constants.

Keywords: weighted differential entropy, Bernoulli random variable, Renyi entropy

1. Introduction

Let U be a random variable (RV) that uniformly distributed in interval $[0, 1]$. Given a realization of this RV p , consider a sequence of conditionally independent identically distributed ξ_i 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 by

$S_n = \xi_1 + \dots + \xi_n$, by $\mathbf{x} = (x_i, i = 1, \dots, n)$ and by $x = x(n) = \sum_{i=1}^n x_i$. The posterior PDF given the information that after n throws we observe x heads takes the form

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

Note that conditional distribution given in (1) is a Beta-distribution $B(x + 1, n - x + 1)$. The RV $Z^{(n)}$ with PDF (1) has the following conditional variance:

$$\mathbb{V}[Z^{(n)}|S_n = x] = \frac{(x + 1)(n - x + 1)}{(n + 3)(n + 2)^2}. \tag{2}$$

Recall the definition of a differential entropy $h(f)$ of a RV Z with the PDF f :

$$h(f) = h_{diff}(f) = - \int_{\mathbb{R}} f(z) \log(f(z)) dz \tag{3}$$

with the convention $0 \log 0 = 0$. While referring to the differential entropy of a RV Z we mean the entropy of its PDF f . Consider a linear transformation $X = b_1 Z + b_2$, then [3,7]:

$$h(g) = h(f) + \log b_1 \tag{4}$$

where g is a PDF of RV X . Let \bar{Z} be the standard Gaussian RV with PDF φ ,

$$\varphi(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}},$$

then the differential entropy of \bar{Z} equals [7]:

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

In our previous paper [8] the standard Shannon entropy of (1) was studied in three particular cases: $x = \lfloor \alpha n \rfloor$, $x \sim n^\beta$, where $0 < \alpha, \beta < 1$ and either x or $n - x$ is a constant. We had demonstrated that the limiting distributions when $n \rightarrow \infty$ in the cases 1 and 2 are Gaussian. However, the asymptotic normality does not imply automatically the limiting form of differential entropy. In general the problem of taking the limits under the sign of entropy is rather delicate and was extensively studied in literature, cf., i.e., [4,6]. In stated problem, it was proved that in the first and second cases the differential entropy is asymptotically Gaussian with corresponding variances. In the third case the limiting distribution is not Gaussian, but still the asymptotics of differential entropy can be found explicitly.

We would like to extend the theory of the Shannon differential entropy. For this reason, let us consider the following statistical experiment with twofold goal: on the initial stage an experimenter mainly concerns whether the coin is approximately fair (i.e. $p \approx \frac{1}{2}$) with a high precision. As the size of a sample grows, he proceeds to estimate the true value of the parameter anyway. We would like to quantify the differential entropy of this experiment taking into account its two sided objective. It seems that quantitative measure of information gain of this experiment is provided by the concept of weighted differential entropy [1,2].

Let $\phi^{(n)} \equiv \phi^{(n)}(\alpha, \gamma, p)$ be a weight function that underlines the importance of some particular value γ ($\gamma = 1/2$ in the problem stated above). The goal of this work is to study the asymptotic behaviour of

the weighted Shannon (5) differential entropy [3,8] of RV $Z^{(n)}$ with PDF $f^{(n)}$ given in (1) and particular RV $Z_\alpha^{(n)}$ with PDF $f_\alpha^{(n)}$ given in (1) with $x = \lfloor \alpha n \rfloor$ where $0 < \alpha < 1$:

$$h^\phi(f_\alpha^{(n)}) = - \int_{\mathbb{R}} \phi^{(n)} f_\alpha^{(n)} \log f_\alpha^{(n)} dp \tag{5}$$

and generalisation of the weighted Shannon entropy, the weighted Renyi differential entropy

$$H_\nu^\phi(f_\alpha^{(n)}) = \frac{1}{1-\nu} \log \int_{\mathbb{R}} \phi^{(n)} (f_\alpha^{(n)})^\nu dp. \tag{6}$$

When the weight function is uniform ($\phi \equiv 1$) we will omit the superscript ϕ . Moreover, we would like to compare asymptotics of the weighted differential entropy and the standard differential entropy. Thus, the following special cases are considered:

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

We assume that $\phi^{(n)}(x) \geq 0$ for all x . Choosing the weight function we adopt the following normalization rule:

$$\int_{\mathbb{R}} \phi^{(n)} f_\alpha^{(n)} dp = 1. \tag{7}$$

It can be easily checked that if weight function $\phi^{(n)}$ satisfies (7) then the Renyi weighted entropy (6) tend to Shannon’s weighted entropy as $\nu \rightarrow 1$. In this paper we consider the weight function of the following form:

$$\phi^{(n)}(p) = \Lambda^{(n)}(\alpha, \gamma) p^{\gamma\sqrt{n}} (1-p)^{(1-\gamma)\sqrt{n}} \tag{8}$$

where $\Lambda^{(n)}(\alpha, \gamma, p)$ is found from the normalizing condition (7) and is given explicitly in (16). This weight function is selected as a model example with a twofold goal to emphasize a particular value γ for moderate n , while preserving the estimate to be asymptotically unbiased

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

2. Main results

Theorem 1. 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 $n \rightarrow \infty$:

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

Theorem 2. For the weighted Shannon differential entropy of RV $Z_\alpha^{(n)}$ with PDF $f_\alpha^{(n)}$ and weight function $\phi^{(n)}$ given in (8) the following limit exists

$$\lim_{n \rightarrow \infty} \left(h^\phi(f_\alpha^{(n)}) - \frac{1}{2} \log \left(\frac{2\pi e \alpha(1-\alpha)}{n} \right) \right) = \frac{(\alpha - \gamma)^2}{2\alpha(1-\alpha)}. \tag{9}$$

If $\alpha = \gamma$ then

$$\lim_{n \rightarrow \infty} (h^\phi(f_\alpha^{(n)}) - h(f_\alpha^{(n)})) = 0 \tag{10}$$

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

Theorem 3. Let $Z^{(n)}$ be a RV with PDF $f^{(n)}$ given in (1), $Z_\alpha^{(n)}$ be a RV with PDF $f_\alpha^{(n)}$ given in (1) with $x = \lfloor \alpha n \rfloor$, $0 < \alpha < 1$ and $H_\nu(f^{(n)})$ be the weighted Renyi differential entropy given in (6).

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

$$\lim_{n \rightarrow \infty} \left(H_\nu(f^{(n)}) - \frac{1}{2} \log \frac{2\pi x(n-x)}{n^3} \right) = -\frac{\log(\nu)}{2(1-\nu)}, \tag{11}$$

and for any fixed n

$$\lim_{\nu \rightarrow 1} (H_\nu(f^{(n)}) - h(f^{(n)})) = 0. \tag{12}$$

(b) When the weight function $\phi^{(n)}$ is given in (8) the following limit for the Renyi weighted entropy of $f_\alpha^{(n)}$ holds

$$\lim_{n \rightarrow \infty} \left(H_\nu^\phi(f_\alpha^{(n)}) - \frac{1}{2} \log \frac{2\pi \alpha(1-\alpha)}{n} \right) = -\frac{\log(\nu)}{2(1-\nu)} + \frac{(\alpha - \gamma)^2}{2\alpha(1-\alpha)\nu}, \tag{13}$$

and for any fixed n

$$\lim_{\nu \rightarrow 1} (H_\nu^\phi(f_\alpha^{(n)}) - h^\phi(f_\alpha^{(n)})) = 0. \tag{14}$$

Proposition 1. For any continuous random variable X with PDF f and for any non-negative weight function $\phi(x)$ which satisfies condition (7) 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

$$\frac{\partial}{\partial \nu} H_\nu^\phi(f) = -\frac{1}{(1-\nu)^2} \int_{\mathbb{R}} z(x) \log \frac{z(x)}{\phi(x)f(x)} dx \tag{15}$$

where

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

3. Proofs

The normalizing constant in the weight function (8) is found from the condition (7). We obtain that

$$\Lambda^{(n)}(\gamma) = \frac{\Gamma(x+1)\Gamma(n-x+1)\Gamma(n+2+\sqrt{n})}{\Gamma(x+\gamma\sqrt{n}+1)\Gamma(n-x+1+\sqrt{n}-\gamma\sqrt{n})\Gamma(n+2)} = \frac{\mathbb{B}(x+1, n-x+1)}{\mathbb{B}(x+\gamma\sqrt{n}+1, n-x+\sqrt{n}-\gamma\sqrt{n}+1)} \tag{16}$$

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

$$\psi^{(j)}(x) = \frac{d^{j+1}}{dx^{j+1}} \log \Gamma(x). \tag{17}$$

In further calculations we will need the asymptotics of digamma functions in two particular cases $j = 0$ and $j = 1$ only

$$\begin{aligned} \psi(x) &= \log(x) - \frac{1}{2x} + O\left(\frac{1}{x^2}\right) \text{ as } x \rightarrow \infty, \\ \psi^{(1)}(x) &= \frac{1}{x} + \frac{1}{2x^2} + O\left(\frac{1}{x^3}\right) \text{ as } x \rightarrow \infty. \end{aligned}$$

Recall also the Stirling formula [5]:

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

3.1. Theorem 1

Proof. The proof can be found in [8]. \square

We will use the following result of Theorem 1. The asymptotics of the differential entropy of unstandartized RV $Z_\alpha^{(n)}$ with PDF $f_\alpha^{(n)}$:

$$\lim_{n \rightarrow \infty} \left[h(f_\alpha^{(n)}) - \frac{1}{2} \log \frac{2\pi e[\alpha(1-\alpha)]}{n} \right] = 0. \tag{19}$$

3.2. Theorem 2 The Shannon differential entropy of PDF $f_\alpha^{(n)}$ given in (1) with the weight function $\phi^{(n)}$ given in (8) takes the form:

$$h^\phi(f_\alpha^{(n)}) = \log \left[(n+1) \binom{n}{x} \right] + x \int_0^1 \log(p) \phi^{(n)} f_\alpha^{(n)} dp + (n-x) \int_0^1 \log(1-p) \phi^{(n)} f_\alpha^{(n)} dp$$

The integrals can be computed explicitly [5] (4.253.1):

$$\int_0^1 x^{\mu-1} (1-x^r)^{\nu-1} \log(x) dx = \frac{1}{r^2} \mathbb{B}\left(\frac{\mu}{r}, \nu\right) \left(\psi\left(\frac{\mu}{r}\right) - \psi\left(\frac{\mu}{r} + \nu\right) \right).$$

Applying this formula, we get

$$\int_0^1 \log(p)\phi^{(n)} f_\alpha^{(n)} dp = \psi(x + z + 1) - \psi(n + \sqrt{n} + 2)$$

and

$$\int_0^1 \log(1 - p)\phi^{(n)} f_\alpha^{(n)} dp = \psi(n - x + \sqrt{n} - z + 1) - \psi(n + \sqrt{n} + 2)$$

where $z = \gamma\sqrt{n}$.

Applying Stirling’s formula (18) and using the asymptotics for digamma function we have that

$$h^\phi(f_\alpha^{(n)}) = \frac{1}{2} \log \frac{2\pi e[\alpha(1 - \alpha)]}{n} + \frac{(\alpha - \gamma)^2}{2\alpha(1 - \alpha)} + O\left(\frac{1}{\sqrt{n}}\right). \tag{20}$$

The leading term in (20) is the Shannon differential entropy of Gaussian RV with weight function $\phi^{(n)} \equiv 1$. Moreover, note that leading term of the asymptotics for the weighted differential entropy exceeds that for the classical differential entropy studied in [8]. The only difference is constant which tends to zero as $\gamma \rightarrow \alpha$.

3.3. Theorem 3 (a) In this case $\phi^{(n)}(p) \equiv 1$ the Renyi entropy has the form

$$(1 - \nu)H_\nu(f^{(n)}) = \nu \log \left[(n + 1) \binom{n}{x} \right] + \log \left[\int_0^1 p^{\nu x} (1 - p)^{\nu(n-x)} dp \right].$$

Consider the integral:

$$\int_0^1 p^{\nu x} (1 - p)^{\nu(n-x)} dp = \mathbb{B}(\nu x + 1, \nu(n - x) + 1) = \frac{\Gamma(\nu x + 1)\Gamma(\nu(n - x) + 1)}{\Gamma(\nu n + 2)}.$$

Applying Stirling formula, we obtain that

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

So, we have that

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

Note that the leading terms in (22) looks like Renyi differential entropy of Gaussian RV with variance $\frac{x(n-x)}{n^3}$.

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

$$H_{\nu \rightarrow 1}(f^{(n)}) = \lim_{\nu \rightarrow 1} H_\nu(f^{(n)}) = \frac{1}{2} \log \left(\frac{2e\pi x(n - x)}{n^3} \right) + O\left(\frac{1}{n}\right). \tag{23}$$

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.

(b) When $\phi^{(n)}$ is given in (8) and $x = \lfloor \alpha n \rfloor$, the weighted Renyi differential entropy of PDF $f_\alpha^{(n)}$ takes the following form

$$H_\nu^\phi(f_\alpha^{(n)}) = \frac{1}{1-\nu} \log \int_0^1 \phi^{(n)} (f_\alpha^{(n)})^\nu dp,$$

$$\int_0^1 \phi^{(n)} (f_\alpha^{(n)})^\nu dp \equiv U_1 U_2 U_3 \text{ where}$$

$$U_1 = \frac{\Gamma(\nu x + \gamma\sqrt{n} + 1)\Gamma(\nu(n-x) + (1-\gamma)\sqrt{n} + 1)}{\Gamma(\nu n + \sqrt{n} + 2)},$$

$$U_2 = \left(\frac{\Gamma(n+2)}{\Gamma(x+1)\Gamma(n-x+1)} \right)^{\nu-1},$$

$$U_3 = \frac{\Gamma(n + \sqrt{n} + 2)}{\Gamma(x+z+1)\Gamma(n-x+\sqrt{n}-z+1)}$$

where $z = \gamma\sqrt{n}$ as before.

Applying Stirling’s formula for each term and taking all parts together, we obtain that

$$H_\nu^\phi(f_\alpha^{(n)}) = \frac{1}{2} \log \frac{2\pi\alpha(1-\alpha)}{n} - \frac{\log(\nu)}{2(1-\nu)} + \frac{(\alpha-\gamma)^2}{2\alpha(1-\alpha)\nu} + O\left(\frac{1}{\sqrt{n}}\right). \tag{24}$$

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

$$H_{\nu \rightarrow 1}^\phi(f_\alpha^{(n)}) = \lim_{\nu \rightarrow 1} H_\nu^\phi(f_\alpha^{(n)}) = \frac{1}{2} \log \frac{2\pi e[\alpha(1-\alpha)]}{n} + \frac{(\alpha-\gamma)^2}{2\alpha(1-\alpha)} + O\left(\frac{1}{\sqrt{n}}\right). \tag{25}$$

So, for any fixed n the weighted Renyi differential entropy tends to Shannon’s weighted differential entropy as $\nu \rightarrow 1$.

3.4. Proposition 1 We need to show that

$$\frac{\partial}{\partial \nu} H_\nu^\phi(f) \leq 0,$$

$$\frac{\partial}{\partial \nu} H_\nu^\phi(f) = \frac{\log \int_{\mathbb{R}} \phi(x)(f(x))^\nu dx}{(1-\nu)^2} + \frac{\int_{\mathbb{R}} \phi(x)(f(x))^\nu \log(f(x)) dx}{(1-\nu) \int_{\mathbb{R}} \phi(x)(f(x))^\nu dx} = I_1 + I_2. \tag{26}$$

Denote

$$z(x) = \frac{\phi(x)(f(x))^\nu}{\int_{\mathbb{R}} \phi(x)(f(x))^\nu dx}. \tag{27}$$

Note that $z(x) \geq 0$ for any x and

$$\int_{\mathbb{R}} z(x) dx = 1.$$

Denote $Q = \log \int_{\mathbb{R}} \phi(x)(f(x))^\nu dx$. Using the substitution (27)

$$Q = \log(\phi(x)) + \nu \log(f(x)) - \log(z(x)) \tag{28}$$

we obtain

$$I_2 = \frac{1}{1-\nu} \int_{\mathbb{R}} z(x) \log(f(x)) dx,$$

$$I_1 + I_2 = \frac{1}{(1-\nu)^2} \left(\log \int_{\mathbb{R}} \phi(x)(f(x))^\nu dx + (1-\nu) \int_{\mathbb{R}} z(x) \log(f(x)) dx \right)$$

By substitution $\log(f(x))$ using (28) we obtain that

$$-\frac{\partial}{\partial \nu} H_\nu^\phi(f) = \frac{1}{(1-\nu)^2} \int_{\mathbb{R}} z(x) \log \left(\frac{z(x)}{\phi(x)f(x)} \right) dx = \frac{1}{(1-\nu)^2} \mathbb{D}_{KL}(z||\phi f). \quad (29)$$

Here $\mathbb{D}_{KL}(z||\phi f)$ is the Kullback-Leibler divergence between z and ϕf which is always non-negative [3,7]. Due to conditions $\phi(x)f(x) \geq 0$ and (7), $\phi(x)f(x)$ is itself a PDF:

$$\int_{\mathbb{R}} \phi(x)f(x) dx = 1$$

4. Conclusion The behaviour of the weighted and standard differential entropies was studied in this paper. We had shown that in the binary outcomes trial the difference of the weighted differential entropy of (1) with the weight function (8) and the standard differential entropy of (1) tends to some constant as $n \rightarrow \infty$. This constant depend on the distance between the point of special interest γ and true parameter α . The same conclusion holds of the weighted Renyi differential entropy. The problem of sensitive estimation is quite common. Thus, it seems that both these facts can find a variety of applications. Moreover, all results can be straightforwardly generalized for larger class of the weight function and the work in this direction should be continued.

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Conflicts of Interest

The authors declare no conflict of interest.

References

1. M. Belis, S. Guiasu, A Quantitative and qualitative measure of information in cybernetic systems, IEEE Trans. Inf. Th. **1968**, 14, 593-594
2. A. Clim, Weighted entropy with application, Analele Universitatii Bucuresti Matematica **2008**, Anul LVII, 223-231
3. T.M. Cover, J.M. Thomas, Elements of Information Theory, NY: Basic Books **2006**
4. R.L. Dobrushin, Passing to the limit under the sign of the information and entropy, Theory Prob.Appl., **1960**, 29-37
5. I.S. Gradshteyn, I.M. Ryzhik, Table of Integrals, Series, and Product, Elsevier **2007**
6. M. Kelbert, Yu. Suhov, Continuity of mutual entropy in the large signal-to-noise ratio limit, Stochastic Analysis **2010**, Berlin: Springer, 281-299
7. M. Kelbert, Yu. Suhov, Information Theory and Coding by Example Cambridge: Cambridge University Press **2013**

8. M. Kelbert, P. Mozgunov, Shannon's differential entropy asymptotic analysis in a Bayesian problem, *Mathematical Communications* **2015**, Vol 20, No 2
9. M. Kelbert, P. Mozgunov, Asymptotic behaviour of weighted differential entropies in a Bayesian problem, arXiv: 1504.01612 **2015**
10. Yu. Suhov, S. Yasaei Sekeh, M. Kelbert, Entropy-power inequality for weighted entropy, arXiv:1502.02188 **2015**

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