

Generalized Informations and Bayesian errors*

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Abstract

Key words: Power informations, Rényi informations, Bayesian errors

1. Introduction

We consider decisions about a random variable $X \sim (\{1, 2, \dots, n\}, P_X)$ on the basis of a random observation $Y \sim (\mathcal{Y}, \mathcal{B}, P_Y)$. By $P_{X|y}$ we denote the conditional probability distribution of X given $Y = y$ for $y \in \mathcal{Y}$. Let

$$e(X) = e(P_X) = 1 - \max_{1 \leq i \leq n} P_X(i) \quad (1.1)$$

be the error of the a priori Bayesian decision about X . We are interested in the error

$$e(X|Y) = \int_{\mathcal{Y}} e(P_{X|y}) dP_Y(y) \quad (1.2)$$

of the a posteriori Bayesian decision about X based on the observation Y .

We shall estimate the Bayesian decision errors $e(X|Y)$ by means of the conditional entropies

$$H_\alpha(X|Y) = \int_{\mathcal{Y}} H_\alpha(P_{X|y}) dP_Y(y) \quad (1.3)$$

of order $\alpha > 0$ where the unconditional (a priori) entropies

$$H_\alpha(P_X) \equiv H_\alpha(X), \quad \alpha > 0.$$

* Supported by grants BMF 2003-04820 and MŠMT 1M0572

are of two possible types, namely the *power entropies*

$$H_\alpha^I(P_X) \equiv H_\alpha^I(X) = \frac{1}{1-\alpha} \left(\sum_{i=1}^n P_X(i)^\alpha - 1 \right), \quad \alpha \neq 1 \quad (1.4)$$

or the *Rényi entropies*

$$H_\alpha^{II}(P_X) \equiv H_\alpha^{II}(X) = \frac{1}{1-\alpha} \ln \sum_{i=1}^n P_X(i)^\alpha, \quad \alpha \neq 1 \quad (1.5)$$

with common limit

$$H_1^I(X) \equiv H_1^{II}(X) = H_1(X) = - \sum_{i=1}^n P_X(i) \ln P_X(i) \quad (1.6)$$

known as the *Shannon entropy*. In (1.3) the particular power conditional entropies will be derived by $H^I(X|Y)$ and the conditional Rényi entropies by $H^{II}(X|Y)$. As it is well known, the differences

$$I_\alpha^I(X, Y) = H_\alpha^I(X) - H_\alpha^I(X|Y) \quad \text{and} \quad I_\alpha^{II}(X, Y) = H_\alpha^{II}(X) - H_\alpha^{II}(X|Y)$$

are generalized informations, namely the power informations and Rényi informations of orders $\alpha > 0$, see Morales et al. (1996). For the particular order these generalized informations reduce to the classical Shannon information

$$I(X, Y) = I_1^I(X, Y) = I_1^{II}(X, Y).$$

Thus in fact, our paper studies relations between the Bayes errors $e(X|Y)$ (Bayes risks for the $0 - 1$ loss functions) and the generalized informations of two types, the power informations $I_\alpha^I(X, Y)$ of orders $\alpha > 0$ and the Rényi informations $I_\alpha^{II}(X, Y)$ of orders $\alpha > 0$.

The Bayesian errors $e(X)$ and $e(X|Y)$ take on values in the interval

$$0 \leq e(X), e(X|Y) \leq e_n \quad (1.7)$$

where here and in the sequel

$$e_n = \frac{n-1}{n}, \quad \text{for } n = 1, 2, \dots \quad (1.8)$$

By Theorem 2 in Morales et al. (1996),

$$e(X) = e \quad \text{implies} \quad \underline{H}_\alpha(e) \leq H_\alpha(X) \leq \overline{H}_\alpha(e) \quad (1.9)$$

where if $\alpha \neq 1$ then the (estimable) bounds are

$$\overline{H}_\alpha^I(e) = \frac{(n-1)^{1-\alpha} e^\alpha + (1-e)^\alpha - 1}{1-\alpha}, \quad (1.10)$$

$$H_\alpha^I(e) = \sum_{k=1}^{n-1} \frac{(1-k(1-e))^\alpha + k(1-e)^\alpha - 1}{1-\alpha} I(e_k < e \leq e_{k+1}) \quad (1.11)$$

and

$$\overline{H}_\alpha^{II}(e) = \frac{1}{\alpha} \ln \left((n-1)^{1-\alpha} e^\alpha + (1-e)^\alpha \right), \quad (1.12)$$

$$\underline{H}_\alpha^{II}(e) = \frac{1}{1-\alpha} \sum_{k=1}^{n-1} \ln ((1-k(1-e))^\alpha + k(1-e)^\alpha) I(e_k < e \leq e_{k+1}) \quad (1.13)$$

For $\alpha = 1$ the bounds are common limits of (1.10), (1.11) and (1.12), (1.13) for $\alpha \rightarrow 1$, namely

$$\overline{H}_1(e) = h(e) + e \ln(n-1), \quad (1.14)$$

$$\underline{H}_1(e) = - \sum_{k=1}^{n-1} \{ [1-k(1-e)] \ln[1-k(1-e)] + k(1-e) \ln(1-e) \} I(e_k < e \leq e_{k+1}) \quad (1.15)$$

where

$$h(e) = -e \ln e - (1-e) \ln(1-e). \quad (1.16)$$

Further, by Theorem 4 in Morales et al. (1996)

$$e(X|Y) = e \quad \text{implies} \quad \underline{\underline{H}}_\alpha(e) \leq H_\alpha(X|Y) \leq \overline{H}_\alpha(e) \quad (1.17)$$

where the (attainable) upper bounds $\overline{H}_\alpha(e)$ are given by (1.10), (1.12) and (1.14) above. The (attainable) lower bounds are for $\alpha \neq 1$.

$$\underline{\underline{H}}_\alpha(e) = \frac{1}{1-\alpha} \sum_{k=1}^{n-1} (a_k + k(k+1)b_k(e - e_k)) I(e_k < e \leq e_{k+1}) \quad (1.18)$$

where

$$a_k^I = k^{1-\alpha} - 1, \quad b_k^I = (k+1)^{1-\alpha} - k^{1-\alpha} \quad (1.19)$$

and

$$a_k^{II} = (1-\alpha) \ln k, \quad b_k^{II} = (1-\alpha) \ln \frac{k+1}{k} \quad (1.20)$$

while the bound

$$\underline{\underline{H}}_1(e) = \sum_{k=1}^{n-1} \left(\ln k + \ln k(k+1)(e - e_k) \ln \frac{k+1}{k} \right) I(e_k < e \leq e_{k+1}) \quad (1.21)$$

is the common limit of $\underline{\underline{H}}_\alpha^I(e)$ and $\underline{\underline{H}}_\alpha^{II}(e)$ for $\alpha \rightarrow 1$.

Problem: To investigate how tight are attainable bounds for $e(X|Y)$ obtained from $H_\alpha^I(X|Y)$ and $H_\alpha^{II}(X|Y)$ as functions of $\alpha > 0$.

We shall reduce this problem to the evaluation of integrals

$$\tau_\alpha = \frac{1}{H_{\alpha,n}} \int_0^{e_n} \left(\overline{H}_\alpha(e) - \underline{\underline{H}}_\alpha(e) \right) de \quad (1.22)$$

where

$$H_{\alpha,n} = H_{\alpha,max}(X), \quad \text{for } n\text{-valued } X. \quad (1.23)$$

In the next two sections are presented the results of numerical evaluation of these integrals for power informations (τ_α^I) and Rényi informations (τ_α^{II}) for selected values of n (rows) and α (columns). The corresponding analytical formulas and conclusions are under the current research. However, an immediate preliminary conclusion which can be even at this stage drawn from the Tables 2.1 and 3.1 is that the quadratic information (power information of the order $\alpha = 2$) is much more closely related to the Bayes decision error than the classical Shannon information.

2. Power informations and Bayesian errors

| n | 0.125 | 0.25 | 0.5 | 1 | 1.5 | 2 | 3 | 4 |
|-------|--------|--------|--------|--------|--------|--------|--------|--------|
| 2 | 0.2163 | 0.1900 | 0.1524 | 0.1107 | 0.0914 | 0.0833 | 0.0833 | 0.0929 |
| 3 | 0.3619 | 0.3183 | 0.2538 | 0.2816 | 0.1345 | 0.1111 | 0.0938 | 0.0958 |
| 4 | 0.4465 | 0.3927 | 0.3121 | 0.3460 | 0.1574 | 0.1250 | 0.0981 | 0.0969 |
| 5 | 0.5034 | 0.4426 | 0.3511 | 0.3804 | 0.1720 | 0.1333 | 0.1004 | 0.0974 |
| 6 | 0.5448 | 0.4790 | 0.3795 | 0.4019 | 0.1821 | 0.1389 | 0.1017 | 0.0976 |
| 7 | 0.5767 | 0.5070 | 0.4014 | 0.4167 | 0.1897 | 0.1429 | 0.1025 | 0.0977 |
| 8 | 0.6021 | 0.5294 | 0.4190 | 0.4275 | 0.1956 | 0.1458 | 0.1031 | 0.0978 |
| 9 | 0.6230 | 0.5479 | 0.4335 | 0.4357 | 0.2004 | 0.1481 | 0.1035 | 0.0978 |
| 10 | 0.6404 | 0.5634 | 0.4457 | 0.4422 | 0.2043 | 0.1500 | 0.1038 | 0.0979 |
| 20 | 0.7319 | 0.6452 | 0.5113 | 0.4705 | 0.2242 | 0.1583 | 0.1047 | 0.0979 |
| 30 | 0.7699 | 0.6800 | 0.5400 | 0.4796 | 0.2322 | 0.1611 | 0.1049 | 0.0979 |
| 40 | 0.7916 | 0.7001 | 0.5571 | 0.4842 | 0.2367 | 0.1625 | 0.1050 | 0.0979 |
| 50 | 0.8058 | 0.7134 | 0.5688 | 0.4869 | 0.2397 | 0.1633 | 0.1050 | 0.0979 |
| 100 | 3.3185 | 2.5337 | 1.5735 | 0.9678 | 0.4113 | 0.2500 | 0.1051 | 0.0479 |
| 200 | 3.6874 | 2.7323 | 1.6311 | 0.9918 | 0.4118 | 0.2500 | 0.1051 | 0.0479 |
| 300 | 0.8668 | 0.7737 | 0.6268 | 0.4963 | 0.2536 | 0.1661 | 0.1051 | 0.0979 |
| 400 | 0.8711 | 0.7784 | 0.6322 | 0.4968 | 0.2547 | 0.1663 | 0.1051 | 0.0979 |
| 500 | 0.8740 | 0.7815 | 0.6358 | 0.4971 | 0.2555 | 0.1663 | 0.1051 | 0.0979 |
| 600 | 0.8759 | 0.7837 | 0.6385 | 0.4974 | 0.2561 | 0.1664 | 0.1051 | 0.0979 |
| 700 | 0.8774 | 0.7854 | 0.6406 | 0.4975 | 0.2566 | 0.1664 | 0.1051 | 0.0979 |
| 800 | 0.8785 | 0.7867 | 0.6423 | 0.4976 | 0.2569 | 0.1665 | 0.1051 | 0.0979 |
| 900 | 0.8794 | 0.7877 | 0.6437 | 0.4977 | 0.2572 | 0.1665 | 0.1051 | 0.0979 |
| 1000 | 0.8802 | 0.7886 | 0.6449 | 0.4978 | 0.2575 | 0.1665 | 0.1051 | 0.0979 |
| 1500 | 0.8826 | 0.7914 | 0.6489 | 0.4981 | 0.2583 | 0.1666 | 0.1051 | 0.0979 |
| 2000 | 0.8839 | 0.7930 | 0.6513 | 0.4982 | 0.2588 | 0.1666 | 0.1051 | 0.0979 |
| 2500 | 0.8847 | 0.7940 | 0.6529 | 0.4983 | 0.2592 | 0.1666 | 0.1051 | 0.0979 |
| 3000 | 0.8853 | 0.7948 | 0.6541 | 0.4984 | 0.2594 | 0.1666 | 0.1051 | 0.0979 |
| 3500 | 0.8857 | 0.7953 | 0.6550 | 0.4984 | 0.2596 | 0.1666 | 0.1051 | 0.0979 |
| 4000 | 0.8860 | 0.7957 | 0.6558 | 0.4985 | 0.2598 | 0.1666 | 0.1051 | 0.0979 |
| 4500 | 0.8863 | 0.7961 | 0.6564 | 0.4985 | 0.2599 | 0.1666 | 0.1051 | 0.0979 |
| 5000 | 0.8865 | 0.7964 | 0.6569 | 0.4985 | 0.2600 | 0.1666 | 0.1051 | 0.0979 |
| 5500 | 0.8867 | 0.7966 | 0.6574 | 0.4986 | 0.2601 | 0.1666 | 0.1051 | 0.0979 |
| 6000 | 0.8868 | 0.7968 | 0.6578 | 0.4986 | 0.2602 | 0.1666 | 0.1051 | 0.0979 |
| 6500 | 0.8869 | 0.7970 | 0.6581 | 0.4986 | 0.2603 | 0.1666 | 0.1051 | 0.0979 |
| 7000 | 0.8871 | 0.7971 | 0.6584 | 0.4986 | 0.2603 | 0.1666 | 0.1051 | 0.0979 |
| 7500 | 0.8872 | 0.7973 | 0.6587 | 0.4986 | 0.2604 | 0.1666 | 0.1051 | 0.0979 |
| 8000 | 0.8873 | 0.7974 | 0.6590 | 0.4987 | 0.2604 | 0.1666 | 0.1051 | 0.0979 |
| 8500 | 0.8873 | 0.7975 | 0.6592 | 0.4987 | 0.2605 | 0.1666 | 0.1051 | 0.0979 |
| 9000 | 0.8874 | 0.7976 | 0.6594 | 0.4987 | 0.2605 | 0.1666 | 0.1051 | 0.0979 |
| 9500 | 0.8875 | 0.7977 | 0.6596 | 0.4987 | 0.2606 | 0.1666 | 0.1051 | 0.0979 |
| 10000 | 0.8875 | 0.7978 | 0.6598 | 0.4987 | 0.2606 | 0.1667 | 0.1051 | 0.0979 |

Table ???.1. Average inaccuracies $\tau_{\alpha,n}^I$ for selected α and n .

3. Rényi informations and Bayesian errors

| n | 0.125 | 0.25 | 0.5 | 1 | 1.5 | 2 | 3 | 4 |
|-------|--------|--------|--------|--------|--------|--------|--------|--------|
| 2 | 0.2235 | 0.2001 | 0.1617 | 0.1107 | 0.0798 | 0.0596 | 0.0352 | 0.0212 |
| 3 | 0.3524 | 0.3208 | 0.2659 | 0.2816 | 0.1433 | 0.1134 | 0.0782 | 0.0585 |
| 4 | 0.4320 | 0.3963 | 0.3319 | 0.3460 | 0.1826 | 0.1466 | 0.1049 | 0.0821 |
| 5 | 0.4863 | 0.4483 | 0.3776 | 0.3804 | 0.2087 | 0.1682 | 0.1221 | 0.0974 |
| 6 | 0.5260 | 0.4867 | 0.4114 | 0.4019 | 0.2270 | 0.1830 | 0.1337 | 0.1077 |
| 7 | 0.5566 | 0.5163 | 0.4376 | 0.4167 | 0.2404 | 0.1937 | 0.1419 | 0.1149 |
| 8 | 0.5809 | 0.5401 | 0.4587 | 0.4275 | 0.2507 | 0.2017 | 0.1479 | 0.1202 |
| 9 | 0.6008 | 0.5596 | 0.4761 | 0.4357 | 0.2587 | 0.2077 | 0.1523 | 0.1241 |
| 10 | 0.6175 | 0.5760 | 0.4908 | 0.4422 | 0.2651 | 0.2124 | 0.1557 | 0.1270 |
| 20 | 0.7047 | 0.6633 | 0.5698 | 0.4705 | 0.2923 | 0.2297 | 0.1662 | 0.1360 |
| 30 | 0.7416 | 0.7012 | 0.6051 | 0.4796 | 0.2992 | 0.2316 | 0.1656 | 0.1353 |
| 40 | 0.7631 | 0.7236 | 0.6266 | 0.4842 | 0.3012 | 0.2305 | 0.1634 | 0.1332 |
| 50 | 0.7776 | 0.7390 | 0.6416 | 0.4869 | 0.3015 | 0.2286 | 0.1608 | 0.1310 |
| 100 | 0.8134 | 0.7777 | 0.6810 | 0.9678 | 0.2975 | 0.2189 | 0.1506 | 0.1220 |
| 200 | 0.8393 | 0.8066 | 0.7126 | 0.9918 | 0.2888 | 0.2061 | 0.1388 | 0.1119 |
| 300 | 0.8513 | 0.8202 | 0.7284 | 0.4963 | 0.2826 | 0.1982 | 0.1320 | 0.1061 |
| 400 | 0.8588 | 0.8288 | 0.7387 | 0.4968 | 0.2778 | 0.1925 | 0.1273 | 0.1022 |
| 500 | 0.8640 | 0.8349 | 0.7462 | 0.4971 | 0.2740 | 0.1881 | 0.1237 | 0.0992 |
| 600 | 0.8680 | 0.8396 | 0.7521 | 0.4974 | 0.2708 | 0.1846 | 0.1209 | 0.0969 |
| 700 | 0.8712 | 0.8433 | 0.7568 | 0.4975 | 0.2681 | 0.1817 | 0.1186 | 0.0949 |
| 800 | 0.8739 | 0.8464 | 0.7608 | 0.4976 | 0.2658 | 0.1792 | 0.1166 | 0.0933 |
| 900 | 0.8761 | 0.8491 | 0.7643 | 0.4977 | 0.2637 | 0.1770 | 0.1149 | 0.0919 |
| 1000 | 0.8780 | 0.8514 | 0.7673 | 0.4978 | 0.2618 | 0.1750 | 0.1134 | 0.0907 |
| 1500 | 0.8849 | 0.8595 | 0.7782 | 0.4981 | 0.2546 | 0.1677 | 0.1079 | 0.0862 |
| 2000 | 0.8893 | 0.8648 | 0.7854 | 0.4982 | 0.2496 | 0.1628 | 0.1042 | 0.0832 |
| 2500 | 0.8925 | 0.8686 | 0.7908 | 0.4983 | 0.2457 | 0.1591 | 0.1015 | 0.0810 |
| 3000 | 0.8950 | 0.8716 | 0.7950 | 0.4984 | 0.2425 | 0.1561 | 0.0994 | 0.0792 |
| 3500 | 0.8970 | 0.8740 | 0.7984 | 0.4984 | 0.2399 | 0.1537 | 0.0977 | 0.0778 |
| 4000 | 0.8986 | 0.8760 | 0.8013 | 0.4985 | 0.2376 | 0.1516 | 0.0962 | 0.0766 |
| 4500 | 0.9001 | 0.8778 | 0.8038 | 0.4985 | 0.2356 | 0.1498 | 0.0949 | 0.0756 |
| 5000 | 0.9013 | 0.8793 | 0.8060 | 0.4985 | 0.2338 | 0.1483 | 0.0938 | 0.0747 |
| 5500 | 0.9024 | 0.8806 | 0.8079 | 0.4986 | 0.2322 | 0.1469 | 0.0929 | 0.0739 |
| 6000 | 0.9034 | 0.8818 | 0.8096 | 0.4986 | 0.2308 | 0.1456 | 0.0920 | 0.0732 |
| 6500 | 0.9043 | 0.8828 | 0.8112 | 0.4986 | 0.2294 | 0.1444 | 0.0912 | 0.0726 |
| 7000 | 0.9051 | 0.8838 | 0.8126 | 0.4986 | 0.2282 | 0.1434 | 0.0905 | 0.0720 |
| 8000 | 0.9065 | 0.8855 | 0.8152 | 0.4986 | 0.2260 | 0.1415 | 0.0892 | 0.0709 |
| 8500 | 0.9071 | 0.8863 | 0.8163 | 0.4987 | 0.2251 | 0.1407 | 0.0886 | 0.0705 |
| 9000 | 0.9077 | 0.8870 | 0.8174 | 0.4987 | 0.2241 | 0.1399 | 0.0881 | 0.0700 |
| 9500 | 0.9082 | 0.8877 | 0.8184 | 0.4987 | 0.2233 | 0.1392 | 0.0876 | 0.0696 |
| 10000 | 0.9087 | 0.8883 | 0.8193 | 0.4987 | 0.2225 | 0.1385 | 0.0871 | 0.0693 |

Table 3.1. Average inaccuracies $\tau_{\alpha,n}^{II}$ for selected α and n .

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