Definition and some Properties of Information Entropy

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Summary. In this article we mainly define the information entropy [3], [11] and prove some its basic properties. First, we discuss some properties on four kinds of transformation functions between vector and matrix. The transformation functions are LineVec2Mx, ColVec2Mx, Vec2DiagMx and Mx2FinS. Mx2FinS is a horizontal concatenation operator for a given matrix, treating rows of the given matrix as finite sequences, yielding a new finite sequence by horizontally joining each row of the given matrix in order to index. Then we define each concept of information entropy for a probability sequence and two kinds of probability matrices, joint and conditional, that are defined in article [25]. Further, we discuss some properties of information entropy including Shannon's lemma, maximum property, additivity and super-additivity properties.

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The papers [21], [23], [1], [20], [24], [6], [14], [8], [4], [22], [17], [7], [9], [2], [5], [15], [16], [12], [10], [13], [18], [25], and [19] provide the terminology and notation for this paper.

1. Preliminaries

For simplicity, we use the following convention: D denotes a non empty set, *i*, *j*, *k*, *l* denote elements of \mathbb{N} , *n* denotes a natural number, *a*, *b*, *c*, *r*, *r*₁, *r*₂ denote real numbers, *p*, *q* denote finite sequences of elements of \mathbb{R} , and M_1 , M_2 denote matrices over \mathbb{R} .

Next we state several propositions:

(1) If $k \neq 0$ and i < l and $l \leq j$ and $k \mid l$, then $i \div k < j \div k$.

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- (2) If r > 0, then $(\log_{-}(e))(r) \le r 1$ and r = 1 iff $(\log_{-}(e))(r) = r 1$ and $r \ne 1$ iff $(\log_{-}(e))(r) < r 1$.
- (3) If r > 0, then $\log_e r \le r 1$ and r = 1 iff $\log_e r = r 1$ and $r \ne 1$ iff $\log_e r < r 1$.
- (4) If a > 1 and b > 1, then $\log_a b > 0$.
- (5) If a > 0 and $a \neq 1$ and b > 0, then $-\log_a b = \log_a(\frac{1}{b})$.
- (6) If a > 0 and $a \neq 1$ and $b \ge 0$ and $c \ge 0$, then $b \cdot c \cdot \log_a(b \cdot c) = b \cdot c \cdot \log_a b + b \cdot c \cdot \log_a c$.
- (7) Let q, q_1, q_2 be finite sequences of elements of \mathbb{R} . Suppose len $q_1 = \text{len } q_1$ and len $q_1 = \text{len } q_2$ and for every k such that $k \in \text{dom } q_1$ holds $q(k) = q_1(k) + q_2(k)$. Then $\sum q = \sum q_1 + \sum q_2$.
- (8) Let q, q_1, q_2 be finite sequences of elements of \mathbb{R} . Suppose len $q_1 = \text{len } q_1$ and len $q_1 = \text{len } q_2$ and for every k such that $k \in \text{dom } q_1$ holds $q(k) = q_1(k) - q_2(k)$. Then $\sum q = \sum q_1 - \sum q_2$.
- (9) Suppose len $p \ge 1$. Then there exists q such that len q = len p and q(1) = p(1) and for every k such that $0 \ne k$ and k < len p holds q(k+1) = q(k) + p(k+1) and $\sum p = q(\text{len } p)$.

Let us consider p. Let us observe that p is non-negative if and only if:

(Def. 1) For every *i* such that $i \in \text{dom } p$ holds $p(i) \ge 0$.

Let us note that there exists a finite sequence of elements of \mathbb{R} which is non-negative.

The following proposition is true

(10) If p is non-negative and $r \ge 0$, then $r \cdot p$ is non-negative.

Let us consider p, k. We say that p has only one value in k if and only if:

(Def. 2) $k \in \text{dom } p$ and for every i such that $i \in \text{dom } p$ and $i \neq k$ holds p(i) = 0. Next we state four propositions:

- (11) If p has only one value in k and $i \neq k$, then p(i) = 0.
- (12) If len p = len q and p has only one value in k, then $p \bullet q$ has only one value in k and $(p \bullet q)(k) = p(k) \cdot q(k)$.
- (13) If p has only one value in k, then $\sum p = p(k)$.
- (14) If p is non-negative, then for every k such that $k \in \text{dom } p$ and $p(k) = \sum p$ holds p has only one value in k.

Let us observe that every finite sequence of elements of \mathbb{R} which is finite probability distribution is also non empty and non-negative.

One can prove the following propositions:

(15) Let p be finite probability distribution finite sequence of elements of \mathbb{R} and given k such that $k \in \text{dom } p$ and p(k) = 1. Then p has only one value in k.

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(16) Let *i* be a non empty natural number. Then $i \mapsto \frac{1}{i}$ is finite probability distribution finite sequence of elements of \mathbb{R} .

One can check that every matrix over \mathbb{R} which is summable-to-1 is also non empty yielding and every matrix over \mathbb{R} which is joint probability is also non empty yielding.

The following propositions are true:

- (17) For every matrix M over \mathbb{R} such that $M = \emptyset$ holds SumAll M = 0.
- (18) For every matrix M over D and for every i such that $i \in \text{dom } M$ holds dom M(i) = Seg width M.
- (19) M_1 is nonnegative iff for every *i* such that $i \in \text{dom } M_1$ holds $\text{Line}(M_1, i)$ is non-negative.
- 2. Properties of Transformations between Vector and Matrix

Next we state four propositions:

- (20) For every j such that $j \in \text{dom } p$ holds $(\text{LineVec2Mx } p)_{\Box,j} = \langle p(j) \rangle$.
- (21) Let p be a non empty finite sequence of elements of \mathbb{R} , q be a finite sequence of elements of \mathbb{R} , and M be a matrix over \mathbb{R} . Then $M = \text{ColVec2Mx } p \cdot \text{LineVec2Mx } q$ if and only if the following conditions are satisfied:
 - (i) $\operatorname{len} M = \operatorname{len} p$,
 - (ii) width $M = \operatorname{len} q$, and
- (iii) for all i, j such that $\langle i, j \rangle \in$ the indices of M holds $M_{i,j} = p(i) \cdot q(j)$.
- (22) Let p be a non empty finite sequence of elements of \mathbb{R} , q be a finite sequence of elements of \mathbb{R} , and M be a matrix over \mathbb{R} . Then $M = \text{ColVec2Mx } p \cdot \text{LineVec2Mx } q$ if and only if the following conditions are satisfied:
 - (i) $\operatorname{len} M = \operatorname{len} p$,
 - (ii) width $M = \operatorname{len} q$, and
- (iii) for every *i* such that $i \in \text{dom } M$ holds $\text{Line}(M, i) = p(i) \cdot q$.
- (23) Let p, q be finite probability distribution finite sequences of elements of \mathbb{R} . Then ColVec2Mx $p \cdot \text{LineVec2Mx } q$ is joint probability.

Let us consider n and let M_1 be a matrix over \mathbb{R} of dimension n. We say that M_1 is diagonal if and only if:

(Def. 3) For all i, j such that $\langle i, j \rangle \in$ the indices of M_1 and $(M_1)_{i,j} \neq 0$ holds i = j.

Let us consider n. Observe that there exists a matrix over \mathbb{R} of dimension n which is diagonal.

The following proposition is true

(24) Let M_1 be a matrix over \mathbb{R} of dimension n. Then M_1 is diagonal if and only if for every i such that $i \in \text{dom } M_1$ holds $\text{Line}(M_1, i)$ has only one value in i.

Let us consider p. The functor Vec2DiagMx p yielding a diagonal matrix over \mathbb{R} of dimension len p is defined as follows:

(Def. 4) For every j such that $j \in \text{dom } p$ holds $(\text{Vec2DiagMx } p)_{j,j} = p(j)$.

One can prove the following propositions:

- (25) $M_1 = \text{Vec2DiagMx } p$ iff $\text{len } M_1 = \text{len } p$ and width $M_1 = \text{len } p$ and for every i such that $i \in \text{dom } M_1$ holds $\text{Line}(M_1, i)$ has only one value in iand $\text{Line}(M_1, i)(i) = p(i)$.
- (26) Suppose len $p = \text{len } M_1$. Then $M_2 = \text{Vec2DiagMx } p \cdot M_1$ if and only if the following conditions are satisfied:
 - (i) $\operatorname{len} M_2 = \operatorname{len} p$,
 - (ii) width M_2 = width M_1 , and
- (iii) for all i, j such that $\langle i, j \rangle \in$ the indices of M_2 holds $(M_2)_{i,j} = p(i) \cdot (M_1)_{i,j}$.
- (27) If len $p = \text{len } M_1$, then $M_2 = \text{Vec2DiagMx } p \cdot M_1$ iff len $M_2 = \text{len } p$ and width $M_2 = \text{width } M_1$ and for every i such that $i \in \text{dom } M_2$ holds $\text{Line}(M_2, i) = p(i) \cdot \text{Line}(M_1, i).$
- (28) Let p be finite probability distribution finite sequence of elements of \mathbb{R} and M be a non empty yielding conditional probability matrix over \mathbb{R} . If len p = len M, then Vec2DiagMx $p \cdot M$ is joint probability.
- (29) Let M be a matrix over D and p be a finite sequence of elements of D^* . Suppose len p = len M and p(1) = M(1) and for every k such that $k \ge 1$ and k < len M holds $p(k+1) = p(k) \cap M(k+1)$. Let given k. If $k \in \text{dom } p$, then len $p(k) = k \cdot \text{width } M$.
- (30) Let M be a matrix over D and p be a finite sequence of elements of D^* . Suppose len p = len M and p(1) = M(1) and for every k such that $k \ge 1$ and k < len M holds $p(k+1) = p(k) \cap M(k+1)$. Let given i, j. If $i \in \text{dom } p$ and $j \in \text{dom } p$ and $i \le j$, then $\text{dom } p(i) \subseteq \text{dom } p(j)$.
- (31) Let M be a matrix over D and p be a finite sequence of elements of D^* . Suppose len p = len M and p(1) = M(1) and for every k such that $k \ge 1$ and k < len M holds $p(k+1) = p(k) \cap M(k+1)$. Then len p(1) = width Mand for every j such that $\langle 1, j \rangle \in \text{the indices of } M$ holds $j \in \text{dom } p(1)$ and $p(1)(j) = M_{1,j}$.
- (32) Let M be a matrix over D and p be a finite sequence of elements of D^* . Suppose len p = len M and p(1) = M(1) and for every k such that $k \ge 1$ and k < len M holds $p(k+1) = p(k) \cap M(k+1)$. Let given j. If $j \ge 1$ and j < len p, then for every l such that $l \in \text{dom } p(j)$ holds p(j)(l) = p(j+1)(l).
- (33) Let M be a matrix over D and p be a finite sequence of elements of D^* .

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Suppose len p = len M and p(1) = M(1) and for every k such that $k \ge 1$ and k < len M holds $p(k+1) = p(k) \cap M(k+1)$. Let given i, j. Suppose $i \in \text{dom } p$ and $j \in \text{dom } p$ and $i \le j$. Let given l. If $l \in \text{dom } p(i)$, then p(i)(l) = p(j)(l).

- (34) Let M be a matrix over D and p be a finite sequence of elements of D^* . Suppose len p = len M and p(1) = M(1) and for every k such that $k \ge 1$ and k < len M holds $p(k+1) = p(k) \cap M(k+1)$. Let given j. Suppose $j \ge 1$ and j < len p. Let given l. If $l \in \text{Seg width } M$, then $j \cdot \text{width } M + l \in \text{dom } p(j+1)$ and $p(j+1)(j \cdot \text{width } M + l) = M(j+1)(l)$.
- (35) Let M be a matrix over D and p be a finite sequence of elements of D^* . Suppose len p = len M and p(1) = M(1) and for every k such that $k \ge 1$ and k < len M holds $p(k+1) = p(k) \cap M(k+1)$. Let given i, j. Suppose $\langle i, j \rangle \in$ the indices of M. Then $(i-1) \cdot \text{width } M + j \in \text{dom } p(i)$ and $M_{i,j} = p(i)((i-1) \cdot \text{width } M + j)$.
- (36) Let M be a matrix over D and p be a finite sequence of elements of D^* . Suppose len p = len M and p(1) = M(1) and for every k such that $k \ge 1$ and k < len M holds $p(k+1) = p(k) \cap M(k+1)$. Let given i, j. Suppose $\langle i, j \rangle \in \text{the indices of } M$. Then $(i-1) \cdot \text{width } M + j \in \text{dom } p(\text{len } M)$ and $M_{i,j} = p(\text{len } M)((i-1) \cdot \text{width } M + j)$.
- (37) Let M be a matrix over \mathbb{R} and p be a finite sequence of elements of \mathbb{R}^* . Suppose len p = len M and p(1) = M(1) and for every k such that $k \ge 1$ and k < len M holds $p(k+1) = p(k) \cap M(k+1)$. Let given k. If $k \ge 1$ and k < len M, then $\sum p(k+1) = \sum p(k) + \sum M(k+1)$.
- (38) Let M be a matrix over \mathbb{R} and p be a finite sequence of elements of \mathbb{R}^* . Suppose len p = len M and p(1) = M(1) and for every k such that $k \ge 1$ and k < len M holds $p(k + 1) = p(k) \cap M(k + 1)$. Then SumAll $M = \sum p(\text{len } M)$.

Let D be a non empty set and let M be a matrix over D. The functor Mx2FinS M yields a finite sequence of elements of D and is defined by:

- (Def. 5)(i) Mx2FinS $M = \emptyset$ if len M = 0,
 - (ii) there exists a finite sequence p of elements of D^* such that Mx2FinS $M = p(\operatorname{len} M)$ and $\operatorname{len} p = \operatorname{len} M$ and p(1) = M(1) and for every k such that $k \ge 1$ and $k < \operatorname{len} M$ holds $p(k+1) = p(k) \cap M(k+1)$, otherwise.

We now state several propositions:

- (39) For every matrix M over D holds len Mx2FinS $M = \text{len } M \cdot \text{width } M$.
- (40) Let M be a matrix over D and given i, j. If $\langle i, j \rangle \in$ the indices of M, then (i-1)·width $M+j \in$ dom Mx2FinS M and $M_{i,j} = (Mx2FinS M)((i-1) \cdot \text{width } M+j)$.
- (41) Let M be a matrix over D and given k, l. Suppose $k \in \text{dom} Mx2 \text{FinS} M$

and l = k - 1. Then $\langle (l \div \text{width } M) + 1, (l \mod \text{width } M) + 1 \rangle \in \text{the indices}$ of M and $(\text{Mx2FinS } M)(k) = M_{(l \div \text{width } M) + 1, (l \mod \text{width } M) + 1}$.

- (42) SumAll $M_1 = \sum Mx 2 FinS M_1$.
- (43) M_1 is nonnegative iff Mx2FinS M_1 is non-negative.
- (44) M_1 is joint probability iff Mx2FinS M_1 is finite probability distribution.
- (45) Let p, q be finite probability distribution finite sequences of elements of \mathbb{R} . Then Mx2FinS(ColVec2Mx $p \cdot \text{LineVec2Mx } q)$ is finite probability distribution.
- (46) Let p be finite probability distribution finite sequence of elements of \mathbb{R} and M be a non empty yielding conditional probability matrix over \mathbb{R} . If len p = len M, then Mx2FinS(Vec2DiagMx $p \cdot M$) is finite probability distribution.

3. INFORMATION ENTROPY

Let us consider a, p. Let us assume that a > 0 and $a \neq 1$ and p is non-negative. The functor $\overrightarrow{\log_a p}$ yields a finite sequence of elements of \mathbb{R} and is defined by:

(Def. 6) $\operatorname{len} \overrightarrow{\log}_a p = \operatorname{len} p$ and for every k such that $k \in \operatorname{dom} \overrightarrow{\log}_a p$ holds if p(k) > 0, then $(\overrightarrow{\log}_a p)(k) = \operatorname{log}_a p(k)$ and if p(k) = 0, then $(\overrightarrow{\log}_a p)(k) = 0$.

Let us consider p. The functor $\operatorname{id} \operatorname{log} p$ yields a finite sequence of elements of \mathbb{R} and is defined by:

(Def. 7) $\overrightarrow{\operatorname{id}} \log p = p \bullet \overrightarrow{\log}_2 p.$

The following propositions are true:

- (47) Let p be a non-negative finite sequence of elements of \mathbb{R} and given q. Then $q = \overrightarrow{\mathrm{id} \log p}$ if and only if the following conditions are satisfied:
 - (i) $\operatorname{len} q = \operatorname{len} p$, and
- (ii) for every k such that $k \in \text{dom } q$ holds $q(k) = p(k) \cdot \log_2 p(k)$.
- (48) Let p be a non-negative finite sequence of elements of \mathbb{R} and given k such that $k \in \text{dom } p$. Then
 - (i) if p(k) = 0, then $(id \log p)(k) = 0$, and
- (ii) if p(k) > 0, then $(\operatorname{id} \log p)(k) = p(k) \cdot \log_2 p(k)$.
- (49) Let p be a non-negative finite sequence of elements of \mathbb{R} and given q. Then $q = -\overrightarrow{id \log p}$ if and only if the following conditions are satisfied:
 - (i) $\operatorname{len} q = \operatorname{len} p$, and
 - (ii) for every k such that $k \in \text{dom } q$ holds $q(k) = p(k) \cdot \log_2(\frac{1}{p(k)})$.
- (50) Let p be a non-negative finite sequence of elements of \mathbb{R} . Suppose $r_1 \ge 0$ and $r_2 \ge 0$. Let given i. If $i \in \text{dom } p$ and $p(i) = r_1 \cdot r_2$, then $(\overrightarrow{\text{id log }} p)(i) = r_1 \cdot r_2 \cdot \log_2 r_1 + r_1 \cdot r_2 \cdot \log_2 r_2$.

- (51) For every non-negative finite sequence p of elements of \mathbb{R} such that $r \ge 0$ holds $\overrightarrow{id \log r} \cdot p = r \cdot \log_2 r \cdot p + r \cdot (p \bullet \overrightarrow{\log_2} p)$.
- (52) Let p be a non empty finite probability distribution finite sequence of elements of \mathbb{R} and given k. If $k \in \text{dom } p$, then $(\overrightarrow{\text{id log }} p)(k) \leq 0$.

Let us consider M_1 . Let us assume that M_1 is nonnegative. The functor $\overrightarrow{\mathrm{id} \log M_1}$ yields a matrix over \mathbb{R} and is defined as follows:

(Def. 8) len $\overrightarrow{\operatorname{id}} \operatorname{og} M_1 = \operatorname{len} M_1$ and width $\overrightarrow{\operatorname{id}} \operatorname{log} M_1 = \operatorname{width} M_1$ and for every k such that $k \in \operatorname{dom} \operatorname{id} \operatorname{log} M_1$ holds $(\operatorname{id} \operatorname{log} M_1)(k) = \operatorname{Line}(M_1, k) \bullet \overrightarrow{\operatorname{log}}_2 \operatorname{Line}(M_1, k).$

The following two propositions are true:

- (53) For every nonnegative matrix M over \mathbb{R} and for every k such that $k \in \operatorname{dom} M$ holds $\operatorname{Line}(\operatorname{id} \log M, k) = \operatorname{id} \log \operatorname{Line}(M, k)$.
- (54) Let M be a nonnegative matrix over \mathbb{R} and M_3 be a matrix over \mathbb{R} . Then $M_3 = \overrightarrow{\mathrm{id} \log M}$ if and only if the following conditions are satisfied:
 - (i) $\operatorname{len} M_3 = \operatorname{len} M$,
 - (ii) width M_3 = width M, and
- (iii) for all i, j such that $\langle i, j \rangle \in$ the indices of M_3 holds $(M_3)_{i,j} = M_{i,j} \cdot \log_2(M_{i,j})$.

Let p be a finite sequence of elements of \mathbb{R} . The functor Entropy p yields a real number and is defined by:

(Def. 9) Entropy $p = -\sum \overrightarrow{\operatorname{id} \log p}$.

We now state several propositions:

- (55) For every non empty finite probability distribution finite sequence p of elements of \mathbb{R} holds Entropy $p \ge 0$.
- (56) Let p be a non empty finite probability distribution finite sequence of elements of \mathbb{R} . If there exists k such that $k \in \text{dom } p$ and p(k) = 1, then Entropy p = 0.
- (57) Let p, q be non empty finite probability distribution finite sequences of elements of \mathbb{R} and p_1, q_3 be finite sequences of elements of \mathbb{R} . Suppose that
 - (i) $\operatorname{len} p = \operatorname{len} q$,
 - (ii) $\operatorname{len} p_1 = \operatorname{len} p$,
- (iii) $\operatorname{len} q_3 = \operatorname{len} q$, and
- (iv) for every k such that $k \in \text{dom } p$ holds p(k) > 0 and q(k) > 0 and $p_1(k) = -p(k) \cdot \log_2 p(k)$ and $q_3(k) = -p(k) \cdot \log_2 q(k)$.

Then

- (v) $\sum p_1 \leq \sum q_3$,
- (vi) for every k such that $k \in \text{dom } p$ holds p(k) = q(k) iff $\sum p_1 = \sum q_3$, and
- (vii) there exists k such that $k \in \text{dom } p$ and $p(k) \neq q(k)$ iff $\sum p_1 < \sum q_3$.

- (58) Let p be a non empty finite probability distribution finite sequence of elements of \mathbb{R} . Suppose that for every k such that $k \in \text{dom } p$ holds p(k) > 0. Then
 - (i) Entropy $p \le \log_2 \ln p$,
 - (ii) for every k such that $k \in \text{dom } p$ holds $p(k) = \frac{1}{\text{len } p}$ iff Entropy $p = \log_2 \text{len } p$, and
- (iii) there exists k such that $k \in \operatorname{dom} p$ and $p(k) \neq \frac{1}{\operatorname{len} p}$ iff Entropy $p < \log_2 \operatorname{len} p$.
- (59) For every nonnegative matrix M over \mathbb{R} holds Mx2FinS $\operatorname{id} \log M = \operatorname{id} \log \operatorname{Mx2FinS} M$.
- (60) Let p, q be finite probability distribution finite sequences of elements of \mathbb{R} and M be a matrix over \mathbb{R} . If $M = \text{ColVec2Mx } p \cdot \text{LineVec2Mx } q$, then SumAll $\overrightarrow{\text{id}} \log M = \sum \overrightarrow{\text{id}} \log p + \sum \overrightarrow{\text{id}} \log q$.

Let us consider M_1 . The entropy of joint probability of M_1 yields a real number and is defined as follows:

(Def. 10) The entropy of joint probability of $M_1 = \text{Entropy Mx2FinS } M_1$.

Next we state the proposition

(61) Let p, q be finite probability distribution finite sequences of elements of \mathbb{R} . Then the entropy of joint probability of ColVec2Mx $p \cdot \text{LineVec2Mx } q =$ Entropy p + Entropy q.

Let us consider M_1 . The entropy of conditional probability of M_1 yields a finite sequence of elements of \mathbb{R} and is defined by the conditions (Def. 11).

- (Def. 11)(i) len (the entropy of conditional probability of M_1) = len M_1 , and
 - (ii) for every k such that $k \in \text{dom}$ (the entropy of conditional probability of M_1) holds (the entropy of conditional probability of M_1)(k) =Entropy Line (M_1, k) .

One can prove the following propositions:

- (62) Let M be a non empty yielding conditional probability matrix over \mathbb{R} and p be a finite sequence of elements of \mathbb{R} . Then p = the entropy of conditional probability of M if and only if len p = len M and for every k such that $k \in \text{dom } p$ holds $p(k) = -\sum (\overrightarrow{\text{id log }} M)(k)$.
- (63) Let M be a non empty yielding conditional probability matrix over \mathbb{R} . Then the entropy of conditional probability of M = -LineSum id log M.
- (64) Let p be finite probability distribution finite sequence of elements of \mathbb{R} and M be a non empty yielding conditional probability matrix over \mathbb{R} . Suppose len p = len M. Let M_3 be a matrix over \mathbb{R} . If $M_3 = \text{Vec2DiagMx } p \cdot M$, then SumAll $\overrightarrow{id} \log M_3 = \sum \overrightarrow{id} \log p + \sum (p \bullet \text{LineSum } \overrightarrow{id} \log M)$.
- (65) Let p be finite probability distribution finite sequence of elements of \mathbb{R} and M be a non empty yielding conditional probability matrix over

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R. Suppose len p = len M. Then the entropy of joint probability of Vec2DiagMx $p \cdot M = \text{Entropy } p + \sum (p \bullet \text{the entropy of conditional probability of } M)$.

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