

Regularities Unseen, Randomness Observed: Addenda

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We make points in [1] that used to be simple simple again and correct a sign error. The first addendum concerns the mutual entropy form of excess entropy, showing additional steps to go from Eq. (A23) to Eq. (A24). The second clarifies and corrects a nonstandard use of information gain in Prop. 1. The third corrects a typographical error in the published version.

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I. PROPOSITION 8 OF [1]

Proposition 8: $\mathbf{E} = I[\vec{S}; \overleftarrow{S}]$.

Proof: We rewrite the definition so that we can use the finite- L forms of various entropies:

$$I[\vec{S}; \overleftarrow{S}] \equiv \lim_{L \rightarrow \infty} I[\vec{S}^{\rightarrow L}; \overleftarrow{S}^{\leftarrow L}]. \quad (1)$$

We begin with the definition of mutual information, Eq. (8) of [1], which expresses I as the difference between two entropies:

$$I[\vec{S}^{\rightarrow L}; \overleftarrow{S}^{\leftarrow L}] = H[\vec{S}^{\rightarrow L}] - H[\vec{S}^{\rightarrow L} | \overleftarrow{S}^{\leftarrow L}]. \quad (2)$$

Recall that $H[\vec{S}^{\rightarrow L}] \equiv H(L)$.

Using the conditional entropy chain rule [2] we have

$$H[\vec{S}^{\rightarrow L} | \overleftarrow{S}^{\leftarrow L}] = H[S_0, S_1, \dots, S_{L-1} | S_{-L}, \dots, S_{-1}] \quad (3)$$

$$= \sum_{i=0}^{L-1} H[S_i | S_{-L} S_{-L+1} \dots S_{i-1}]. \quad (4)$$

Putting these together we have

$$\begin{aligned} I[\vec{S}; \overleftarrow{S}] &= \lim_{L \rightarrow \infty} \left[H(L) - \sum_{i=0}^{L-1} H[S_i | S_{-L} S_{-L+1} \dots S_{i-1}] \right] \\ &= \lim_{L \rightarrow \infty} \left[H(L) - \sum_{i=0}^{L-1} H(S_L | S^{L+i}) \right]. \end{aligned} \quad (5)$$

Add and subtract $L \times h_\mu$ and rearrange:

$$\begin{aligned} I[\vec{S}; \overleftarrow{S}] &= \lim_{L \rightarrow \infty} \left[H(L) - Lh_\mu + Lh_\mu - \sum_{i=0}^{L-1} H(S_L | S^{L+i}) \right] \\ &= \lim_{L \rightarrow \infty} [H(L) - Lh_\mu] \\ &\quad + \lim_{L \rightarrow \infty} \left[Lh_\mu - \sum_{i=0}^{L-1} H(S_L | S^{L+i}) \right] \end{aligned} \quad (6)$$

Focusing on the second limit, we see that it vanishes:

$$\begin{aligned} &\lim_{L \rightarrow \infty} \sum_{i=0}^{L-1} [h_\mu - H(S_L | S^{L+i})] \\ &= \sum_{i=0}^{\infty} \lim_{L \rightarrow \infty} [h_\mu - H(S_L | S^{L+i})], \end{aligned} \quad (7)$$

since, by definition, $\lim_{L \rightarrow \infty} H(S_L | S^{L+i}) = h_\mu$ for any fixed i .

In short,

$$I[\vec{S}; \overleftarrow{S}] = \lim_{L \rightarrow \infty} [H(L) - Lh_\mu]. \quad (8)$$

The LHS of which is \mathbf{E} by Prop. 7 of [1]. \square

II. PROPOSITION 1 OF [1]

Proposition 1 introduces a nonstandard form of information gain that compares, in effect, a joint distribution to one of its marginals. Unlike the usual information gain that is non-negative, this form is non-positive.

We have the following corrections (adding minus signs) to make.

1. Eq. (24) of that proposition should read:

$$\Delta H(L) = -\mathcal{D}(\Pr(s^L) || \Pr(s^{L-1})) \quad (9)$$

This change should also be made to the restatement of Prop. 1 in Appendix A1.

2. Following Prop. 1 in the main text replace the two sentences:

When this is the case, We then sum ... sequences.

with

We appeal to the consistency conditions for word distributions: $\Pr(s^{L-1}) = \sum_{s_{L-1}} \Pr(s^L)$. This means, in particular, that (i) $\Pr(s^L) > 0 \Rightarrow \Pr(s^{L-1}) > 0$ and (ii) $\Pr(s^{L-1}) = 0 \Rightarrow \Pr(s^L) = 0$. This addresses the boundary case for the information gain when the second distribution assigns zero probability to events of positive probability under the first distribution. (The information gain diverges.) That is, this cannot occur. This and realizing that $\Pr(s^{L-1}) \leq \sum_{s_{L-1}} \Pr(s^L)$ means that this form of the information gain is nonpositive.

3. The first sentence of the immediately following paragraph should read:

Note that, due to this, it follows from Prop. 1 that $\Delta H(L) \equiv H(L) - H(L-1) \geq 0$, as remarked earlier.

4. Eqs (A3) and (A4) have the incorrect sign. For example, Eq. (A4) should read $= -H(L) + H(L-1)$.

III. EXCESS ENTROPY AS MUTUAL INFORMATION OF [1]

Eq. (54) of [1] is missing an ellipsis and should read:

$$\mathbf{E} = \lim_{L \rightarrow \infty} I[S_0 S_1 \dots S_{L-1}; S_L S_{L+1} \dots S_{2L-1}] \quad (10)$$

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