

# DIFFERENTIAL ENTROPY\*

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## Abstract

In this module we consider differential entropy.

Consider the entropy of **continuous** random variables. Whereas the (normal) entropy is the entropy of a **discrete** random variable, the differential entropy is the entropy of a continuous random variable.

## 1 Differential Entropy

### Definition 1: Differential entropy

The differential entropy  $h(X)$  of a continuous random variable  $X$  with a pdf  $f(x)$  is defined as

$$h(X) = - \int_{-\infty}^{\infty} f(x) \log f(x) dx \quad (1)$$

Usually the logarithm is taken to be base 2, so that the unit of the differential entropy is bits/symbol. Note that in the discrete case,  $h(X)$  depends only on the pdf of  $X$ . Finally, we note that the differential entropy is the expected value of  $-\log f(x)$ , i.e.,

$$h(X) = -E(\log f(x)) \quad (2)$$

Now, consider calculating the differential entropy of some random variables.

### Example 1

Consider a uniformly distributed random variable  $X$  from  $c$  to  $c + \Delta$ . Then its density is  $\frac{1}{\Delta}$  from  $c$  to  $c + \Delta$ , and zero otherwise.

We can then find its differential entropy as follows,

$$\begin{aligned} h(X) &= - \int_c^{c+\Delta} \frac{1}{\Delta} \log \frac{1}{\Delta} dx \\ &= \log \Delta \end{aligned} \quad (3)$$

Note that by making  $\Delta$  arbitrarily small, the differential entropy can be made arbitrarily negative, while taking  $\Delta$  arbitrarily large, the differential entropy becomes arbitrarily positive.

### Example 2

Consider a normal distributed random variable  $X$ , with mean  $m$  and variance  $\sigma^2$ . Then its density is  $\sqrt{\frac{1}{2\pi\sigma^2}} e^{-\frac{(x-m)^2}{2\sigma^2}}$ .

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We can then find its differential entropy as follows, first calculate  $-\log f(x)$ :

$$-\log f(x) = \frac{1}{2} \log(2\pi\sigma^2) + \log e^{-\frac{(x-m)^2}{2\sigma^2}} \quad (4)$$

Then since  $E((X-m)^2) = \sigma^2$ , we have

$$\begin{aligned} h(X) &= \frac{1}{2} \log(2\pi\sigma^2) + \frac{1}{2} \log e \\ &= \frac{1}{2} \log(2\pi e\sigma^2) \end{aligned} \quad (5)$$

## 2 Properties of the differential entropy

In the section we list some properties of the differential entropy.

- The differential entropy can be negative
- $h(X+c) = h(X)$ , that is translation **does not** change the differential entropy.
- $h(aX) = h(X) + \log|a|$ , that is scaling **does** change the differential entropy.

The first property is seen from both Example 1 and Example 2. The two latter can be shown by using (1).