# The mutual information: information theory lecture 9

COMSM0075 Information Processing and Brain

comsm0075.github.io

October 2020



The probability density isn't any old function. It is a density.











$$P(x \in [x_0, x_1)) = \int_{x_0}^{x_1} p_X(x) dx$$

Now what happens if we do a change of variable to y(x).

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Now what happens if we do a change of variable to y(x). For simplicity, assume y(x) is strictly monotonic so we can invert to get x(y).

Remember how to change variables in an integral:

$$dx = \left| \frac{dx}{dy} \right| dy$$

so if  $y_0 = y(x_0)$  and  $y_1 = y(x_1)$  we have

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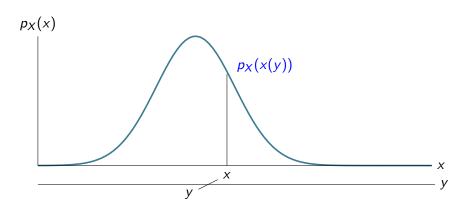
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The *y* probability density

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SO

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### The probability shouldn't change

$$\int_{y_0}^{y_1} p_Y(y) dy = \int_{y_0}^{y_1} p_X(x(y)) \left| \frac{dx}{dy} \right| dy$$

hence

$$p_Y(y) = \frac{p_X(x(y))}{|dy/dx|}$$

This behaviour is the definition of a density.

The entropy is not invariant under a change of variable!

$$h(Y) = h(X) + \int p_X(x) \log_2 \left| \frac{dy}{dx} \right| dx$$

## Scaling

Using y = ax in this formula

$$h(aX) = h(X) + \log |a|$$

#### The mutual information

or

$$I(X,Y) = h(X) + h(Y) - h(X,Y)$$

$$I(X,Y) = \int p(x,y) \log_2 \frac{p(x,y)}{p(x)p(y)}$$

The mutual information has nice properties

$$I(X, Y) = h(X) + h(Y) - h(X, Y)$$

is invariant under a change of variable; roughly speaking the Jacobian factors cancel!

The mutual information is the same for mutual information

$$I(X^{\delta_x}, Y^{\delta_y}) \to I(X, Y)$$

as  $\delta_{x}$  and  $\delta_{y}$  approach zero.

The mutual information is non-negative

$$I(X,Y)\geq 0$$