

## Independence

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*Definition:* Two events  $A$  and  $B$  are independent if  $P(A \cap B) = P(A)P(B)$ . Alternatively, they are independent if  $P(A|B) = P(A)$  and  $P(B|A) = P(B)$ .

Informally, two events are independent if they don't influence each other; knowing that  $A$  happened doesn't give you any additional information about  $B$ .

### Independence Example - Discrete

- Suppose that our sample space consists of  $N$  flips of a coin that has probability  $p$  of giving heads.
- The events corresponding to a  $H$  in position  $i$  and in position  $j$  are independent.
- The chance of getting  $k$  heads in  $N$  flips is

$$P(k, N) = \binom{N}{k} p^k (1-p)^{N-k}$$

The probability distribution on the set  $0, \dots, N$  given by this formula is called the *binomial distribution* for parameters  $p$  and  $N$ .

## Independence Example - Continuous

Suppose we have a thermometer that measures the temperature  $t$  within an error  $x = t - t_0$  from the true temperature, where  $x$  is normally distributed with standard deviation  $\sigma$ .

Suppose we make  $N$  independent measurements of the temperature. How are the errors distributed?

$$P(|x_1| < \delta, \dots, |x_N| < \delta) = \left( \frac{1}{\sigma\sqrt{2\pi}} \right)^N \int_{x_1=-\delta}^{\delta} \dots \int_{x_N=-\delta}^{\delta} e^{-(\sum_{i=1}^N x_i^2)/(2\sigma^2)} dx_1 \dots dx_N$$

This is the *multivariate gaussian* distribution.

### Non-independent events

Suppose we draw a pair of real numbers  $(x, y)$  from the plane  $R^2$  controlled by the distribution

$$P((x, y) \in U) = A \int_U e^{(-x^2 - xy - y^2)/(2\sigma^2)} dx dy$$

This density function has a bump at the origin and its level curves are ellipses.

The two coordinates are not independent of each other.

## Non-independent events

Multivariate Gaussian

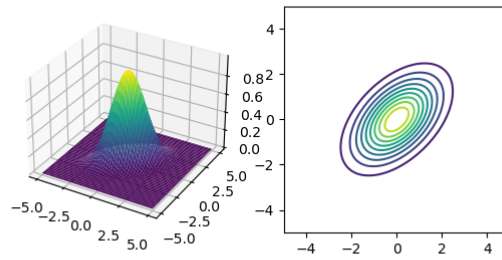


Figure 1: Multivariate Gaussian