

# Jensen's inequality

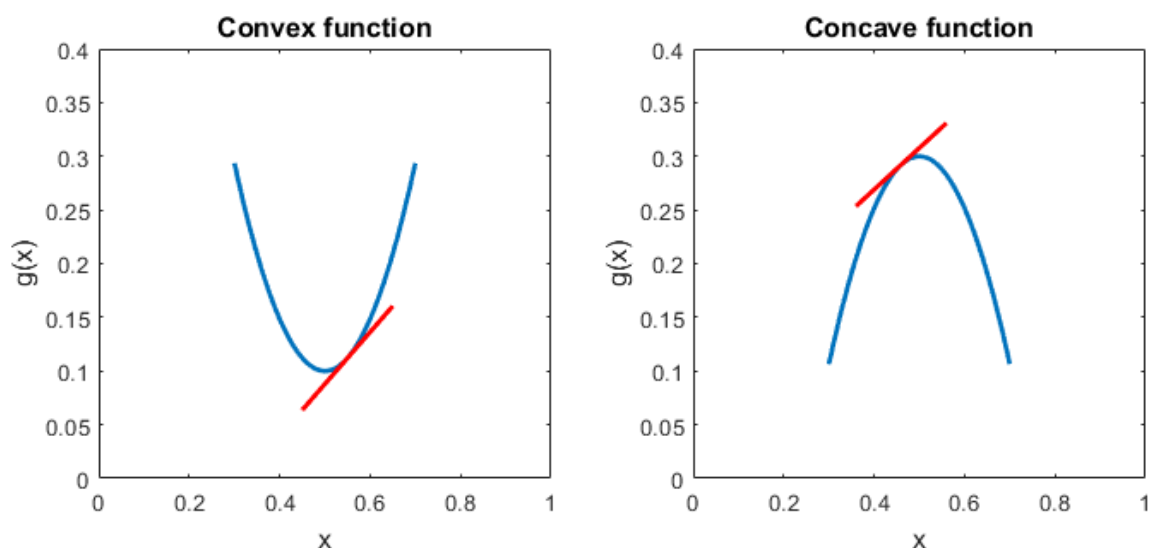
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Jensen's inequality is a probabilistic inequality that concerns the expected value of convex and concave transformations of a random variable.



## Convex and concave functions

Jensen's inequality applies to convex and concave functions.



The properties of these functions that are relevant for understanding the proof of the inequality are:

- the tangents of a convex function lie entirely below its graph;
- the tangents of a concave function lie entirely above its graph.

Also remember that a differentiable function is:

- (strictly) convex if its second derivative is (strictly) positive;
- (strictly) concave if its second derivative is (strictly) negative.

## Statement

The following is a formal statement of the inequality.

**Proposition** Let  $X$  be an integrable random variable. Let  $g : \mathbb{R} \rightarrow \mathbb{R}$  be a convex function such that

$$Y = g(X)$$

is also integrable. Then, the following inequality, called Jensen's inequality, holds:

$$E[g(X)] \geq g(E[X])$$

## Proof

If the function  $g$  is strictly convex and  $X$  is not **almost surely** constant, then we have a strict inequality:

$$E[g(X)] > g(E[X])$$

### Proof

If the function  $g$  is concave, then

$$E[g(X)] \leq g(E[X])$$

### Proof

If the function  $g$  is strictly concave and  $X$  is not almost surely constant, then

$$E[g(X)] < g(E[X])$$

### Proof

## Example

Suppose that a strictly positive random variable  $X$  has expected value

$$E[X] = 1$$

and it is not constant with probability one.

What can we say about the expected value of  $\ln(X)$ , by using Jensen's inequality?

The natural logarithm is a strictly concave function because its second derivative

$$\frac{d^2}{dx^2} \ln(x) = -x^{-2}$$

is strictly negative on its domain of definition.

As a consequence, by Jensen's inequality, we have

$$E[\ln(X)] < \ln(E[X]) = \ln(1) = 0$$

Therefore,  $\ln(X)$  has a strictly negative expected value.

## Important applications

Jensen's inequality has many applications in statistics. Two important ones are in the proofs of:

- the non-negativity of the Kullback-Leibler divergence;
- the information inequality concerning the expected value of the log-likelihood.

## Other inequalities

If you like this page, StatLect has other pages on probabilistic inequalities:

- [Markov's inequality](#);
- [Chebyshev's inequality](#).

## Solved exercises

Below you can find some exercises with explained solutions.

### Exercise 1

Let  $X$  be a random variable having finite mean and variance  $\sigma^2 > 0$ .

Use Jensen's inequality to find a bound on the expected value of  $X^2$ .

**|** Solution

### Exercise 2

Let  $X$  be a positive integrable random variable.

Find a bound on the mean of  $\sqrt{X}$ .

**|** Solution

## How to cite

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