

Convergence, LLN and CLT

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Convergence of Functions

In calculus, we used to work with *sequences of numbers*:

$$x_1, x_2, x_3, \dots, x_n, \dots$$

and in case the sequence got closer and closer to some number x as n gets larger and larger, we said that the sequence *converges* to x :

$$\lim_{n \rightarrow \infty} x_n = x$$

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If instead, we had a *sequence of functions*:

$$f_1(a), f_2(a), f_3(a), \dots, f_n(a), \dots$$

and at some point a , their values got closer to some number b as $n \rightarrow \infty$, we say that the sequence $\{f_n\}$ converges to b **at the point a** :

$$\lim_{n \rightarrow \infty} f_n(a) = b$$

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As we see, this sequence of functions converges to 0 at **every point** $a \in \mathbb{R}$:

$$\lim_{n \rightarrow \infty} f_n(a) = 0 \quad \text{for all } a \in \mathbb{R}$$

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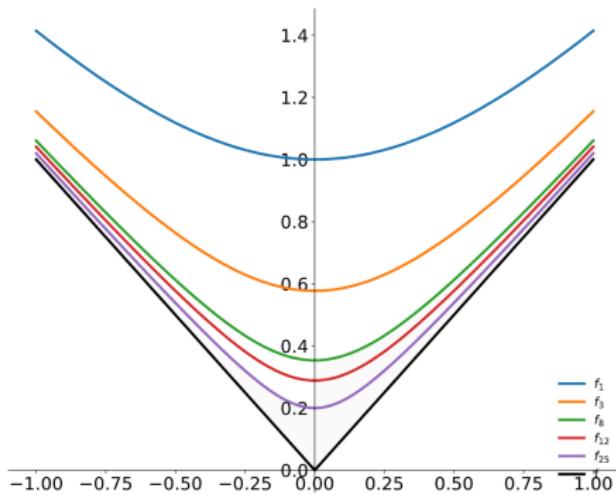
or simply that the sequence of functions $\{f_n\}$ converges to the function g .

Convergence of Functions

Definition

A sequence of functions $\{f_n\}$ *converges* to a function g , if for every point a and for every $\varepsilon > 0$, there exists a natural number N such that for all $n \geq N$,

$$|f_n(a) - g(a)| < \varepsilon$$



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i.e.

```
X = lambda omega: the height of omega
X(Arthur Abraham)
>> 1.75
```

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So can we talk about the convergence of random variables?

As it turns out, the answer is yes.

Convergence of Random Variables

Everyday Anahit randomly chooses a number from the interval $[0, 1]$ and multiplies the number by itself n times, where n is the number of days since she started doing this.

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- If on the first day, she chooses 0.5, then $X_1 = 0.5^1 = 0.5$

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Question

What happens to X_n as n gets larger and larger? Does it converge to some number?

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If something happens with probability 1 (i.e. it can also not happen, but this event has probability zero) we say that it happens *almost surely (a.s.)*.

We distinguish between three possible ways how a sequence of random variables $\{X_n\}$ can converge to a random variable X :

- The strongest case: $X_n \rightarrow X$ *almost surely*

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- The weakest case: $X_n \xrightarrow{d} X$ *in distribution*
when just the PMF/PDFs of X_n converge to the PMF/PDF of X – but the values of X_n do not necessarily get closer to the value of X .

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Almost sure \Rightarrow in probability \Rightarrow in distribution

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What can we say about the convergence of X_n ?

Law of Large Numbers

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Law of Large Numbers (LLN)

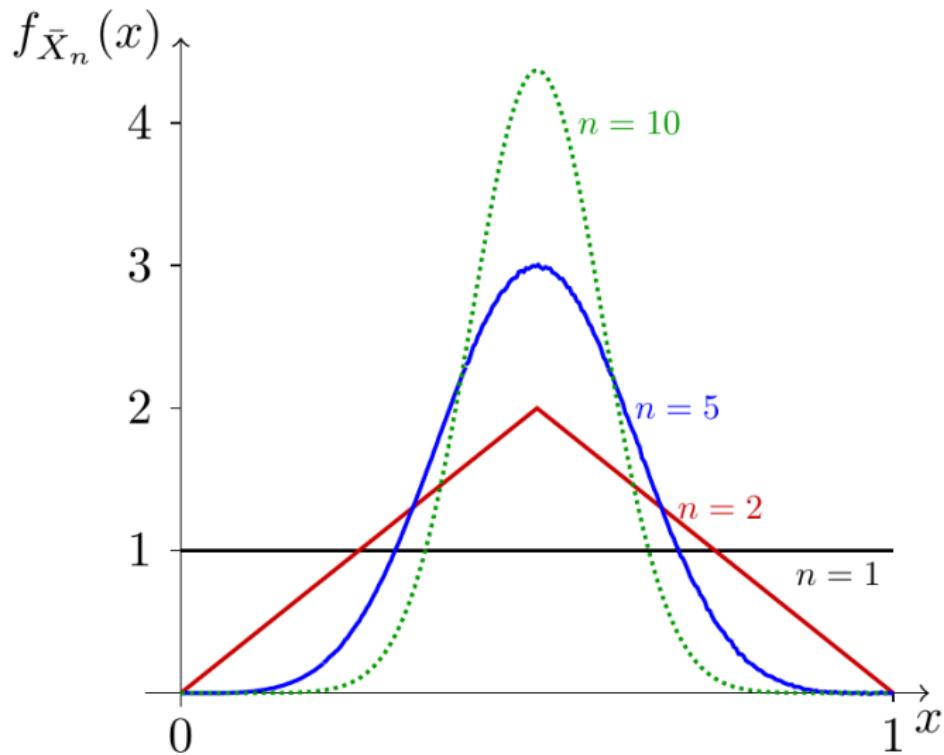
Let X_1, X_2, \dots be a sequence of independent and identically distributed (i.i.d.) random variables with the same expected value:

$$\mathbb{E}[X_k] = m$$

and with finite variance (again, same). Then their average

$$\frac{X_1 + X_2 + \cdots + X_n}{n} \xrightarrow{p} m \quad \text{as } n \rightarrow \infty$$

Law of Large Numbers



Central Limit Theorem

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Central Limit Theorem (CLT)

Let X_1, X_2, \dots be a sequence of independent and identically distributed (i.i.d.) random variables with the same expected value:

$$\mathbb{E}[X_k] = m$$

and with finite variance

$$\text{Var}[X_k] = \sigma^2$$

Then their normalized average

$$\sum_{k=1}^n \frac{X_k - m}{\sigma\sqrt{n}} \xrightarrow{d} N(0, 1) \quad \text{as } n \rightarrow \infty$$