# Applied Statistics Central Limit Theorem







hali Mar Sulp - 1823





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"Statistics is merely a quantisation of common sense"

# Law of large numbers

#### LAW OF LARGE NUMBERS IN AVERAGE OF DIE ROLLS

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# Adding random numbers

If each of you chose a random number from your own favorit distribution\*, and we added all these numbers, repeating this many times...

# What would you expect?

\* OK - to be nice to me, you agree to have similar RMSs in these distributions! 3

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Central Limit Theorem:

The sum of N *independent* continuous random variables  $x_i$  with means  $\mu_i$  and variances  $\sigma_i^2$  becomes a Gaussian random variable with mean  $\mu = \Sigma_i \mu_i$  and variance  $\sigma^2 = \Sigma_i \sigma_i^2$  in the limit that N approaches infinity.

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The Central Limit Theorem holds under fairly general conditions, which means that the Gaussian distribution takes a central role in statistics...

#### The Gaussian is "the unit" of distributions!

Since measurements are often affected by many small effects, uncertainties tend to be Gaussian (until otherwise proven!).

Statistical rules often require Gaussian uncertainties, and so **the central limit theorem is your new good friend.** 



Take the sum of 100 uniform numbers! Repeat 100000 times to see what distribution the sum has...



The result is a bell shaped curve, a so-called **normal** or **Gaussian** distribution.

It turns out, that this is very general!!!

Now take the sum of just 10 uniform numbers!



Now take the sum of just 5 uniform numbers!



Now take the sum of just 3 uniform numbers!



This time we will try with a much more "**nasty**" function. Take the sum of 100 *exponential* numbers! Repeat 100000 times to see the sum's distribution...



Even with such a non-Gaussian skewed distribution, the sum quickly becomes

Gaussian!!!

It turns out, that this fact saves us from much trouble: Makes statistics "easy"!

Distribution in the Population

Sampling Distribution of the Mean,  $\bar{X}$ 

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# Central Limit Theorem for general PDFs

It doesn't matter what shape the input PDF has, as long as it has finite mean and width (which all numbers from the real world has!)



Looking at z-coordinate of tracks at vertex from proton collisions in CERNs LHC accelerator by the ATLAS detector, this is what you get:



# The Gaussian distribution

Almost every field of science have their own terms for features of the Gaussian, also known as the "normal" distribution.



# The Gaussian distribution

It is useful to know just a few of the most common Gaussian integrals:

2.1%

2σ

 $-1\sigma$ 

0.1%

-3σ

4

0.3

0.2

0.1

0.0

	1		
few of the	Range	Inside	Outside
ntegrals:	$\pm 1\sigma$	<b>68</b> %	32~%
	$\pm 2\sigma$	95 %	5 %
	$\pm 3\sigma$	99.7 %	0.3~%
	$\pm 5\sigma$	99.99995~%	0.00005~%
34.	1% 34.19	2 1	%
13.6%		13.6%	0.1%

1σ

μ

2σ

3σ

## Summary

# The Central Limit Theorem

... is your good friend because it...

#### ensures that uncertainties tend to be Gaussian

...which are the easiest to work with!

