

# Central Limit Theorem

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- ▶ Used in many cases, perhaps most importantly in regression
- ▶ Establishes that when independent random variables are summed, their properly normalized sum tends towards a normal distribution *irrelevant* of the variables' distributions
- ▶ Implies that probabilistic and statistical methods that work for normal distributions can be applicable to many problems involving other types of distributions

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- ▶  $Z$  is a *standard* normal distribution

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- ▶ If this procedure is performed many times, the central limit theorem says that the probability distribution of the average will closely approximate a normal distribution
- ▶ A simple example of this is that if one flips a coin many times, the probability of getting a given number of heads will approach a normal distribution, with the mean equal to half the total number of flips
- ▶ At the limit of an infinite number of flips, it will equal a normal distribution

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- ▶ Various types of statistical inference on the regression assume that the error term is normally distributed
- ▶ This assumption can be justified by assuming that the error term is actually the sum of many independent error terms; even if the individual error terms are not normally distributed, by the central limit theorem their sum can be well approximated by a normal distribution



## Illustration

```
http://195.134.76.37/applets/  
AppletCentralLimit/App1_CentralLimit2.  
html
```