

# STAT 801=830

## Convergence of RVs

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# Purposes of These Notes

- Define convergence in probability, in mean, in quadratic mean.
- Define almost sure convergence.
- Define convergence in distribution.
- Compare and contrast these definitions.



- Contrast two statements:
  - ▶  $X$  and  $Y$  are close together.
  - ▶  $X$  and  $Y$  have similar distributions.
- Truth of first statement depends on *joint* distribution of  $X$  and  $Y$ .
- Truth of second statement depends on *marginal* distributions of  $X$  and  $Y$ .
- Both ideas used in *large sample theory*: describing behaviour of statistical procedures *approximately* in presence of lots of data.



# Relation between convergence and approximation

- Some approximations and the limits they come from.
- Stirling's approximation:

$$n! \approx \sqrt{2\pi n} n^{n+1/2} e^{-n} \equiv s_n$$

$n$	$n!$	$s_n$	$n!/s_n$
5	120	118.019	1.01678
10	3628800	3598695.619	1.008365

- Normal approximation to Binomial. Toss coin 100 times, get  $X$  heads.

$$P(40 \leq X \leq 60) \approx \Phi\left(\frac{60 - 50}{\sqrt{25}}\right) - \Phi\left(\frac{40 - 50}{\sqrt{25}}\right) = 0.9544997.$$

- Same context better approximation

$$P(40 \leq X \leq 60) \approx \Phi\left(\frac{60.5 - 50}{\sqrt{25}}\right) - \Phi\left(\frac{39.5 - 50}{\sqrt{25}}\right) = 0.9642712$$

- Same context

$$P(X = 50) \approx ?$$



# Associated Limits

- Theorem:

$$\lim_{n \rightarrow \infty} \frac{n!}{\sqrt{2\pi n} n^{n+1/2} e^{-n}} = 1.$$

- Used when  $n = 100$  to give, for instance,

$$100! \approx \sqrt{2\pi 100} 100^{100} e^{-100} = 9.3326215443944152682 \times 10^{157}$$

- All those digits are right!
- Theorem: if  $X_n \sim \text{Binomial}(n, 1/2)$  then

$$\lim_{n \rightarrow \infty} P\left(\frac{X_n - n/2}{\sqrt{n/4}} \leq x\right) = \Phi(x)$$

- Used with  $x\sqrt{100/4} + 100/2$  equal to 60 and 40 or 60.5 and 39.5.
- Used with 49.5 and 50.5 to get  $P(X_{100} = 50)$  approximately.



# Summary

- If we want to compute  $x_{100}$  we compute  $y = \lim_{n \rightarrow \infty} x_n$  and approximate  $x_{100} \approx y$ .
- There are often many different ways to think of  $x_{100}$  as an entry in some sequence! Get slightly different approximations.
- And some of the approximations are lousy; some are great.



# Limits of Random Variables NOT Distributions

- We do an experiment to measure probability that a dropped tack lands point up.
- Drop tack  $n$  times, observe  $X_n \sim \text{Binomial}(n, p)$ , which is number of times tack lands point up.
- Two common random variables to study:

$$U_n \equiv \frac{X_n - np}{\sqrt{np(1-p)}} \quad \text{and} \quad V_n \equiv \frac{X_n - np}{\sqrt{n\hat{p}(1-\hat{p})}}$$

where  $\hat{p} = X/n$ .

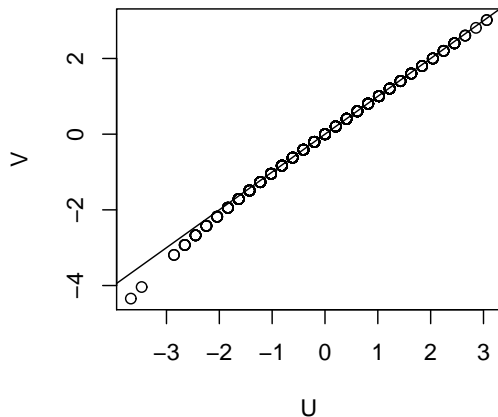
- First used in testing, second to form confidence intervals.
- Approximation is

$$U_n \approx V_n$$

- Generate 1000 values of  $X$  with  $p = 0.4$ . Plot  $U$  vs  $V$ .



# Effect of estimating standard error



# Convergence in Probability and Almost Surely pp 72-75,81

- In this case the plot shows

$$P(|U_n - V_n| \text{ is big})$$

is small. (The points are close to the line  $y = x$ .)

- **Def'n:** A sequence of random variables  $X_n$  *converges in probability* to a random variable  $X$  if for every  $\epsilon > 0$  we have

$$\lim_{n \rightarrow \infty} P(|X_n - X| > \epsilon) = 0.$$

- **Def'n:** A sequence of random variables  $X_n$  *converges almost surely* to a random variable  $X$  if

$$P\left(\lim_{n \rightarrow \infty} X_n = X\right) = 1.$$



## Convergence in $p$ th mean especially quadratic

- **Def'n:** A sequence of random variables  $X_n$  *converges in mean* or *converges in  $L_1$*  to a random variable  $X$  if

$$\lim_{n \rightarrow \infty} E(|X_n - X|) = 0.$$

- **Def'n:** A sequence of random variables  $X_n$  *converges in quadratic mean* or *converges in  $L_2$*  to a random variable  $X$  if

$$\lim_{n \rightarrow \infty} E(|X_n - X|^2) = 0.$$

- For  $p$ th mean we use

$$\lim_{n \rightarrow \infty} E(|X_n - X|^p) = 0.$$



## Back to our example

- In fact  $U_n - V_n$  converges to 0 in probability.
- And  $U_n - V_n$  converges to 0 almost surely.
- But they do not converge in  $p$ th mean because  $V_n$  does not have a finite mean ( $P(\hat{p} = 0) > 0$ ).

