

AUA CS 108, Statistics, Fall 2019

Lecture 16

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Contents

- ▶ Convergence Types of R.V. Sequences
- ▶ LLN and CLT

Last Lecture ReCap

- ▶ Give the definition of the convergence in the a.s./ Probability / QM / Distributions sense.

Example

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and X_n are defined on the same Probability Space. Which of the followings are true (use only the definitions):

- ▶ $X_n \xrightarrow{\mathbb{P}} 0;$
- ▶ $X_n \xrightarrow{qm} 0;$
- ▶ $X_n \xrightarrow{D} 0 ?$

Solution: OTB

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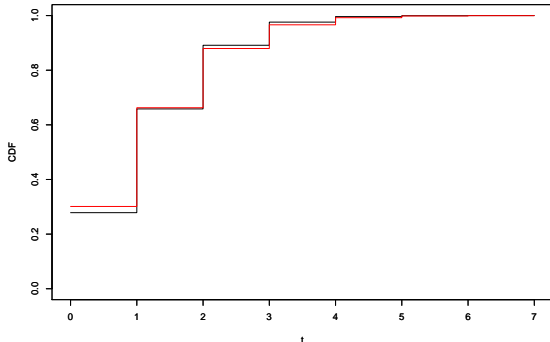
Note: Note that when using $X_n \xrightarrow{D} \text{Pois}(\lambda)$ we mean $X_n \xrightarrow{D} X$, where $X \sim \text{Pois}(\lambda)$.

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```
lambda <- 1.2; n <- 10; t <- seq(0,7, 0.1)
plot(t,pbinom(t, size = n, prob = lambda/n), type = "s", ylim = c(0,1), ylab = "CDF")
par(new = T)
plot(t, ppois(t, lambda = lambda), type = "s", col = "red", ylim = c(0,1), ylab = "CDF")
```



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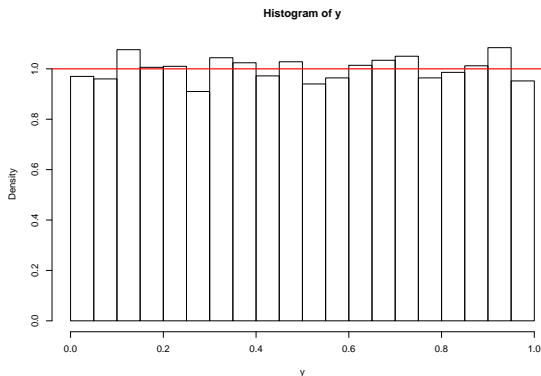
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```
n <- 10000 ## We use Y_n  
m <- 10000 ## No. of generated numbers  
y <- runif(m, min = 0, max = n)/n  
hist(y, freq = F)  
abline(h = 1, col = "red", lwd = 2)
```



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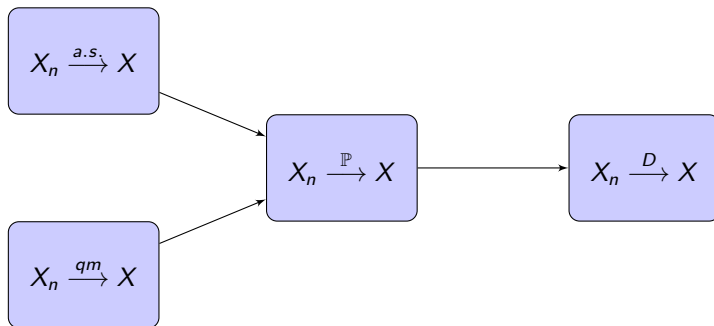
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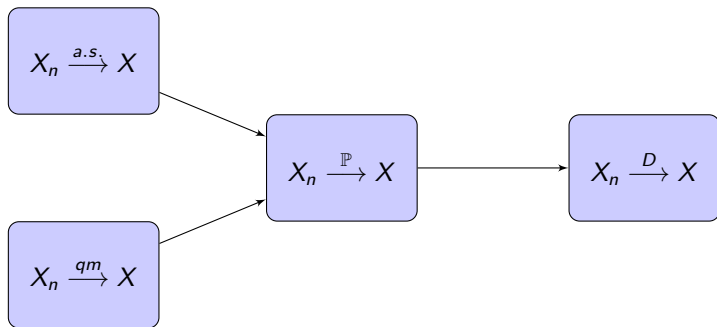
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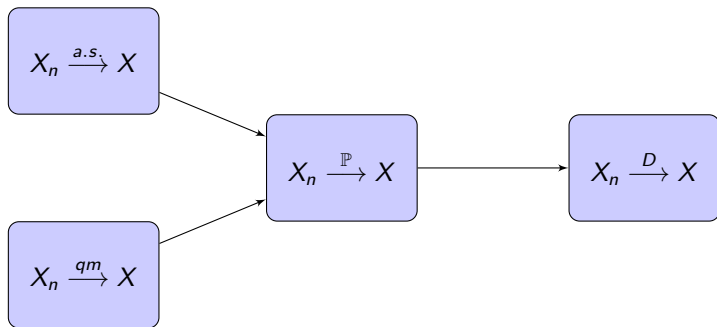
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Note: Inverse implications are not always correct. But, say, the following holds: If $X_n \xrightarrow{D} X$ and $X \equiv \text{constant}$, then $X_n \xrightarrow{\mathbb{P}} X$ (X_n and X are defined on the same Probability space).

Limit Theorems

Sequence of IID r.v.

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- ▶ X_n -s are independent. Say, in particular,

$$\text{Var}(X_1 + X_2 + \dots + X_n) = \text{Var}(X_1) + \text{Var}(X_2) + \dots + \text{Var}(X_n) = n \cdot \text{Var}(X_1).$$

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