

## Recap: Modes of Convergence

$\{X_n, n \geq 0\}$  converges to  $X$  pointwise or surely if for all  $\omega \in \Omega$  we have  $\lim_{n \rightarrow \infty} X_n(\omega) = X(\omega)$

$X_n$  converges to  $X$  almost surely if  $P(\omega \in \Omega : \lim_{n \rightarrow \infty} X_n(\omega) = X(\omega)) = 1$ .

$\{X_n, n \geq 0\}$  is a sequence of i.i.d random variables with mean  $\mu$  and  $S_n = \sum_{i=1}^n X_i$ . Then  $\hat{\mu}_n := \frac{S_n}{n} \rightarrow \mu$  a.s. (SLLN)

- ▶ Estimator  $\hat{\mu}_n$  has mean  $\mu$  and Variance  $\frac{\sigma^2}{n}$ .
- ▶  $\hat{\mu}_{n+1} = \hat{\mu}_n + \frac{1}{n+1} [X_{n+1} - \hat{\mu}_n]$

# Borel Cantelli Lemma

Self-Study: Theorem 7.5 (probabilitycourse.com)

Consider a sequence of random variables  $X_1, X_2, \dots$ . If for all  $\epsilon$  we have

$$\sum_{n=1}^{\infty} P(|X_n - X| > \epsilon) < \infty$$

then  $X_n \rightarrow X$  a.s.

- ▶ This is only a sufficient condition for almost sure convergence!
- ▶ Thm 7.6 (HW) gives necessary and sufficient conditions.
- ▶ Lot of problems in probabilitycourse, practice them!

## Another example of a.s. convergence

- ▶ Consider a uniform r.v.  $U$  and define  $X_n = n1_{\{U \leq \frac{1}{n}\}}$ .
- ▶  $X_n = n$  when  $U \leq \frac{1}{n}$  and  $X_n = 0$  otherwise.
- ▶ Given a realization of  $U$ , what can you say about the sequence  $\{X_n\}$  ?
- ▶ Once an  $X_n$  is zero, all higher indexed variables are also zero!
- ▶ This happens for all realizations  $U$  other than  $U = 0$ . In this case since  $0 \leq \frac{1}{n}$  for all  $n$ ,  $X'_n$ s run off to infinity and we don't see convergence to 0.
- ▶ But  $P(U = 0) = 0$ .
- ▶ Does  $E[X_n] \rightarrow 0$  ?
- ▶ Almost sure convergence does not imply their means converge!

# Towards convergence in probability

- ▶ Now define  $X_n = n1_{\{U_n \leq \frac{1}{n}\}}$  where  $\{U_n\}$  are i.i.d uniform.
- ▶  $X_n = n$  when  $U_n \leq \frac{1}{n}$  and  $X_n = 0$  otherwise.
- ▶ What can you say about the sequence  $\{X_n\}$  ?
- ▶ Is it true that once an  $X_n$  is zero, all higher indexed variables are also zero!? No!
- ▶ Every time (on every run of the experiment or every sample path), we will have a sequence of zero and non-zero values, where the non-zero values become rarer and rarer but will keep happening once in a while.
- ▶ On no sample path would you see convergence to zero but occurrence of non-zero values become rare.
- ▶ We now characterize this notion of convergence.

# Convergence in probability (w.h.p)

$X_n$  converges to  $X$  in probability if

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

- ▶ How would you compute  $P(|X_n - X| > \epsilon)$  when  $X_n, X$  are either continuous or discrete random variables ?
- ▶ Ex:  $X_n = n$  with probability  $\frac{1}{n}$  and  $X_n = 0$  otherwise.
- ▶  $P(|X_n - X| > \epsilon) = P(X_n > \epsilon) = \frac{1}{n}$  when  $n > \epsilon$ .
- ▶ When  $n < \epsilon$ , we have  $P(|X_n - X| > \epsilon) = 0$ .
- ▶ Once  $n > \epsilon$  we have  $\lim_{n \rightarrow \infty} P(X_n > \epsilon) = \lim_{n \rightarrow \infty} \frac{1}{n} = 0$ .
- ▶  $X_n$  converges to 0 in probability, but not almost surely.
- ▶ a.s. convergence implies convergence in probability

# Convergence in $r^{th}$ mean

$X_n$  converges to  $X$  in  $r^{th}$  mean if

$$\lim_{n \rightarrow \infty} E[|X_n - X|^r] = 0.$$

- ▶ How will you compute  $E[|X_n - X|^r]$ ?
- ▶ When  $r = 2$ , it is convergence in mean squared sense. In addition if  $X = 0$ , it implies that the second moments converge to 0.
- ▶ In the convergence in probability example, do we have convergence in mean or mean square?
- ▶ Convergence in  $r^{th}$  mean implies convergence in probability.

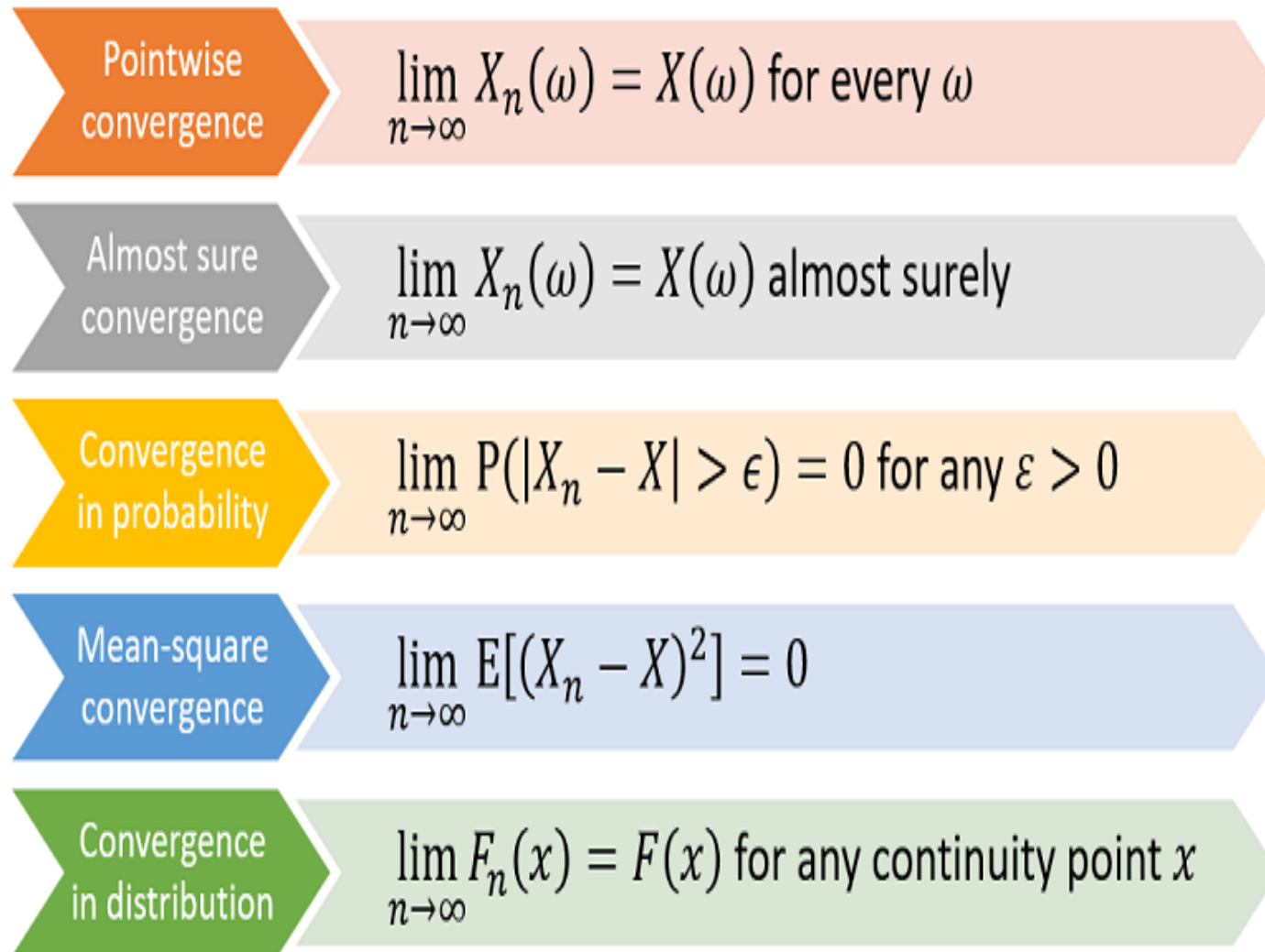
# Weak convergence (in distribution)

$X_n$  converges to  $X$  in distribution if

$$\lim_{n \rightarrow \infty} F_{X_n}(x) = F_X(x) \text{ for all continuity points of } F_X(\cdot).$$

- ▶ a.s. convergence and convergence in probability imply convergence in distribution.
- ▶ Example:  $X_n$  is an exponential random variable with parameter  $\lambda n$ .
- ▶ In this case,  $F_{X_n}(x) = 1 - e^{-n\lambda x}$  and  $F_X(x) = 1$  for all  $x$ .
- ▶ Note  $x = 0$  is point of discontinuity as  $F_X(0) = 1$  and  $F_{X_n}(0) = 0$ .
- ▶ HW EX2:  $X_n$  are i.i.d Binomial( $n, \frac{\lambda}{n}$ ). It converges in distribution to Poisson( $\lambda$ ).

# Summary



[https://en.wikipedia.org/wiki/Convergence\\_of\\_random\\_variables](https://en.wikipedia.org/wiki/Convergence_of_random_variables)



# Relation between modes of convergence (no proofs)

Almost sure  
convergence

Mean square  
convergence



Convergence  
in probability



Convergence  
in distribution

[https://en.wikipedia.org/wiki/Proofs\\_of\\_convergence\\_of\\_random\\_variables](https://en.wikipedia.org/wiki/Proofs_of_convergence_of_random_variables)