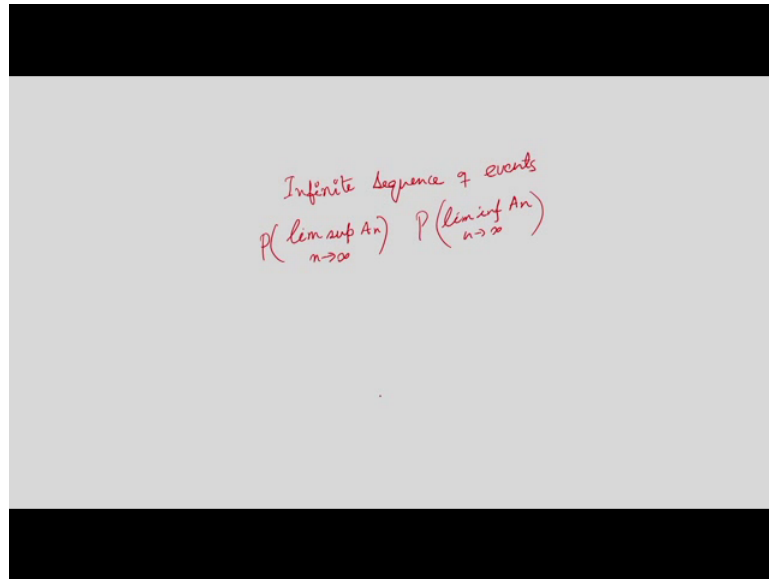


Advanced Topics in Probability and Random Processes
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Lecture - 07
Convergence of a Sequence of Random Variables

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In the last lectures we discussed sequence of events that is infinite sequence of events and limits. Limits of these events, limits $\limsup$ we define, $\limsup$ and $\lim$. Suppose, sequence is n , this event we defined. Similarly, $\liminf$ as n tends to infinity, this type of limit we define and then how to find out the probability for this limiting events. We discussed different theorems and particularly Borel- Cantelli lemmas how to find out this probability, probability of $\limsup A_n$ under two different conditions.

Now, this application of convergence of infinite sequence of events is the convergence of sequence of random variables. Now, we will be discussing the convergence concepts for random variables and these concepts are useful in many practical applications, just like it is the basis of stochastic calculus. Also we will be discussing, very important applications like central limit theorem and we log up large numbers.

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Convergence of a sequence of real numbers

Consider a sequence of real numbers $\{x_n\}_{n=1}^{\infty}$. The sequence converges to a limit x if corresponding to every $\varepsilon > 0$, we can find a positive integer N such that $|x - x_n| < \varepsilon$ for all $n > N$. For example, the sequence $\{x_n = \frac{1}{n}\}$ converges to the number 0. Because, for any $\varepsilon > 0$, we can choose a positive integer $N > \frac{1}{\varepsilon}$ such that $|0 - x_n| = \frac{1}{n} < \varepsilon$ for $n > N$.

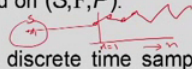
First of all let us see: what is convergence of a sequence of real numbers. Consider a sequence of real numbers x_n , n going from 1 to infinity. This sequence converges to a limit x if corresponding every epsilon greater than 0, we can find a positive integer N such that $|x - x_n|$ can be made smaller than epsilon for all n greater than N . So, what does it say that, this difference between the limit and the sequence can be made as small as possible where epsilon is an arbitrarily small number positive number, for sufficiently large N . Then we say that x_n converges to x . So, for example, if I consider this sequence suppose x_n is equal to $\frac{1}{n}$ simple example.

Now, we know that limit of $\frac{1}{n}$ as n tends to infinity 0. Now, how it confirms to this definition, suppose for any epsilon greater than 0, we can choose a positive integer N greater than $\frac{1}{\varepsilon}$. So, that in that case $|0 - x_n|$ will be $\frac{1}{n}$ by that is equal to $\frac{1}{n}$ and this will be less than epsilon. So, because we are considering N is greater than $\frac{1}{\varepsilon}$ therefore, for any n greater than capital N this will be valid. So, that way this 0 will be limit of this sequence. And now we have to extend this definition of convergence or limit to sequence of random variables.

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Sequence of Random Variables

Consider a sequence of random variables $\{X_n\}_{n=1}^{\infty}$ defined on $(S, \mathcal{F}, P)$.



- For each $s \in S$, $X_1(s), X_2(s), \dots, X_n(s), \dots$ represent a discrete time sample function.
- $\{X_n\}_{n=1}^{\infty}$ represents a family of sequences.
- Convergence of a random sequence $\{X_n\}_{n=1}^{\infty}$ is to be defined using different criteria.

Consider a sequence of random variable X_n , n going from 1 to infinity defined on $S, \mathcal{F}, P$. We are defining suppose a sequence of random variables. Now what does it mean? That means, if we have a sample space suppose S , this is the sample space and now corresponding to suppose any element s here. So, that will map to suppose, this is my time axis n , n is equal to 1 here. So, n is equal to 1 it will be 1 point. Similarly, n is equal to 2 it may be another point like that this s will be mapped to different points at different instant of time. So, that way I may get a sequence, may be if I join this may be something like this.

So, that is that means, corresponding to every sample point I will get a sample functions like this. So, that means, fall is S belonging to sample space $X_1(s), X_2(s)$ up to $X_n(s)$ represent a discrete time sample function particular for each s . So, that will be, we will have a sample function like this. Therefore, since the sample space will contain other sample points therefore, corresponding to each sample point will get a sample function like this. So, unlike sequence of real numbers action sequence of random variables a represents a family of sequences.

So, each sample function is a sequence therefore, it is a family of sequences and here are the problem in defining the convergence for the sequence of a real numbers. So, convergence of a random sequence is to be defined using different criteria because, it is

since it is not a single sequence, we have to use different criteria to, to the define the convergence of a sequence of random variables.

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Convergence Everywhere *or Point-wise Convergence*

A sequence of random variables $\{X_n\}_{n=1}^{\infty}$ is said to converge everywhere to X if

$$\lim_{n \rightarrow \infty} X_n(s) = X(s) \quad \forall s \in S.$$

Example: Suppose $S = \{s_1, s_2\}$. Define the sequence of RVS by

$$X_n(s_1) = 1 + \frac{1}{n} \text{ and } X_n(s_2) = -1 + \frac{1}{n}$$

Define a random variable X such that

$$X(s_1) = 1 \text{ and } X(s_2) = -1$$

Now, $\lim_{n \rightarrow \infty} X_n(s_1) = 1$ and $\lim_{n \rightarrow \infty} X_n(s_2) = -1$

$$\therefore \lim_{n \rightarrow \infty} X_n(s) = X(s) \quad \forall s$$

This case is the simple extension of convergence of a sequence of real numbers.

Therefore, there are different modes of convergence of a sequence of random variables. The most elementary convergence concept is convergence everywhere or point - wise convergence, it is also called point - wise convergence. A sequence of random variables X_n is said to converge everywhere to X , if limit of X_n is equal to X of s for all s belonging to S . So, suppose corresponding to each sample point we will find a limit. So, in that case this is the point wise convergent or convergent everywhere. We can consider one simple example. Suppose define a sequence of random variables suppose X only sample space comprises of two sample points s_1 and s_2 .

Now, suppose $X_n(s_1)$ is equal to $1 + \frac{1}{n}$ and $X_n(s_2)$ is equal to $-1 + \frac{1}{n}$. We observe that as n tends to it infinity, this point will become 1 and this point will become minus 1. So, that way we can now define the a random variable X such that suppose X of s_1 is equal to 1 and X of s_2 is equal to minus 1. So obviously, limit of X_n of s_1 will be 1 and limit of X_n of s_2 will be minus 1. Therefore, in this case the sequence X_n , this X_n sequence converges point wise or converges everywhere to random variable X . So, that way convergence everywhere is this just an extension of convergence of a sequence of real numbers. So, here we are not, not considering the

property of the random variable, we are simply considering this as a sequence of function.

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Almost sure (a.s.) convergence or convergence with probability 1

Consider the event $\{s \mid \lim_{n \rightarrow \infty} X_n(s) = X(s)\}$ defined on $(S, \mathcal{F}, P)$. *Strong convergence*

The sequence $\{X_n\}_{n=1}^{\infty}$ is said to converge to X almost sure or with probability 1 if $P\left(\{s \mid \lim_{n \rightarrow \infty} X_n(s) = X(s)\}\right) = 1$

Equivalently, for every $\varepsilon > 0$, there exists N such that $P\{s \mid |X_n(s) - X(s)| < \varepsilon \text{ for all } n \geq N\} = 1$. We write $\{X_n\} \xrightarrow{a.s.} X$ in this case

a.s. convergence is a mode of strong convergence

Next we will, consider one concept which is known as convergence almost here or convergence with probability 1, this is also known as the strong mode of convergence, this is strong convergence. Let us consider an event that is it is comprising of points S such that limit of $X_n(s)$ is equal to $X(s)$. So, we are considering all those sample points for which limit of $X_n(s)$ is equal to $X(s)$ there may be some sample points where this limit is not equal to $X(s)$ ok, but we are considering only those sample points for which this limit is equal to $X(s)$.

Since, decision even define on the on $S, \mathcal{F}, P$ we can find out the probability of this event. This sequence now X_n is said to converge to X almost here or with probability 1, if probability of this event that is probability of the event s such that limit of $X_n(s)$ equal to $X(s)$ is equal to 1. So, probability of this event is equal to 1. So, that is the definition of almost your convergence; that means, wherever there is a convergence. So, those sample points if we consider then the corresponding event should have probability 1.

Equivalently for every epsilon greater than 0 there exists N such that a probability of this event is less than epsilon for all n greater than N , this is the event we are considering that is S such that $|X_n(s) - X(s)| < \varepsilon$ for all n greater than certain N . Suppose epsilon corresponding to each epsilon we will get a big N number

and suppose where this convergence takes place for all n greater than equal to N , that probability is equal to 1. We write X_n converges to X almost here this is the symbol we use, as I have told this is a, mod of strong conversions.

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Example: Suppose $S = \{s_1, s_2, s_3\}$ and $\{X_n\}_{n=1}^{\infty}$ be a sequence of random variables with $X_n(s_1) = 1, X_n(s_2) = -1$ and $X_n(s_3) = n$. Define a random variable X such that $X(s_1) = 1, X(s_2) = -1$ and $X(s_3) = 1$.

$$\therefore \{s \mid X_n(s) \rightarrow X(s)\} = \{s_1, s_2\}$$

$$\Rightarrow P(\{s \mid \lim_{n \rightarrow \infty} X_n(s) = X(s)\}) = P(\{s_1, s_2\})$$

Therefore $\{X_n\} \xrightarrow{a.s.} X$ if $P(\{s_1, s_2\}) = 1$

Let us consider one example, suppose sample space is s_1, s_2, s_3 , it comprises of three elements and the X_n be a sequence of random variables with X_n of s_1 is equal to 1, X_n of s_2 is equal to minus 1. And, suppose this is a divergent sequence X_n of s_3 is equal to n and now, we will define another random variable on the same sample space X s_1 is equal to 1, X s_2 is equal to minus 1 and X of s_3 is equal to 1. So, here it is divergent, here it is 1.

Therefore, if I consider the event for where, which this convergence is taking place I, I we see that X_n s_1 is 1 and here also 1, X_n s_2 is minus 1, here also minus 1, but X_n s_3 is 1, here it is 1 therefore, it is not converging for s_3 . So, that way this event now where that convergence is taking place that is that comprises of s_1 and s_2 therefore, probability of this limiting event now is equal to probability of s_1 s_2 . So, when X_n will converge to X almost here if probability of this event is equal to 1, so that way the basic concepts of almost sure convergence is introduced.

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Test for a.s. Convergence

$$\{X_n\} \xrightarrow{a.s.} X \text{ if } P\left(\left\{s \mid \lim_{n \rightarrow \infty} X_n(s) = X(s)\right\}\right) = 1$$

equivalently, if $P\left(\left\{s \mid \lim_{n \rightarrow \infty} X_n(s) \neq X(s)\right\}\right) = 0$

The set of divergence

$$D = \left\{s \mid \lim_{n \rightarrow \infty} X_n(s) \neq X(s)\right\}$$

$$= \left\{s \mid |X_n(s) - X(s)| \geq \varepsilon \text{ i.o. for all arbitrarily small } \varepsilon > 0\right\}$$

For each arbitrarily small $\varepsilon > 0$, we can find $m \in \mathbb{N}$ such that $\frac{1}{m} \leq \varepsilon$.

The set of divergence

$$D = \left\{s \mid |X_n(s) - X(s)| \geq \frac{1}{m} \text{ i.o. for all } m \in \mathbb{N}\right\} = \bigcup_{m=1}^{\infty} D_m$$

Now, how to test almost sure convergence, X_n converges almost sure to X , if probability of this event s such that limit of $X_n(s)$ equal to $X(s)$, the probability of this event is equal to 1 or equivalently, if this limit is not equal to $X(s)$, if we consider that event that probability should be equal to 0. Now let us define a set of divergence where $X_n(s)$ is not equal to s .

So, that event is we are calling this event as D . Now, that event is equal to, the set as such that mod of X_n minus s mod of $X_n(s)$ minus $X(s)$ is greater than equal to epsilon and now infinitely of 1 for all arbitrarily small epsilon greater than 0. So, what we are defining that the set of divergence is this set comprising of element such that mod of $X_n(s)$ minus $X(s)$ is greater than epsilon infinitely often for all arbitrarily small epsilon greater than 0.

So, if suppose there is only divergence in the case of finite number of n , then we can always consider the sequence starting with that number where it is diverging suppose after that the sequence we can consider. So, that way this is important that we define the event s such that mod of $X_n(s)$ minus $X(s)$ is greater than equal to epsilon for in for infinitely many n because if it is finitely many n then we can always go beyond those numbers to find out the convergence sequence.

Now, for is arbitrarily small epsilon we can find a positive integer m , m belonging to set of natural number, this is the set of natural number such that $1/m$ is less than equal to

epsilon. So, any epsilon, you consider we can get a, positive integer like this such that $\frac{1}{m}$ is less than equal to epsilon. So, that we can replace epsilon by $\frac{1}{m}$ what is the idea because if it is in terms of m we can enumerate.

So, this set of divergence now we can define as D is equal to the set comprising of element s such that $\text{mod of } X_n(s) - X(s)$ is greater than equal to $\frac{1}{m}$. Instead of epsilon we are writing greater than equal to $\frac{1}{m}$ infinitely often for all m will belonging to n , that is equal to suppose now m is we can count m therefore, we can write in terms of suppose Union of D_m m going from 1 to infinity because all m we have to consider. So, that way we have define a set of divergence where the random sequence diverges from $X(s)$. So, that we are considering and that is a union D_m because for each time we have to define that divergence set; so, union D_m m going from 1 to infinity.

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Condition for a.s. convergence

The set of divergence can be rewritten as

$$D = \left\{ s \mid |X_n(s) - X(s)| \geq \frac{1}{m} \text{ i.o. for all positive integer } m \right\}$$

$$= \bigcup_{m=1}^{\infty} D_m$$

$$D_m = \limsup_{n \rightarrow \infty} \left\{ s \mid |X_n(s) - X(s)| \geq \frac{1}{m} \right\}$$

$$= \bigcap_{n=1}^{\infty} \bigcup_{j=n}^{\infty} \left\{ s \mid |X_j(s) - X(s)| \geq \frac{1}{m} \right\}$$

$$\{X_n\} \xrightarrow{a.s.} X \text{ if and only if } P\left(\bigcup_{m=1}^{\infty} D_m\right) = 0$$

Now we get a condition for s convergence convergence in terms of D . Suppose this set of divergence can be rewritten as which we have already done that it is a Union of D_m m going from 1 to infinity. Now this infinitely often that is the, that is the lim sup of s such that $\text{mod of } X, X_n(s) - X(s)$ is greater than equal to $\frac{1}{m}$ as n tends to infinity, that is the event of this, that $X_n - X_n(s) - X(s)$ absolute value of that is greater than $\frac{1}{m}$ infinity often that we are writing in terms of lim, lim sup.

Therefore, now this is equal to as from definition of lim sup intersection n going from 1 to infinity of Union there going from n to infinity of the event s such that $\text{mod of } X$ s

minus X is greater than equal to $1/m$. So, this is the basic event where it is greater and this deviation is greater than $1/m$ then we are considering the lim sup of that event sequence.

Obviously, now X_n will converge to X almost sure if and only if probability of Union of D_m is equal to 0. So, we have the condition now if and only if condition that X_n will converge to X almost sure if and only if probability of union of D_m m going from 1 to infinity is equal to 0, this is a very strong condition.

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Theorem: $\{X_n\} \xrightarrow{a.s.} X$ if and only if $P(D_m) = 0$ for each positive integer m

Proof: Suppose $\{X_n\} \xrightarrow{a.s.} X$

$$\text{then } P\left(\bigcup_{m=1}^{\infty} D_m\right) = 0$$

$$\text{but } D_m \subseteq \bigcup_{i=1}^{\infty} D_i$$

$$\text{thus } P(D_m) \leq P\left(\bigcup_{i=1}^{\infty} D_i\right) = 0$$

Next, if $P(D_m) = 0 \forall m \geq 1$, then

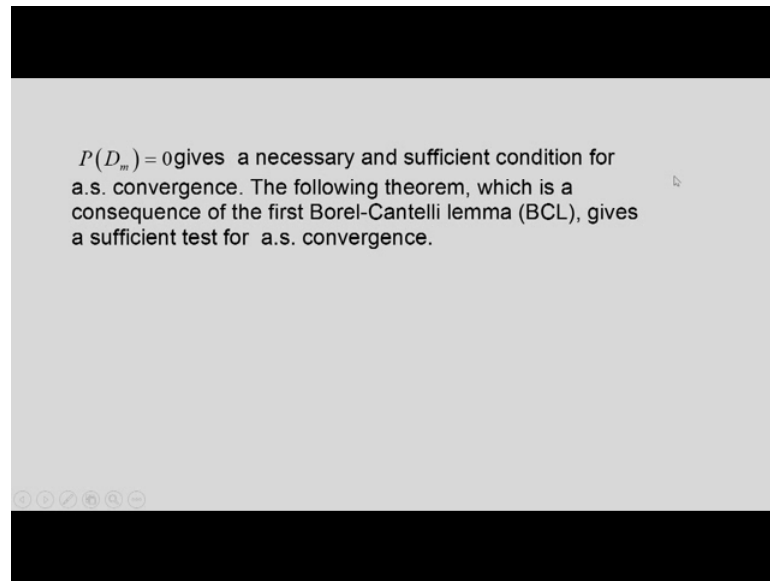
$$P\left(\bigcup_{m=1}^{\infty} D_m\right) \leq \sum_{m=1}^{\infty} P(D_m) = 0$$

Now, we will because here we have the union, union of D_m m going from 1 to infinity, this condition we want to simplify. We will state the theorem the sequence X_n converges almost sure to X if and only if probability of D_m is equal to 0 for each positive integer m . So, this is a simpler condition because we have to consider P of D_m instead of the probability of union D_m .

Now, as the proof suppose X_n converges to X , almost here then probability of union of D_m m going from 1 to infinity must be equal to 0 but we also know that this D_m , any D_m any member is a subset of the union, union D_i , i going from 1 to infinity. Therefore, we can write that probability of D_m because it is a subset it is a less than equal to probability of this union but, we know that probability of this union is equal to 0. Therefore, probability D_m must be equal to 0. So, that way if X_n converges almost sure to X , then probability of D_m must be equal to 0.

Now, next if probability of D_m is equal to 0 for all m greater than equal to 1 then probability of a union of D_m , I know that that is less than equal to sum of D corresponding probability, this inequality we proved earlier therefore, and I know that each of this probability is equal to 0 therefore, probability of this union will be equal to 0.

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So, that way we have proved that X_n converges to X if and only if probability of D_m is equal to 0 for each positive integer m so, this is a condition that is, if and only if condition probability of D_m is equal to 0 gives a necessary and sufficient condition for almost your convergence. So, we can find out this probability to prove almost your convergence but this is still a, difficult task because we have to consider D_m then D_m is, D_m for each m and we know that D_m is defined through that $\limsup$ operation.

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Theorem: If for each $m \geq 1$, $\sum_{n=1}^{\infty} P\left(|X_n - X| \geq \frac{1}{m}\right) < \infty$, then

$$\{X_n\} \xrightarrow{a.s.} X$$

Proof:

Suppose for each $m \geq 1$,

$$\sum_{n=1}^{\infty} P\left(|X_n - X| \geq \frac{1}{m}\right) < \infty$$

$$\Rightarrow P\left(\limsup_{n \rightarrow \infty} \left\{s \mid |X_n(s) - X| > \frac{1}{m}\right\}\right) = 0 \quad (\text{Using BCL 1})$$

$$\Rightarrow P(D_m) = 0$$

$$\Rightarrow \{X_n\} \xrightarrow{a.s.} X$$

Now, we will state a theorem that is a consequence of first Borel- Cantelli lemma that gives a sufficient condition for almost sure convergence that we will discuss now., theorem if for each m greater than equal to 1, now we are considering this sum probability of mod of X_n minus X greater than equal to $\frac{1}{m}$. So, this is for each n , we will consider and then sum up; so, therefore infinite sum of probability of mod of X_n greater than equal to $\frac{1}{m}$.

So, if this infinite sum, sum of the infinite series is less than infinity, then X_n converges almost sure to X . So, we have to consider the probability of deviation for each m and then sum up, if this sum is less than infinity then X_n converges almost sure to X . Suppose, for each m greater than equal to 1 this sum is convergent. So, what does it mean? Now, we can apply the Borel - Cantelli lemma if the sequence of event. suppose I can consider this as a sequence of event, if the probability of this sequence if the sum of the probability of this sequence is less than infinity, then a probability of lim sup of that event sequence is equal to 0, that is the Borel Cantelli lemma 1.

So, what we will get here this implies that probability that lim sup of, the event s such that mod of $X_n(s)$ minus X , that is greater than $\frac{1}{m}$ as n tends to infinity is equal to 0, that is the conclusion. We draw with the help of Borel-Cantelli lemma 1. Therefore, but this is by definition this is the D_m therefore, probability of D_m is equal to 0 the probability of D_m is equal to 0, then X_n converges almost sure to X .

So, that way we can apply the Borel- Cantelli lemma 1 and that is first part of Borel -Cantelli lemma to prove the, prove the convergence of a sequence and thus we can apply this condition that is probability then mod of X_n minus X greater than equal to $\frac{1}{m}$ by m , that is the sum from n going from infinity, if that sum is convergent then X_n converges to X almost sure but this is your sufficient condition unlike probability of D_m is equal to 0 that is a necessary and sufficient condition but this is an this is a sufficient condition only.

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Example: Suppose the random sequence $\{X_n\}$ is given by

$$X_n = \begin{cases} 0 & \text{with probability } 1 - \frac{1}{2^n} \\ n & \text{with probability } \frac{1}{2^n} \end{cases}$$

Then,

$$P\left(\left\{|X_n - 0| \geq \frac{1}{m}\right\}\right) = P(X_n = n) = \frac{1}{2^n},$$

$$\sum_{n=1}^{\infty} P\left(\left\{|X_n - 0| \geq \frac{1}{m}\right\}\right) = \sum_{n=1}^{\infty} \frac{1}{2^n} < \infty$$

We shall consider one example suppose the random sequence X_n is given by X_n is equal to 0 with probability $1 - \frac{1}{2^n}$ and equal to n with probability $\frac{1}{2^n}$. So, we see that, this, this probability will become 1 as n tends to infinity and correspondingly this probability will become 0. So, therefore, X_n approaches 0.

Now, whether this is the limit we get in the almost your sense let us see. So,, we see that probability of $|X_n - 0| \geq \frac{1}{m}$ that is equal to probability of X_n is equal to n because for n only it is this quantity is becoming different from 0 here it is exactly 0. Therefore, only n we have to consider probability X_n is equal to n that is equal to $\frac{1}{2^n}$.

So, we have to consider this sum now, sum of these probabilities. So, if we consider sum up from n is equal to 1 to infinity that is equal to sum of $\frac{1}{2^n}$ as n going

from 1 to infinity and we know that this sum is less than infinity this is a geometric series we can find out this sum. So, this sum is less than infinity. Therefore, this sequence X_n converges almost sure to X . So, that way, we can prove whether a sequence convergence to a random variable X in the almost sure sense.

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To Summarise

- $\{X_n\}_{n=1}^{\infty}$ represents a family of sequences.
- Convergence of a random sequence is to be defined using different criteria.
- $\{X_n\}_{n=1}^{\infty}$ is said to converge everywhere to X if

$$\lim_{n \rightarrow \infty} X_n(s) = X(s) \quad \forall s \in S.$$
- $\{X_n\}_{n=1}^{\infty}$ is said to converge to X almost sure or with probability 1 if

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

Let us summarize the lecture, recall that this X_n this is sequence of random variable, it represents a family of sequence. So, unlike in the case of sequence of real numbers, the sequence of real random variables represents a family of sequence. Convergence of a random sequence is to be defined using different criteria the sequence is said to converge everywhere to X or point wise point, point wise converge to X if limit of $X_n(s)$ is equal to $X(s)$ for all s belonging to S , that is the convergence everywhere.

Now this sequence X_n n going from 1 to infinity is said to converge to X almost sure or with probability 1 if probability of this event that is S such that limit of $X_n(s)$ equal to $X(s)$ that is equal to 1. So, this is the, definition of almost sure convergence which we have introduced here and this is a strong mode of convergence.

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To Summarise..

- For testing a.s. convergence, we consider

$$D_m = \limsup_{n \rightarrow \infty} \left\{ s : |X_j(s) - X(s)| \geq \frac{1}{m} \right\} = \bigcap_{n=1}^{\infty} \bigcup_{j=n}^{\infty} \left\{ s : |X_j(s) - X(s)| \geq \frac{1}{m} \right\}$$

- $\{X_n\} \xrightarrow{a.s.} X$ if and only if $P\left(\bigcup_{m=1}^{\infty} D_m\right) = 0$

The test is simplified: $\{X_n\} \xrightarrow{a.s.} X$ if and only if $P(D_m) = 0$ for each positive integer m .

- Using First Borel-Cantelli lemma, a sufficient test for a.s. convergence

- If for each $m \geq 1$, $\sum_{n=1}^{\infty} P\left(|X_n - X| \geq \frac{1}{m}\right) < \infty$, then $\{X_n\} \xrightarrow{a.s.} X$

THANK YOU

And then, for testing almost sure convergence we consider this event D_m , what is the D_m $\limsup$ of the event s such that $|X_j(s) - X(s)| \geq \frac{1}{m}$. So, $\limsup$ of this event and this is given by this definition of $\limsup$ only we are considering here. Now, we have a necessary and sufficient condition X_n converges almost 0 to X if and only if probability of union of D_m m going from 1 to infinity is equal to 0. The test is simplified that is X_n converges almost sure to X if and only if probability of D_m is equal to 0 for each positive integer m .

Now, using the first Borel - Cantelli lemma, we got a sufficient contest for almost sure convergence if for each m greater than 1, this sum n is equal to 1 to infinity probability of mod of X_n minus X greater than $\frac{1}{m}$. So, the, if the sum is less than infinity, then X_n sequence X converges almost 0 to X .

Thank you.