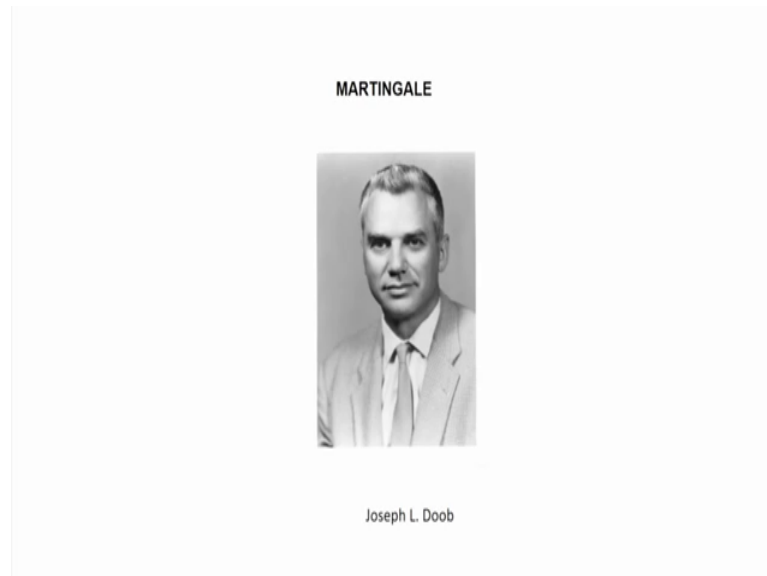


Advanced Topics in Probability and Random Processes
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Lecture – 24
Martingale Process-2

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MARTINGALE

Recall that a discrete-time random process $\{X_n, n \geq 0\}$ is called a martingale process if for all $n \geq 0$,

- (i) $E|X_n| < \infty$, and
- (ii) $E(X_{n+1} / X_0, X_1, \dots, X_n) = X_n$

If the equality sign in (ii) above is replaced by $\leq$, then $\{X_n, n \geq 0\}$ is called a *supermartingale* and if it is replaced by $\geq$, then $\{X_n, n \geq 0\}$ is a *submartingale*.

Recall also that a martingale has constant mean.

$E X^{(n)} = \text{constant} \forall n$

In the last lecture we introduced martingale. Let us recall the definition of martingale or a discrete time random process $X_n, n \geq 0$ is called a martingale

process. If for all n greater than equal to 0 E of X_n is less than infinity. So that means, average value of the magnitude of X_n should be finite. And second thing is conditional expectation of X and plus 1 given X_0, X_1 up to X_n is equal to X_n . Also we define supermartingale and submartingale. In the case of supermartingale this equal this sign is replaced by less than equal to inequality.

Similarly, in the case of submartingale, this equality sign is replaced by greater than equal to inequality. So, that you know that last class we have just told that it is less than inequality, but it should be less than equal to inequality. Similarly, here also it is greater than equal to inequality. Also proved very important property of martingale that, a martingale has a constant mean; that means, E of X_n is equal to constant for all n . So, this is an important property; that means, martingale process is stationary in mean. So, this property is used in deriving 2 important inequalities, other inequalities are also there, but we will not discuss those inequalities, but to basic inequalities we will discuss.

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Martingale Inequalities
Maximal Inequality- If $\{X_n, n \geq 0\}$ is a non-negative martingale,
then $P(\max_{n \geq 0} (X_n) \geq a) \leq \frac{EX_0}{a}$
Proof Using Markov inequality

$$P(X_n \geq a) \leq \frac{EX_n}{a}, n \geq 0$$

$$= \frac{EX_0}{a}$$

$$\therefore P(\max_{n \geq 0} (X_n) \geq a) \leq \frac{EX_0}{a}$$

X_n nonnegative RV

a) EX_n

$P(X_n \geq a) \leq \frac{EX_n}{a}$

Martingale inequalities, maximal inequality if X_n, n greater than equal to 0 is a non negative martingale; then, probability of maximum of X_n, n greater than equal to 0, greater than equal to a is less than equal to E of X_0 divided by a .

So, X_n is a non negative martingale that is important and probability of maximum of X_n that, X that is greater than equal to a is bounded by E of X_0 divided by a . So, this is a statement of Markov inequality in the case of martingale process. Here we recall that

Markov inequality stay there if X_n is a nonnegative random variable and a is a number greater than E of X_n then, the probability that X_n greater than equal to a is bounded by E of X_n divided by a . So, that way we get this result probability that X_n greater than equal to a is bounded by E of X_n divided by a for all n greater than equal to 0.

Now, we know that E of X_n is the equal to E of x_0 that is martingale is a constant mean process. So we can write that probability of X_n greater than equal to a is less than equal to E of X_n divided by a . Now this is true for all n greater than equal to 0 therefore, it will be true for maximum of action n greater than equal to 0. Therefore, we can write probability of maximum of action and greater than equal to 0, greater than equal to a is bounded by E of X_n divided by a . So, this is known as the maximal inequality.

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Kolmogorov Inequality
 Suppose $X_0 = 0$. Clearly, $EX_n = 0, n = 0, 1, \dots$
 Then by applying Chebyshev inequality

$$P(|X_n| \geq a) \leq \frac{EX_n^2}{a^2}, n = 0, 1, \dots, n$$

$EX_n^2 = \sigma_n^2 = \text{variance}$

Another inequality is Kolmogorov inequality. Suppose, X_0 is equal to 0 clearly E of X_n is equal to 0 because X_0 is equal to 0 then E of X_n is equal to 0. Therefore, E of X_n square will be the variance. So, E of X_n square is equal to sigma square is the variance; sigma square is the variance. Why, because X_0 is equal to 0 because of that E of X_0 is equal to 0 and E of X_n is equal to 0 for all n . Then, by applying Chebyshev inequality now, probability of mod of X_n greater than equal to a is less than equal to now, sigma is here in this case E of X_n square divided by a square for n is equal to 0 etcetera.

So, this inequality is known as Kolmogorov inequality; sometimes these inequalities are helpful in deriving some important bounds.

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$$\begin{aligned} & \text{one-step prediction} \\ & E(X_{n+1} | X_0, X_1, \dots, X_n) \\ & = X_n \end{aligned}$$

So far we discussed one step prediction. So, what we discussed that E of X_{n+1} given X_0, X_1 up to X_n is equal to X_n . This is the martingale property.

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m-step prediction

We have observed that for a martingale process $\{X_n, n \geq 0\}$,

$$E(X_{n+1} | X_0, X_1, \dots, X_n) = X_n \text{ (one-step prediction)}$$

The following theorem says about the m-step prediction

Theorem: For a Martingale process $\{X_n\}$,

$$E(X_{n+m} | X_0, X_1, \dots, X_n) = X_n$$

Proof Recall the property of the conditional expectation:

$$E(EY | X, Z) | X = EY | X$$

Therefore,

$$\begin{aligned} LHS &= E\{E(X_{n+m} | X_0, X_1, \dots, X_n, \dots, X_{n+m-1}) | X_0, X_1, \dots, X_n\} \\ &= E(X_{n+m-1} | X_0, X_1, \dots, X_n) \\ &\text{Repeating this we get} \\ &= E(X_{n+1} | X_0, X_1, \dots, X_n) \\ &= X_n \end{aligned}$$

Now, let us see what happened to m step prediction. Suppose, we want to predict beyond X_{n+1} we have one important result. So for a martingale process X_n expected value of X_{n+m} given X_0, X_1 up to X_n is equal to X_n for a martingale process X_n . So, m step prediction is also equal to X_n . So, to prove this, let us apply this theorem to this important result here. So, left hand side will be expected value of E of X_{n+m}

1 given X_0, X_1 up to X_n , then we introduced the term X_{n+1} up to X_{n+M-1} plus 1. So, like that here we are introducing new terms given X_0, X_1 up to X_n .

Because, we are interested to find out this quantity therefore, we are again taking the conditional expectation with respect to X_0, X_1 up to X_n . This is same as E of X_{n+M-1} plus m minus one given X_0, X_1 up to X_n .

So, left hand side, this is the quantity. So, we introduce additional members here and now we introduce the martingale property. So, if I have to predict this conditional expectation of X_{n+M} giving X_{n+M-1} then, this will be same as X_{n+M-1} given X_0, X_1 up to X_n .

So, that way what we have shown that, this quantity, conditional expectation of X_{n+M} plus m given X_0, X_1 up to X_n is equal to conditional expectation of X_{n+M-1} plus M minus 1 given X_0, X_1 up to X_n . Again we can introduce term up to suppose X_{n+M-1} plus 1, X_{n+M-2} etcetera up to X_{n+M-2} . Then we can so apply this theorem and we can show that this expression is same as E of X_{n+M-2} plus m minus 2 given X_0, X_1 up to X_n .

Repeating in this manner we can show that this is equal to ultimately, it will come down to E of X_{n+1} given X_0, X_1 up to X_n and that is equal to X_n . So, that way we have proved the remarkable property that m step prediction for the martingale process is also equal to X_n and this result gives some additional important properties of the martingale process that we will be discussing next.

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Corollary 1 For a Martingale process $\{X_n\}$,

$$E(X_n X_{n+m}) = EX_n^2, m \geq 0$$

$$EEX/Y = EX$$

Proof

$$\begin{aligned} E(X_n X_{n+m}) &= EE(X_n X_{n+m} / X_0, X_1, \dots, X_n) \\ &= EX_n E(X_{n+m} / X_0, X_1, \dots, X_n) \\ &= EX_n X_n \\ &= EX_n^2 \end{aligned}$$

First we will prove that for a martingale process E of X_n into X_{n+m} is equal to E of X_n squared. So, now we use the again property of conditional expectation that E of X given Y is equal to E of X . So, this is the property. This property also we proved therefore, here we can write that this is the E of X_n into X_{n+m} is equal to expectation of the conditional expectation of X_n into X_{n+m} given X_0, X_1 up to X_n . So, first we will take the conditional expectation then, since it will be a function of these quantities, we will take expectation again then we will get the expectation of X_n into X_{n+m} .

Now, this is conditional expectation with respect to X_n . Therefore, this term will result X_n itself, given X_n this expectation will be X_n only. So, we will have only E of X_{n+m} given X_0, X_1 up to X_n . So, that way, what we will have now? Now this quantity again I know that this is equal to X_n . Therefore, we will have E of X_n into X_n that is equal to E of X_n squared.

So, what we have is that, this is in fact, autocorrelation function E' of X_n into X_{n+m} is equal to E of X_n squared. So, this is one important property of the martingale process.

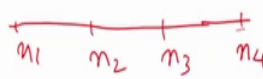
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Corollary 2 A martingale $\{X_n\}$ is an orthogonal increment process, i.e. for $n_1 < n_2 < n_3 < n_4$ $E(X_{n_2} - X_{n_1})(X_{n_4} - X_{n_3}) = 0$

Proof

$$\begin{aligned}
 E(X_{n_2} - X_{n_1})(X_{n_4} - X_{n_3}) &= EX_{n_2}X_{n_4} - EX_{n_2}X_{n_3} - EX_{n_1}X_{n_4} + EX_{n_1}X_{n_3} \\
 &= EX_{n_2}^2 - EX_{n_2}^2 - EX_{n_1}^2 + EX_{n_1}^2 \\
 &= 0
 \end{aligned}$$

X and Y are orthogonal if $EXY=0$



Next property is a martingale X_n is an orthogonal increment process. So, what is orthogonal process? Suppose, first of all orthogonal random variable, X and Y are orthogonal, X and Y are orthogonal, if E of XY is equal to 0. Now, martingale is an orthogonal increment process. What does it say that suppose, we have some this is n_1 n_2 n_3 this is somewhere n_3 suppose this is somewhere n_4 , what we have is that this interval and this interval are non overlapping interval. And if you considered increments in non overlapping interval that is X of n_2 minus X of n_1 , that is the increment X of n_2 minus X of n_1 and similarly, increment here is X of n_4 minus X of n_3 . These intervals are non overlapping therefore, the expectation of the increments X of n_2 minus X of n_1 into X of n_4 minus X of n_3 that must be equal to 0. This also can be easily proved.

Now, how we can prove, we will use the earlier result that is equal to suppose, first I will multiplied by X of n_2 into X of n_4 similarly, minus E of X of n_2 into X of n_3 . Now, if I consider this E of X of n_1 into X of n_4 plus E of X of n_1 into X of n_3 . So, that way this we expand, now we know what is E of X of n_1 into X of n_4 because, here n_4 is greater than n_2 therefore, it will be E of X of n_2 squared. Similarly, this quantity also will be E of X of n_2 squared so; this and this will get cancel. This quantity will be E of X of n_1 squared and this is plus E of X of n_1 squared.

So, what we will have is that, this is equal to 0. So, what we have got is that a martingale process is an orthogonal increment process. Earlier we discussed independent increment

process for example, Poisson process is an independent increment process, but a martingale process is an orthogonal increment process.

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Corollary 3 For a martingale process $\{X_n\}$, EX_n^2 is a monotonically increasing sequence.

Proof We have

$$0 \leq E(X_{n+1} - X_n)^2$$

$$= EX_{n+1}^2 + EX_n^2 - 2EX_n X_{n+1}$$

$$= EX_{n+1}^2 + EX_n^2 - 2EX_n^2$$

$$= EX_{n+1}^2 - EX_n^2$$

$\therefore EX_n^2$ is a monotonically increasing sequence.

Handwritten notes:
 $EX_0^2 \leq EX_1^2 \leq EX_2^2 \dots$
 $EX_{n+1}^2 - EX_n^2 \geq 0$
 $\Rightarrow EX_{n+1}^2 \geq EX_n^2$

We have another corollary, this is also important for a martingale process X_n , E of X_n square that is mean square value is a monotonically increasing sequence. So, E of X_n square it is a deterministic quantity. Now sequence of E of X_n square is a monotonically increasing quantity; that means, E of X_0 square is less than equal to E of X_1 squared is less than equal to E of X_2 square like that.

So, to prove this let us consider this quantity, what happen? The difference between E of X of n plus 1 and X_n . So, what we get? 0 is less than equal to E of X of n plus 1 minus X_n whole square because, whole square we are taking. Therefore, this is a positive quantity, so 0 is less than equal to this. Now, we if we expand it, this will be E of X n plus 1 square E of X plus E of X n square minus twice E of X n into X of n plus 1.

Now, we will apply the property of the martingale that we have derived earlier that is E of X n into X of n plus 1 is equal to E of X n square. So, that way E of X n plus 1 square plus E of X n square minus twice E of X n squared. Now, this minus 2 n plus 1 will get minus E of X n square. So, what we have established that, this quantity E of X n plus 1 square minus E of X n square is greater than equal to 0 because, 0 is less than equal to here and this will imply that E of X n plus 1 squared is greater than equal to E of X n

square. Therefore, what it says that E of X_n square is a monotonically increasing sequence.

So, this is important that, the mean squared value away martingale process is monotonically increasing with respect to n . Through mean square value away a martingale process is monotonically increasing with respect to n .

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Martingale convergence theorem
Let $\{X_n, n \geq 0\}$ be a martingale and $EX_n^2 \leq M < \infty$ for all n . Then $\{X_n\}$ converges in the m.s. sense as $n \rightarrow \infty$ to a random variable X .
Proof:

$$E(X_{n+m} - X_n)^2$$

$$= EX_{n+m}^2 + EX_n^2 - 2EX_n X_{n+m}$$

$$= EX_{n+m}^2 + EX_n^2 - 2EX_n^2$$

$$= EX_{n+m}^2 - EX_n^2$$

$$EX_n^2$$
 is a bounded monotonically increasing sequence and hence convergent

$$\therefore \lim_{n \rightarrow \infty} E(X_{n+m} - X_n)^2 = 0$$

$$\Rightarrow \{X_n\}$$
 is converges in M.S. to some RV X .

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

Handwritten notes:
 $E(X_n - X)^2 \xrightarrow{a.s.} 0$
 $E(X_{n+m} - X_n)^2 \xrightarrow{a.s.} 0$

Now, we will come to the martingale convergence theorem, very important theorem, it is important in establishing many results in various areas science and engineering. Let X_n n greater than or equal to 0 be a martingale and E of X_n square is bounded, so it is a less than equal to m where m is less than infinity, m is less than infinity for all n . Then X_n converges in the mean square sense as n to infinity to a random variable x .

So, X_n converges in mean square sense to a random variable X . So, what we want to essentially we want to prove that, E of X_n minus X whole square that converges to 0 as n tends to infinity, as n tends to infinity. So, to prove this, we will use the quasi criterion for mean square convergence. So, what it says that, if you can show that E of X_{n+m} minus X_n whole squared. So, if this quantity goes down to 0, as n tends to infinity and m greater than 0 for all m greater than 0 then, this sequence X_n will be mean square convergence. This is a if and only if criterion. So, here without knowing the random variable X we can prove that, this sequence is convergence; this is the quasi convergence criterion for mean square convergence. We will prove this, E of X_{n+m} minus X_n

whole square is equal to E of X of n plus m squared plus E of X n squared minus twice E of X n into X of n plus m and that is equal to E of X n plus m square plus E of X n square minus.

Now, this quantity is E of X n square so, that way we have E of X n X of n plus m square minus E of X n square. Now is of this quantity is a bounded quantity and monotonically increasing quantity. Because, bounded because of this and we know that E of X n square is a monotonically increasing sequence. Therefore, both will converge to some limit and because of that this quantity will become 0 as n tends to infinity, which implied that E of x n plus m minus X n whole square is equal to 0. Therefore, X n converges in mean square sense to sum random variable X random variable X .

So, this is a very powerful result. In fact, it can be shown that a martingales bounded martingale sequence, that is if E of n square is bounded. Then a martingale sequence convergence not only in mean square sense, but it converges almost here also. So that means, this X n converges almost shear to some random variable X , but we will not prove this, but this is a very powerful result. And therefore, if it is a martingale sequence and E of X n square is bounded by m . Then, this martingale sequence will converge in almost shear sense in mean square sense in probability sense to a random variable X . So, this is the martingale convergence theorem.

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To summarise

- A discrete-time random process $\{X_n, n \geq 0\}$ is called martingale process if for all $n \geq 0$
 - $E|X_n| < \infty$, and
 - $E(X_{n+1} / X_0, X_1, \dots, X_n) = X_n$.

If the equality sign in (ii) above is replaced by $\leq$, then $\{X_n, n \geq 0\}$ is called a *supermartingale* and if it is replaced by $\geq$, then $\{X_n, n \geq 0\}$ is a *submartingale*.

- A martingale has constant mean
For a Martingale process $\{X_n\}$,
 $E(X_{n+1} / X_0, X_1, \dots, X_n) = X_n$

Handwritten notes:

Maximal inequality and Kolmogorov Inequality

$X_0 = 0$
 $E X_n = 0 \forall n$
 $P(|X_n| > a) \leq \frac{E X_n^2}{a^2}$

$P(X_n \geq a) \leq \frac{E X_0}{a} \quad \forall n$
 $P(\max_{n \geq 0} (X_n) \geq a) \leq \frac{E X_0}{a}$

Let us summarize the lecture, a discrete time random process X_n , $n \geq 0$ is called a martingale process, if for all $n \geq 0$, $E[X_{n+1} | \mathcal{F}_n]$ is finite and conditional expectation of X_{n+1} given $X_0, X_1, \dots, X_n$ is equal to X_n itself is the present value itself. And also we remembered at if we replace this equality by less than equal to inequality then, X_n is a supermartingale and if it is replaced by greater than or equal to inequality, then X_n will be a submartingale.

So, we also find the supermartingale and submartingale. And a martingale has a constant mean. So, that was proved in the earlier class, but because of that we could establish 2 important inequalities that is that maximal inequality and Kolmogorov inequality. So, what does it say, probability that X_n if X_n is a non negative process, probability of $X_n \geq a$ is less than equal to $E[X_0] / a$. This is for all n therefore, we take the maximum of this quantity, so that way probability of maximum of X_n over $n \geq 0$, greater than equal to a is less than equal to $E[X_0] / a$ and the Kolmogorov inequality was for martingale which X_0 is equal to 0.

So, Kolmogorov inequality was for a martingale process, which X_0 is equal to 0 and because of that, $E[X_n] = 0$ for all n and therefore, probability of maximum of X_n greater than equal to a . So, this probability is less than equal to $E[X_n^2] / a^2$. So, this is the Kolmogorov inequality. First, we derived the maximal inequality which is an extension of Markov inequality and the Kolmogorov inequality is a special case of Chebyshev inequality.

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To summarise...

- For a Martingale process $\{X_n\}$, $E(X_n X_{n+m}) = EX_n^2$, $m \geq 0$
- A martingale $\{X_n\}$ is an orthogonal increment process, i.e. for

$$n_1 < n_2 < n_3 < n_4$$

$$E(X_{n_2} - X_{n_1})(X_{n_4} - X_{n_3}) = 0$$

- For a martingale process $\{X_n\}$, EX_n^2 is a monotonically increasing sequence.
- Let $\{X_n, n \geq 0\}$ be a martingale and $EX_n^2 \leq M < \infty \forall n$. Then $\{X_n\}$ converges in the m.s. sense as $n \rightarrow \infty$ to a random variable X .

THANK YOU

Ah then we establish one important result basically, first we consider a m step prediction, so this is another important result for a martingale process X_n E of X of n plus m given $X_0 X_1$ up to X_n is equal to X_n that is m step prediction is also equal to present value.

So, this result resulted into several important results, first of all E of X_n into X of n plus m for m greater than equal to 0 is equal to E of X_n square. Then, we establish the that martingale is an orthogonal increment process, that is for n_1 less than n_2 less than n_3 less than n_4 E of X of n_2 minus X of n_1 , that is the increment, X of n_4 minus X of n_3 , that is another increment. So, joint expectation of these increments is equal to 0.

So, that way a martingale process is an orthogonal increment process. For a martingale process X_n E of X , X_n square is a monotonically increasing sequence. So, this using this result we proved a very important result that is a martingale sequence for which E of X_n square is less than equal to m less than infinity then X_n converges in mean square sense to a random variable X . So, a martingale sequence with mean square value bounded by some number m converges in ms sense to a random variable X . We also told that this convergence is also in the almost shear sense, but that we did not prove it, but this is a very powerful result.

Thank you.