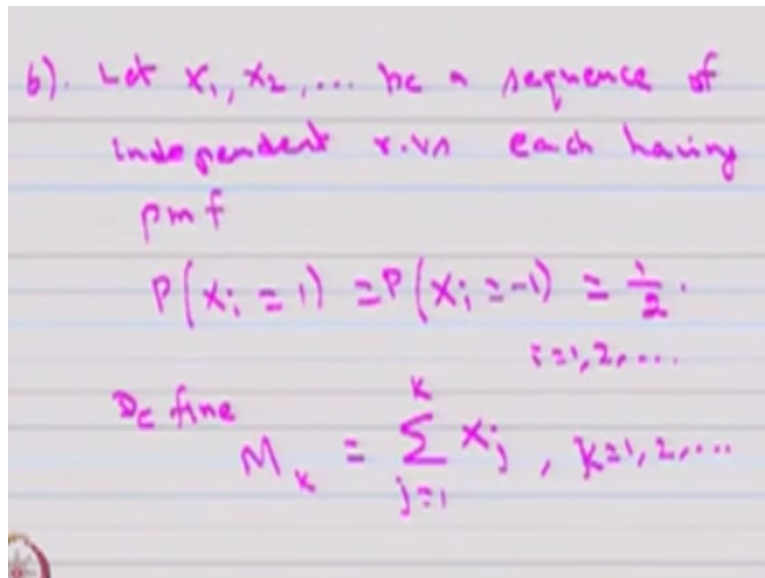


**Stochastic Processes-1**  
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**Lecture – 16**  
**Problems in Sequence of Random Variables (Contd...)**

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Now we move into next example, example 6, Let  $x_1, x_2, \dots$  be a sequence of independent random variables each having probability mass function, probability of  $x_i$  is equal to 1 that is same as probability of  $x_i$  takes the value minus 1, the probability is 1 by 2. This is valid for that means it is a sequence of iid random variables and they are discrete type. Define  $M$  suffix  $k$  as the sum of first  $k$   $x_i$  random variables.

So this running index is  $k$  is equal to 1, 2 and so on. So we are defining a sequence of a random variable  $M_k$  by summing first  $k$   $x_i$  random variables.

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For a fixed integer  $n$ , define

$$W^{(n)}(t) = \frac{1}{\sqrt{n}} M_{nt}$$

for all  $t \geq 0$  such that  $nt$  is an integer

Now, for  $0 \leq s \leq t$

$$E(W^{(n)}(t) - W^{(n)}(s)) = 0$$

$$\text{Var}(W^{(n)}(t) - W^{(n)}(s)) = t - s$$

For a fixed integer  $n$ , we define another sequence of random variable that is denoted by  $W$  superscript  $n$  of  $t$  that is nothing but 1 divided by square root of  $n$   $M$  suffix  $n$  times  $t$ . This is for all  $t$  greater than or equal to zero such that  $n$  times  $t$  is an integer. So we are defining another sequence of a random variables  $W$  superscript  $n$  of  $t$  that is 1 divided by square root of  $n$  times  $m_n$  of  $t$  where  $n$  of  $t$  is an integer.

So this is valid for all  $t$  greater than or equal to zero. If you find out the mean and variants for the difference of the random variable of a  $n$  of  $t$  minus  $W$   $n$  of  $s$  for zero less than or equal to  $s$  or less than or equal to  $t$ , this quantity will be zero. That means  $W$   $n$  of  $t$  is a 1 divided by square root of  $n$   $m_n$  of  $t$  and the way we define the  $m_n$  of  $t$  that is the summation of  $x_i$  and the probability of  $x_i$  is equal to 1 and the probability of  $x$  is equal to minus 1, minus 1 is 1 by 2.

Therefore, the mean of  $x_i$  are going to be zero because of that the expectation of or mean of  $W$   $n$  of  $t$  minus  $W$   $n$  of  $s$  that is equal to zero. Also if you evaluate the variants of  $W$   $n$  of  $t$  minus  $W$   $n$  of  $s$  by finding first variants of excise using that you find out the variants of  $m_n$  of  $t$  then find out the variants of  $W$   $n$  of  $t$  minus  $W$   $n$  of  $s$  that is going to be  $t$  minus  $s$ .

It need a calculation of expectation of a  $x_i$  square then using expectation of a  $x_i$  square and the expectation of  $x_i$ , you can find out the variants of  $x_i$  using variants of  $x_i$ , you can find out the variants of  $W$   $n$  of  $t$ , then you find out the variants of  $W$   $n$  of  $t$  minus  $W$   $n$  of  $s$ .

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Now, for  $0 \leq s \leq t$

$$E(W^{(n)}(t) - W^{(n)}(s)) = 0$$

$$\text{Var}(W^{(n)}(t) - W^{(n)}(s)) = t - s$$

Fix  $t \geq 0$ ,

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as  $n \rightarrow \infty$

$$W^{(n)}(t) \xrightarrow{d} X$$

where  $X \sim N(0, t)$

By using mean and variants, for fixing  $t$  greater than or equal to zero as  $n$  tends to infinity we can conclude  $W_n$  of  $t$  tends to a random variable  $x$  and this converges takes place in distribution using CLT, one can conclude  $W_n$  of  $t$  converges to the random variable  $x$  and the converges in distribution where  $x$  is normal distribution with the mean zero and the variants  $t$ .

Using a central limit theorem, one can prove,  $W_n$  of  $t$  converges to  $x$  in distribution where  $x$  is normal distribution with a mean zero and the variants  $t$ . This result is very useful in Brownian motion and this same problem will be discussed in detail when we are discussing the module of a Brownian motion.