

3/

pdf  $X_i$  is iid  
 $f(x) = \frac{1}{2} e^{-|x|}$

$$S_n = \sum_{i=1}^n X_i$$

$$X \rightarrow \mu = 0$$

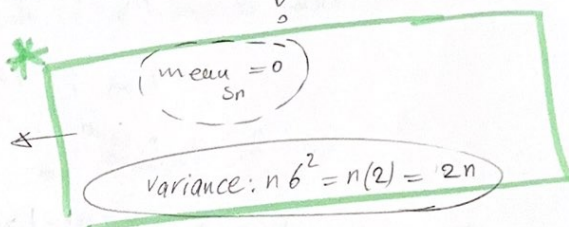
$$\sigma^2 = 2$$

$$\lim_{n \rightarrow \infty} P(S_n^2 > n) = ?$$

$X_i \rightarrow \text{iid} \Rightarrow$  Central Limit theorem

$\Rightarrow \lim_{n \rightarrow \infty} S_n \rightarrow$  the distribution of  $S_n$  will approach a normal distribution with a mean  $= n\mu = 0 \rightarrow$

$$\frac{S_n}{\sqrt{2n}} \xrightarrow[n \rightarrow \infty]{} N(0, 1)$$



$\Rightarrow S_n \rightarrow$  normal distributed  $N(0, n)$

$$\Rightarrow \lim_{n \rightarrow \infty} P(S_n^2 > n) \leftrightarrow P(|S_n| > \sqrt{n})$$

$\Rightarrow P(|S_n| > \sqrt{n}) \rightarrow 0$   
 as: The probability that a standard normal variable is outside of the

$S_n$  is normally distributed as:  $N(0, n)$   $\left[ -\sqrt{\frac{n}{2n}}, \sqrt{\frac{n}{2n}} \right]$  is  $\frac{0}{\sqrt{2n}}$  as  $n \rightarrow \infty$

$\Rightarrow \frac{S_n}{\sqrt{2n}}$  will be a standard normal random variable

as  $n \rightarrow \infty$

$\lim_{n \rightarrow \infty} P(S_n^2 > n)$ : The probability that a standard normal random variable is greater than  $\frac{1}{\sqrt{2}}$  or less than  $-\frac{1}{\sqrt{2}}$ .  $P(Z > 1) + P(Z < -1)$

standard normal r.v.

$P(Z > 1) = 1 - P(Z \leq 1)$ . as the standard normal distribution is symmetric,  $P(Z < -1) = P(Z > 1) \Rightarrow P(|Z| > 1) = 2 P(Z > 1)$

3)

$\Rightarrow$  python code:

$$P(|Z| > 1) = 1 - \underbrace{P(Z \leq 1)}_{\text{norm. cdf}(1)} \rightarrow = 0.317$$

Also:

$$\text{markov} \rightarrow \underbrace{P(|s_n^2| \geq n)}_{\text{non-negative}} \leq \frac{E[s_n^2]}{n} = 0 \quad n \geq 0$$

$$P(|s_n^2| \geq n) = 0 \Rightarrow$$

$$\lim_{n \rightarrow \infty} P(|s_n^2| \geq n) = 0$$

$$\text{chebyshev} \rightarrow P(|s_n| > \sqrt{n}) \leq \frac{1}{n} \Rightarrow$$

$$\lim_{n \rightarrow \infty} P(s_n^2 > n) \leq \lim_{n \rightarrow \infty} \left( \frac{1}{n} + \frac{1}{n} \right) = \lim_{n \rightarrow \infty} \left( \frac{2}{n} \right) = 0$$

4)

$$V_j = |\{n \geq 1 : X_n = j\}|$$

↓

The number of visits of Markov chain  $X_n$  to  $j$ .

$$\nu_{ij} = P(V_j = \infty \mid X_0 = i)$$

q)  $\nu_{ii} = ?$

A state "i" is persistent (or recurrent) if, once visited, the probability of returning to it infinitely often is 1.  $\rightarrow E(V_i \mid X_0 = i)$  is infinite mathematically.

and it is transient if the total expected number of visits is finite.

if "i" is persistent  $\Rightarrow$  if start from state "i" the probability that we will return to state "i" infinitely often is "1".  $\Rightarrow \nu_{ii} = 1$ : The probability that we will return to state "i" at least once is "1". (since the state is recurrent) and due to the Markov property, each time we return to state "i", the chain "restarts", and we have the same probability of returning to it again. since it can happen an infinite number of times,  $\nu_{ii} = 1$



(4)

if  $i$  is transient;

the expected number of visits is finite, which means there is a probability of not returning to state  $i$  that is greater than 0.

As we keep visiting state  $i$ , the probability of not returning to it again increases until it becomes certain that we won't return.

In this case, the probability that we will never return to state  $i$  after some finite number  $n$  of visits is  $> 0$ , which implies that the probability of returning to state  $i$  infinitely often is less than 1. Since the state is transient, this probability is actually 0,  $\Rightarrow f_{ii} = 0$

$$\Rightarrow f_{ii} = \begin{cases} 1 & \text{if } i \text{ is persistent} \\ 0 & \text{if } i \text{ is transient} \end{cases}$$

(13)

(4)

$$b) \nu_{ij} = ? \Rightarrow \nu_{ij} = \begin{cases} P(T_j < \infty | X_0 = i) & \text{if } j \text{ is persistent} \\ 0 & \text{if } j \text{ is transient} \end{cases}$$

$\nu_{ij} \rightarrow$  The probability that starting from state " $i$ " the Markov chain will visit state " $j$ " at least once.

$$T_j = \min \{ n \geq 1 | X_n = j \}$$

is the hitting time for state " $j$ "

if  $j$  is persistent (recurrent)

$\Rightarrow$  probability that the Markov chain, starting from any state  $i$ , will eventually hit state " $j$ " is 1.

$$\Rightarrow P(T_j < \infty | X_0 = i) = 1 \text{ for all } i.$$

Since the state is persistent, it will be visited eventually, and the hitting time  $T_j$  will be finite with the probability of "1".

if  $j$  is transient:  $\Rightarrow$  there is a positive probability that starting from state " $i$ " the chain will never visit state " $j$ "  $\Rightarrow$  meaning the hitting time  $T_j$  would be infinite with positive probability.

consequently,  $P(T_j < \infty | X_0 = i) < 1$ . However since we are looking at the probability that  $T_j$  is finite, this probability is " $\nu_{ij}$ ". because a transient state, by definition, is not expected to be hit an infinite number of times.

$$\Rightarrow \nu_{ij} = P(T_j < \infty | X_0 = i), \text{ when } j \text{ is persistent. and}$$

$$\nu_{ij} = 0, \text{ when } j \text{ is transient}$$