

Extension: Central Limit Theorem

• What is it?

It says[†]: Many independent random variables x_n each has finite expected value (mean) and variance, form a variable

$$X = \sum_{n=1}^N x_n$$

for which $P(X)$ is a Gaussian distribution function in the limit $N \rightarrow \infty$.

$$P(X) = \frac{1}{\sqrt{2\pi V_N}} e^{-\frac{(X - E_N)^2}{2V_N}}$$

$$E_N = \text{Expected value of } X \equiv \sum_{n=1}^N e_n$$

sum of means
expected value of x_n (might call it μ_n)

$$V_N = \text{Variance (of } X) \equiv \sum_{n=1}^N v_n$$

variance of x_n (might call it σ_n^2)

Valid when $V_N \rightarrow \infty$, $\frac{v_n}{V_N} \rightarrow 0$ as $N \rightarrow \infty$.

[†] CLT is not stated rigorously here (about the conditions). The point is to convey what CLT says. The conditions are usually valid in physics contexts.

Illustration: Back to spatial distribution of gas molecules

• Consider one molecule: Prob. p in fictitious box

Prob. $1-p=q$ not in box

$$e = \text{mean [what one molecule contributes to the number of molecules in box on average]} \\ = p \cdot \underset{\substack{\uparrow \\ \text{in}}}{1} + (1-p) \cdot \underset{\substack{\uparrow \\ \text{not in}}}{0} = p$$

$$v = \text{variance} = \langle (x-e)^2 \rangle = \langle x^2 \rangle - e^2 = \langle x^2 \rangle - p^2 \\ \langle x^2 \rangle = p \cdot 1^2 + (1-p) \cdot 0^2 = p$$

$$\therefore v = p - p^2 = p(1-p) = pq$$

• Now we have $N (\gg 1)$ molecules

$$X = \sum_{n=1}^N x_n \quad (\text{variable giving \# molecules in box})$$

$$E_N = \langle X \rangle = \sum_{n=1}^N e_n = Np \quad (\text{same result as before})$$

$$V_N = \sum_{n=1}^N v_n = Npq \quad (\text{same result as before})$$

Note that $\frac{v_n}{V_N} = \frac{1}{N} \rightarrow 0$ as $N \rightarrow \infty$ and $V_N = Npq \rightarrow \infty$ as $N \rightarrow \infty$

By CLT, Variable X follows the distribution function:

$$P(X) = \frac{1}{\sqrt{2\pi V_N}} e^{-\frac{(X - \langle X \rangle)^2}{2V_N}} \quad \text{as } N \rightarrow \infty \\ (\text{same result as before})$$

Other cases:

- Random Walk: Each x_n is a variable

prob. p (or $\frac{1}{2}$) walk to the right
prob. $(1-p)$ (or $\frac{1}{2}$) walk to the left

$$X = \sum_{n=1}^N x_n \quad \text{variable giving location after } N \text{ steps}$$

Try it many times (each N steps), end points fall on different places

$$E_N = \text{mean position of end point} = N \cdot (p - q) \quad (= 0 \text{ for } p = \frac{1}{2})$$

$$V_N = \text{Variance of end point positions} = N \cdot 4pq \quad (= 1 \text{ for } p = \frac{1}{2})$$

$$SD = \sqrt{V_N} = \sqrt{4pq} \cdot \sqrt{N}$$

On average
how far from origin
after time (N steps)

$\sqrt{\text{time}}$

Diffusion