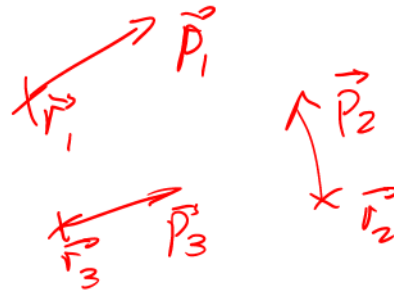


Chapter 2.1: Statistics

Probability distribution



$j = 1, \dots, N$
 $6N$ variables

Every classical microstate $\{\vec{r}_1, \vec{p}_1, \vec{r}_2, \vec{p}_2, \vec{r}_3, \vec{p}_3, \dots, \vec{r}_N, \vec{p}_N\}$ has probability density $P(\{\vec{r}_j, \vec{p}_j\})$, where

$$\int P(\{\vec{r}_i, \vec{p}_i\}) d^3\vec{r}_1 d^3\vec{p}_1 d^3\vec{r}_2 d^3\vec{p}_2 \dots d^3\vec{r}_N d^3\vec{p}_N = 1$$

Phase space integration: $\prod_j d^3\vec{r}_j d^3\vec{p}_j = d\Gamma^{6N}$

$d\Gamma$ has dim. of momenta and positions

$P(\Gamma) d\Gamma$ is dimensionless

the probability to find state in $[\Gamma, \Gamma + d\Gamma]$

Goal: S, T, E, \dots

A given probability distribution may not be diagonal in any desired basis (e.g. position)

Quantum states

can be in superposition of a large number of basis states.

A probability distribution must define a preferred basis $|r\rangle$ with quantum numbers r and probability P_r

Def. 2.1: The density matrix $\hat{\rho}$ with $\text{tr}\hat{\rho} = 1$ and positive definite eigenvalues $P_r \geq 0$ defines a quantum mechanical probability distribution. In general the system is in a mixed state.

in eigenbasis:
$$\hat{\rho} = \sum_r P_r |r\rangle\langle r| \quad \text{tr}\hat{\rho} = \sum_r P_r = 1$$

in general not given in diagonal form

pure state: definite state $|u\rangle = |r=1\rangle \quad P_1 = 1$

in this case $\hat{\rho}^2 = \hat{\rho} \quad P_{\text{rest}} = 0$

all other cases: mixed state

is represented by operator $\hat{\rho}$ not by a vector

(Not superposition)

Measurements

$$\Lambda(\{\vec{r}_j, \vec{p}_j\}) \quad \text{or} \quad \langle \hat{\Lambda} | r \rangle$$

Def. 2.2: Statistical expectation values of a quantity $\hat{\Lambda}$ are obtained by using the probability distribution

$$\langle \Lambda \rangle = \int d\Gamma P(\{\vec{r}_j, \vec{p}_j\}) \Lambda(\{\vec{r}_j, \vec{p}_j\}) \quad \text{average value}$$

$$\langle \hat{\Lambda} \rangle = \sum_r \langle r | \hat{\Lambda} | r \rangle P_r = \text{tr} \hat{\Lambda} \hat{\rho}$$

$$= \sum_{r, r'} \langle r | \hat{\Lambda} | r' \rangle \langle r' | r \rangle P_r = \sum_r \langle r | \hat{\Lambda} | r \rangle P_r = \text{tr} \hat{\Lambda} \hat{\rho}$$

also fluctuations are interesting

Example find particle at x

pure state $|\psi_r(x)\rangle^2 = |\langle r | x \rangle|^2 = \langle r | x \rangle \langle x | r \rangle = \langle r | \hat{P}_x | r \rangle$

"QM expectation value" of $\hat{\Lambda} = \hat{P}_x$

$$\hat{P}_x = |x\rangle\langle x| \hat{=} \text{mass coefficient}$$

thermal/statistical expectation

$$\langle \hat{P}_x \rangle = \sum_r \langle r | \hat{P}_x | r \rangle P_r = \sum_r \langle r | x \rangle \langle x | r \rangle P_r = \sum_r |\psi_r(x)|^2 P_r$$

Fluctuations

Def. 2.3: The standard deviation $\Delta\Lambda$ of a quantity $\hat{\Lambda}$ is given by

$$\Delta\Lambda^2 = \langle (\hat{\Lambda} - \langle \hat{\Lambda} \rangle)^2 \rangle \geq 0 \quad \text{"root mean squared" (RMS)}$$

$$\text{mean: } \lambda = \langle \hat{\Lambda} \rangle \in \mathbb{C} \text{ or } \mathbb{R}$$

$$\Delta\Lambda^2 = \langle \hat{\Lambda}^2 - 2\hat{\Lambda}\langle \hat{\Lambda} \rangle + \langle \hat{\Lambda} \rangle^2 \rangle$$

$$\Delta\Lambda = \sqrt{\langle (\hat{\Lambda} - \langle \hat{\Lambda} \rangle)^2 \rangle}$$

$$= \langle \hat{\Lambda}^2 \rangle - 2\langle \hat{\Lambda} \rangle \langle \hat{\Lambda} \rangle + \langle \hat{\Lambda} \rangle^2$$

$$\langle \lambda \rangle = \lambda \\ \forall \lambda \in \mathbb{C}$$

$$= \langle \hat{\Lambda}^2 \rangle - \langle \hat{\Lambda} \rangle^2$$

$$\Delta\Lambda^2 \geq 0 \Rightarrow \langle \hat{\Lambda}^2 \rangle \geq \langle \hat{\Lambda} \rangle^2$$

but sometimes $\langle \hat{\Lambda} \rangle \in \mathbb{C}$ exists

but $\langle \hat{\Lambda}^2 \rangle$ may diverge

Def. 2.4: The central limit theorem holds for a sum $\Lambda = \sum_{j=1}^n Y_j$ of very many independent statistical variables Y_j which have the same probability distribution $P_Y(Y_j)$ and finite ΔY . In the limit $n \rightarrow \infty$ the probability distribution for Λ approaches a Gaussian distribution

$$P(\Lambda) = \frac{1}{\sqrt{2\pi\Delta\Lambda}} \exp\left(-\frac{1}{2} \left(\frac{\Lambda - \langle\Lambda\rangle}{\Delta\Lambda}\right)^2\right)$$

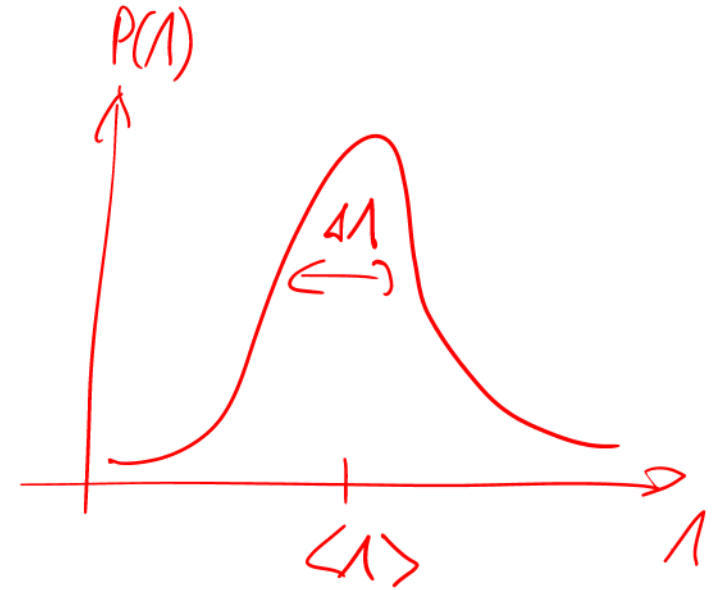
"normal distribution"

e.g. random walk

where

$$\langle\Lambda\rangle = n\langle Y\rangle \quad \text{and} \quad \Delta\Lambda = \sqrt{n}\Delta Y$$

$P(\Lambda)d\Lambda$ is probability to measure $[\Lambda, \Lambda+d\Lambda]$



$$\frac{\Delta\Lambda}{\langle\Lambda\rangle} = \frac{1}{\sqrt{n}} \frac{\Delta Y}{\langle Y\rangle} \quad \text{fluctuation decrease with more measurements}$$

examples: Binomial distribution: random walk,
Repeated measurements

Def. 2.5: A cumulative distribution function $F_\Lambda(\lambda) = P(\Lambda \leq \lambda)$ is the probability that the random variable Λ is measured with a value less than λ . Two variables are identically distributed iff $F_{\Lambda_1}(\lambda) = F_{\Lambda_2}(\lambda) \forall \lambda$. They are independent iff $F_{\Lambda_1, \Lambda_2}(\lambda_1, \lambda_2) = P(\Lambda_1 \leq \lambda_1 \wedge \Lambda_2 \leq \lambda_2) = F_{\Lambda_1}(\lambda_1)F_{\Lambda_2}(\lambda_2)$

probability densities λ

$$F_\Lambda(\lambda) = \int_{-\infty}^{\lambda} P(\Lambda) d\Lambda$$

in general

$$\Lambda_{\text{median}} \neq \Lambda_{\text{max}} \neq \langle \Lambda \rangle$$

Example: wealth distribution

