

**Notes on Physics 415:**  
**Statistical and Thermal Physics**  
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## 1 Basic Statistical Methods

### 1.1 Random walk: Binomial distribution and the Emergence of Gaussian

- **Example:** We introduce important ideas from Probability via an example of **1D random walk**:

Consider a drunkard walking along a straight line, starting from Origin  $x = 0$ , and taking random steps of length  $l$  at regular intervals. Each step is independent of the last. He takes a probability  $p$  of stepping to the left, and  $1 - p$  to step to the right. After taking  $N$  steps, what is the probability that the walker is at position  $x = ml$ ?

Let  $P_N(m)$  be the position  $x = ml$  of the drunkard after  $N$  steps; denote  $n_1$  = number of steps to the left,  $n_2 = N - n_1$  number of steps to the right. Notice that  $-N \leq m \leq N$ ,  $N = n_1 + n_2$ ,  $m = n_1 - n_2$ . Then the number of walking combinations, indexed with either direction left ( $n_1$ ) or right ( $n_2$ ), is given by the binomial coefficient:

$$\binom{N}{n_1} = \frac{N!}{n_1!(N - n_1)!} = \frac{N!}{n_1! n_2!} = \binom{N}{n_2}. \quad (1)$$

Then the probability of the walker taking  $n_1$  steps to the left and  $n_2$  steps to the right is given by the **binomial distribution**:

$$P_N(n_1) = \frac{N!}{n_1! n_2!} p^{n_1} q^{n_2}. \quad (2)$$

Noticing  $n_1 = \left(\frac{N+m}{2}\right)$  and  $n_2 = \left(\frac{N-m}{2}\right)$ :

$$P_N(m) = \frac{N!}{\left[\frac{N+m}{2}\right]! \left[\frac{N-m}{2}\right]!} p^{\frac{N+m}{2}} (1-p)^{\frac{N-m}{2}}. \quad (3)$$

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#### 1.1.1 General Notions from probability:

Let  $X$  be a random variable, taking  $N$  possible values  $x_1, x_2, \dots, x_N$  with associated probabilities  $P(x_1), P(x_2), \dots, P(x_N)$ . Note that  $0 \leq P(x_i) \leq 1$ ,  $\sum_{i=1}^N P(x_i) = 1$

- Mean:  $\bar{x} := \sum_{i=1}^N P(x_i) x_i$       Var:  $\text{var}(x) := \overline{(x - \bar{x})^2} = \sum_{i=1}^N P(x_i) (x_i - \bar{x})^2 = \overline{x^2} - \bar{x}^2$
  - RMS:  $\Delta x_{\text{rms}} = \sqrt{\overline{x^2} - \bar{x}^2}$
  - For Binomial Distribution:  $\bar{x} = Np$ , Dispersion :  $\text{var}(x) = Npq$ ,  $\Delta x_{\text{rms}} = \sqrt{Npq}$
- Relative Width:  $\frac{\Delta x_{\text{rms}}}{\bar{x}} = \frac{q}{p} \frac{1}{\sqrt{N}} \rightarrow 0 \quad (n \gg 1)$

#### 1.1.2 Central Limit Theorem: Appox. of Binom.

Recall Equation 2, taking logarithm on both sides:

$$\ln(P_N(m)) = \ln(N!) - \ln(n_1!) - \ln(N - n_1)! + n_1 \ln(p) + (N - n_1) \ln(q). \quad (4)$$

For  $N \gg 1$ , we can approximate using Stirling's formula:

$$N! \approx \sqrt{2\pi N} N^N e^{-N}, \quad (5)$$

and further algebra gives

$$P_N(m) \approx \sqrt{\frac{N}{2\pi n(N-n)}} \exp\left[-N f\left(\frac{n}{N}\right)\right], \quad (N \gg 1) \quad (6)$$

where

$$f(x) = [x \ln x + (1-x) \ln(1-x)] - [x \ln p + (1-x) \ln q]. \quad (7)$$

For  $N$  large,  $P_N$  peaks sharply near  $\max \tilde{n} = Np$ , which is found by maximizing  $f(x)$ . Expanding  $f(x)$  about  $\tilde{n}$ , and taking  $n \approx \tilde{n}$  in  $P_N$  we have :

$$\boxed{P_N(m) \approx \frac{1}{\sqrt{2\pi Npq}} \exp\left[-\frac{(n - Np)^2}{2Npq}\right]}, \quad (8)$$

which is a Gaussian distribution with mean  $\mu = \bar{x} = Np$ ,  $\sigma^2 = Npq$ ,  $\Delta x_{\text{rms}} = \sqrt{Npq}$ .

## 1.2 Probability Distribution with Multivariables

Consider two r.v.  $u, v$ , which can assume possible values  $u_i, v_j$  for  $i = 1, 2, \dots, M; j = 1, 2, \dots, N$ .

- Normalization conditoin

$$\sum_{i=1}^M \sum_{j=1}^N P(u_i, v_j) = 1. \quad (9)$$

- Unconditioned prob. distribution:

$$P(u_i) = \sum_{j=1}^N P(u_i, v_j), \quad P(v_j) = \sum_{i=1}^M P(u_i, v_j). \quad (10)$$

- Statistical independence:

$$P(u_i, v_j) = P(u_i)P(v_j), \quad (11)$$

in which case the mean of the product is the product of the means:

$$\overline{uv} = \overline{u} \overline{v}. \quad (12)$$

## 1.3 Continuous probability distribution

For continuous r.v.  $x \in (a_1, a_2)$ , assign value of r.v. to  $f(x)$ .

The probability density function  $p(x)$  is normalized:

$$\int_{a_1}^{a_2} p(x) dx = 1. \quad (13)$$

The mean and variance are defined as:

$$\bar{x} = \int_{a_1}^{a_2} x p(x) dx, \quad \text{var}(x) = \int_{a_1}^{a_2} (x - \bar{x})^2 p(x) dx. \quad (14)$$

- Especially,  $p(x) dx$  represents prob. to find  $x$  in  $[x, x + dx]$ .

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