# Topics in Physical Chemistry and Biophysics

### 1 Review of probability

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### Definition 1

Probability. If N is the total number of outcomes, and  $n_A$  fall in category A, then

$$P_A = \frac{n_A}{N} = \frac{\text{outcomes cat. } A}{\text{all outcomes}}$$

Rules of composite events:

- 1. Mutually exclusive: outcomes  $(A_1, A_2,...)$  are mutually exclusive if one outcome precludes another outcomes. (Event  $A_1$  prevents even  $A_2$  from happening simultaneously.)
- 2. Collectively exhaustive: if all known outcomes are also all possible outcomes.  $\sum P_i = 1$ .
- 3. Independence: outcomes do not depend on each other.
- 4. Multiplicity: total number of ways in which outcomes occur.

Rules of calculation:

1. Let there be 3 outcomes A, B, C with probability  $P_A, P_B, P_C$ . What is the probability that either one occurs (A or B or C)?

$$P(A \cup B \cup C) = P_A + P_B + P_C$$

That's the addition rule.

2. Probability that all outcomes occur? (Assuming independence)

$$P(A \cap B \cap C) = P_A P_B P_C$$

3. Probability that an event A is not happening?  $P = 1 - P_A$ 

**Example 1.** We roll a die twice. What is the probability of rolling a 1 first **or** a 4 second? Split the problem to parts. Note that the events are not mutually exclusive. Condition applies if:

- 1 first and not a 4 second:  $\frac{1}{6} \cdot \frac{5}{6}$
- not a 1 first and a 4 second:  $\frac{5}{6} \cdot \frac{1}{6}$
- 1 first and 4 second:  $\frac{1}{6} \cdot \frac{1}{6}$

Now sum up all of the options to get result.

#### Definition 2

Correlated events. P(B|A) is the probability that B occurs given A has occured.

Joint probability. P(AB) that both A and B occur.

#### Definition 3

General multiplication rule.

$$P(AB) = P(B|A) P(A)$$

P(A) is called the a priori probability and P(B|A) is called the a posterior probability

### Theorem 1

Bayes theorem.

$$P(B|A) P(A) = P(A|B) P(B)$$

**Example 2.** 1% of population has breast cancer. We use mammography to detect cancer.

Event A: breast cancer. P(A) = 0.01.  $P(\bar{A}) = 1 - P(A) = 0.99$ .

Event B: diagnosis. P(B|A) = 0.8.  $P(B|\bar{A}) = 0.096$ . (i.e. false positive)

What is the chance that a doctor has diagnosed someone with cancer? i.e. P(A|B)

$$P(A|B) = \frac{P(B|A) P(A)}{P(B)}$$

P(B) is the diagnosis of breast cancer irrespective wheter it's there or not there.

$$P(B) = P(BA) + P(B\bar{A}) = P(B|A)P(A) + P(B|\bar{A})P(\bar{A}) = 0.8 \cdot 0.01 + 0.096 \cdot 0.99 = 0.103$$

$$P(A|B) = \frac{0.8 \cdot 0.01}{0.103} = 0.078 = 7.8\%$$

The reason that P(A|B) is so small is that the rate of false positive is really low and the rate of having breast cancer is really low.

Combinatorics: concerned with composition of events, and not their order.

**Example 3.** How many combinations of there are of N amino acids?

$$W = N! = N (N - 1) (N - 2) \cdots$$

**Example 4.** Distinguish or not Distinguish: What are the possible number of ways to arrange N amino acids? Divide all permutations (assuming objects are distinguishable) by the number of permutations of objects that are indistinguishable.

$$W = \frac{N!}{N_A}$$

In general, for N objects consisting of t categories in which the objects are indistinguishable:

$$W = \frac{N!}{(n_1!)(n_2!)\cdots(n_t!)}$$

So, if t=2, (e.g. possible number of ways to arrange three acids A,A,H)

$$W = \frac{N!}{n_1! \cdot n_2!} = \frac{N!}{n_1! (N - n_1)!} = {N \choose n}$$

## Definition 4

Distribution functions. Describe collections of probabilities. Relevant for continuous variables.

$$\sum_{i} p_{i} \to \int_{a}^{b} p(x) \, \mathrm{d}x$$

Popular distributions:

1. Binomial Distribution. Relevent when there are only two outcomes.

**Example 5.** What is the probability that a series of N trials has  $n_H$  heads and  $n_T$  tails in any order?

 $P_H, P_T$  are mutually exclusive, so the probability of one sequence is

$$P_H^{n_H} \cdot P_T^{n_T} = P_H^{n_H} (1 - P_H)^{N - n_H}; \quad N = n_H + n_T$$

and the number of ways to arrange the coins is

$$W = \frac{N!}{n_H! \left(N - n_H\right)!}$$

therefore, the possibility for the outcome (getting  $n_H$  and  $n_T$ ) in any order is

$$P(n_H, N) = {N \choose n_H} p_H^{n_H} (1 - p_H)^{N - N_H}$$

that's the binomial distribution.

**Example 6.** Given the molecule  $C_{27}H_{44}O$  such that 1.1% is  $^{13}C$  and the rest are  $^{12}C$ , the fraction of molecules without a single  $^{13}C$  is given by the binomial distribution.

2. Multinomial distribution. Basically the extension of the binomial distribution.

$$P(n_1, n_2, \dots, n_t, N) = \left(\frac{N!}{n_1! \, n_2! \cdots n_t!}\right) p_1^{n_1} \, p_2^{n_2} \cdots p_t^{n_t}$$

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## Definition 5

Moments of distributions. Averages and Variances of distribution functions.

Given p(i) s.t.  $\sum_{i} p_{(i)} = 1$ , the **Average** is defined as

$$\langle i \rangle = \sum_{i} i p(i) \rightarrow \langle x \rangle = \int x p(x) dx$$

Given f(x),

$$\langle f(x) \rangle = \int f(x) p(x) dx$$

Given  $a \in \mathbb{R}$ 

$$\langle a f(x) \rangle = \int a f(x) p(x) dx = a \langle f(x) \rangle$$

Given 2 functions f(x), g(x),

$$\langle f(x) + g(x) \rangle = \langle f(x) \rangle + \langle g(x) \rangle$$

$$\langle f(x) \cdot g(x) \rangle \neq \langle f(x) \rangle \langle g(x) \rangle$$

The 2nd and 3nd **Moments** of the distributions p(x) are

$$\langle x^2 \rangle = \int x^2 p(x) \, \mathrm{d}x$$

$$\langle x^3 \rangle = \int x^3 p(x) \, \mathrm{d}x$$

The **Variance** of the distribution,  $\sigma^2$  is defined as

$$\sigma^2 = \langle x^2 \rangle - \langle x \rangle^2 = \langle (x - \langle x \rangle)^2 \rangle$$