

# STATISTICAL PHYSICS & THERMODYNAMICS

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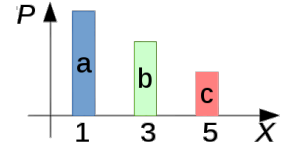
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## SAMPLE SOLUTIONS EXERCISE 3

### EXERCISE 3.1: RECONSTRUCT PROBABILITIES FROM GIVEN MOMENTS (4P)

Consider a system with three configurations  $\Omega = \{a, b, c\}$  together with an associated map  $X : \Omega \rightarrow \mathbb{R}$  defined by

$$X_a = 1, \quad X_b = 3, \quad X_c = 5.$$



- How many moments do you have to know in order to be able to reconstruct the probability distribution  $\{P_a, P_b, P_c\}$  and why?
- Suppose that  $M_1$  and  $M_2$  are given (with the map  $X$  defined above). Find the solution for the probabilities  $P_a, P_b$ , and  $P_c$ .
- Not all possible values for  $M_1$  and  $M_2$  lead to consistent results. Give a counterexample.
- Let  $M_1 = \frac{7}{2}$  and  $M_2 = 15$ . Compute the probabilities  $P_a, P_b, P_c$  and find a general formula for the higher moments  $M_n$  with  $n = 3, 4, \dots, \infty$ .

### SAMPLE SOLUTION

- Two! We have three unknowns  $\{P_a, P_b, P_c\}$  and therefore we need three equations in total. One of them is the normalization condition  $P_a + P_b + P_c = 1$ . Therefore, we have to know two moments which provide remaining two equations.
- Solving the equations<sup>1</sup>

$$\begin{aligned} 1 &= P_a + P_b + P_c \\ M_1 &= P_a X_a + P_b X_b + P_c X_c \\ M_2 &= P_a X_a^2 + P_b X_b^2 + P_c X_c^2 \end{aligned}$$

we find the solution

$$\begin{aligned} P_a &= \frac{M_2 + X_b X_c - M_1(X_b + X_c)}{(X_a - X_b)(X_b - X_c)} = \frac{1}{8}(15 - 8M_1 + M_2) \\ P_b &= \frac{M_2 + X_a X_c - M_1(X_a + X_c)}{(X_a - X_b)(X_c - X_b)} = \frac{1}{4}(-5 + 6M_1 - M_2) \\ P_c &= \frac{M_2 + X_a X_b - M_1(X_a + X_b)}{(X_a - X_c)(X_b - X_c)} = \frac{1}{8}(3 - 4M_1 + M_2) \end{aligned}$$

- Probabilities must not be negative or larger than 1. A possible counterexample is e.g.  $M_1 = 1, M_2 = 2$  where  $P_b = -1/4$ . There are many more examples of that kind.

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<sup>1</sup>We encourage the use of algebraic software packages like *Mathematica*<sup>®</sup>. But please do not submit uncommented notebooks as solutions.

(d) Inserting the solution into  $M_n = P_a X_a^n + P_b X_b^n + P_c X_c^n$  we get

$$M_n = \frac{1}{4}(6M_1 - M_2 - 5)3^n + \frac{1}{8}(-4M_1 + M_2 + 3)5^n + \frac{1}{8}(-8M_1 + M_2 + 15)$$

For  $n = 1, 2$  this expression reduces correctly to  $M_1, M_2$ .

**Note:** The purpose is to make the students aware of the fact that knowing all moments is like knowing the whole probability distribution (up to subtle mathematical exceptions).

### EXERCISE 3.2: POISSON DISTRIBUTION

(6P)

The poisson distribution  $P_\lambda(k) = \frac{\lambda^k}{k!} e^{-\lambda}$  can be understood as the limit of the binomial distribution in the case of “rare events”.

- Let  $p = \lambda/N$  and take  $N \rightarrow \infty$  while keeping  $\lambda$  and  $k$  constant. Show that in this limit we can approximate  $(1 - p)^{N-k} \approx e^{-\lambda}$ . (1P)
- Show similarly that  $\binom{N}{k} \approx \frac{N^k}{k!}$ . (1P)
- Use (a) and (b) to show that in this limit the binomial distribution tends to the Poisson distribution. (1P)
- Check that the Poisson distribution is properly normalized. (1P)
- Compute the moment- and cumulant-generating functions. (1P)
- Determine all cumulants. (1P)

### SAMPLE SOLUTION

- Let us consider the logarithm of the left-hand side  $(N - k) \ln(1 - p)$ . Since  $p \rightarrow 0$  for  $N \rightarrow \infty$  we can approximate

$$\ln(1 - p) \approx -p + \mathcal{O}(p^2).$$

Therefore  $(N - k) \ln(1 - p) \approx -Np + kp \approx -Np = -\lambda$ , hence  $(1 - p)^{N-k} \approx e^{-\lambda}$

- For this we have to show that  $N!/(N - k)! \approx N^k$ . We take the logarithm and apply Stirlings formula

$$\ln N! - \ln(N - k)! \approx N \ln N - N - (N - k) \ln(N - k) + (n - k)$$

Since  $k \ll N$  we can further approximate  $\ln(N - k) \approx \ln N - k/N$ . Inserting this approximation in the formula given above and simplifying the expression we arrive at

$$\ln N! - \ln(N - k)! \approx k \ln N.$$

Exponentiating this we finally obtain  $\binom{N}{k} \approx \frac{N^k}{k!}$ .

- Simply insert the results from (a) and (b):

$$\binom{N}{k} p^k (1 - p)^{(N-k)} \approx \frac{N^k}{k!} p^k e^{-\lambda} = \frac{\lambda^k e^{-\lambda}}{k!}$$

(d) Check normalization:

$$\sum_{k=0}^{\infty} \frac{\lambda^k e^{-\lambda}}{k!} = e^{-\lambda} \sum_{k=0}^{\infty} \frac{\lambda^k}{k!} = e^{-\lambda} e^{\lambda} = 1.$$

(e) The moment-generating function can also be found in the lecture notes:

$$M(t) = \langle e^{kt} \rangle = \sum_{k=0}^{\infty} \frac{\lambda^k}{k!} e^{-\lambda} e^{kt} = e^{-\lambda} \sum_{k=0}^{\infty} \frac{(\lambda e^t)^k}{k!} = e^{\lambda(e^t - 1)}.$$
$$K(t) = \ln M(t) = \lambda(e^t - 1).$$

(f) Take the derivative of the cumulant-generating function:

$$\kappa_n = \left. \frac{d^n}{dt^n} K(t) \right|_{t=0} = \left. \lambda e^t \right|_{t=0} = \lambda.$$

Therefore, the Poisson distribution is special in so far as all cumulants coincide.

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### EXERCISE 3.3: TRANSFORMATION OF A PROBABILITY DENSITY (2P)

Suppose that the random variable  $X \in [0, 2]$  is distributed according to the probability density  $p_X(x) = \frac{3}{4}x(2-x)$ . How is the random variable  $Y = X^2$  distributed? Check the normalization of  $p_Y(y)$ .

#### SAMPLE SOLUTION

For completeness we first check the normalization of the given distribution (not required):

$$\frac{3}{4} \int_0^2 x(2-x) dx = 1.$$

The new random variable  $Y = X^2$  is distributed according to (see Lecture Notes)

$$p_Y(y) = \frac{p_X(f^{-1}(y))}{|f'(f^{-1}(y))|}.$$

Inserting  $p_X(x) = \frac{3}{4}x(2-x)$  and  $y = f(x) = x^2 \Rightarrow x = f^{-1}(y) = \sqrt{y}$  we get (1P)

$$p_Y(y) = \frac{3}{8}(2 - \sqrt{y}).$$

The new domain of  $Y$  is no longer  $[0, 2]$  but  $[0, 4]$ . The normalization is (1P)

$$\int_0^4 p_Y(y) dy = \frac{3}{8} \int_0^4 (2 - \sqrt{y}) dy = 1.$$

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( $\Sigma = 12P$ )