

# 作业一

杨国威 222081202016

August 2023

**task 1** Compute entropy of geometric distribution:

$$P(X = k) = p(1 - p)^{k-1} \quad (1)$$

The entropy of a geometric distribution with parameter  $p$  is given by the formula:

$$H(X) = - \sum_{k=1}^{\infty} P(X = k) \log_2 P(X = k) \quad (2)$$

Substituting the probability mass function of the geometric distribution, we get:

$$H(X) = - \sum_{k=1}^{\infty} p(1 - p)^{k-1} \log_2 [p(1 - p)^{k-1}] \quad (3)$$

Using the properties of logarithms, we can simplify this expression as follows:

$$H(X) = - \sum_{k=1}^{\infty} p(1 - p)^{k-1} [\log_2 p + (k - 1) \log_2 (1 - p)] \quad (4)$$

$$H(X) = -p \log_2 p \sum_{k=1}^{\infty} (1 - p)^{k-1} - p \log_2 (1 - p) \sum_{k=1}^{\infty} (k - 1)(1 - p)^{k-1} \quad (5)$$

The first summation is a geometric series with common ratio  $(1 - p)$ , which converges to  $\frac{1}{p}$ . The second summation is the derivative of the first summation with respect to  $(1 - p)$ , which converges to  $\frac{1}{p(1-p)}$ . Therefore, we can write:

$$H(X) = -p \log_2 p \frac{1}{p} - p \log_2 (1 - p) \frac{1}{p(1 - p)} \quad (6)$$

Simplifying further, we get the final answer:

$$H(X) = -\log_2 p - (1 - p) \log_2 (1 - p) \quad (7)$$

**task 2** Construct a discrete distribution with infinite entropy

One possible way to construct a discrete distribution with infinite entropy is to use a **Zipf distribution** The general formula for a Zipf distribution is:

$$P(X = k) = \frac{1}{k^s} \frac{1}{\delta(s)} \quad (8)$$

where  $k$  is the rank of the event,  $s$  is a positive parameter that controls the shape of the distribution, and  $\delta(s)$  is the Riemann zeta function, which is a normalization constant that ensures the probabilities sum to 1. The entropy of a Zipf distribution is given by:

$$H(X) = \log_2 \delta(s) + s \frac{\delta'(s)}{\delta(s)} \quad (9)$$

where  $\delta'(s)$  is the derivative of the Riemann zeta function. As  $s$  approaches 1 from above, the entropy of the Zipf distribution diverges to infinity. This is because the distribution becomes more and more uniform as  $s$  decreases, and a uniform distribution has maximum entropy.

**task 3** Compute KL divergence between Gaussians:  $N(\mu_1, \sigma_1), N(\mu_2, \sigma_2)$ .