Statistical Inference I

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## Lecture Notes 11

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## 0.1 Discrete Distributions

**Definition 1 (Discrete Uniform Distribution)** Suppose X follows discrete uniform distribution then it has density

$$f_X(|N) = \frac{1}{N} \mathbb{1}_{\{1,2,3...N\}}(x)$$

where N is an integer, with notation  $X \sim Discrete\ Uniform(N)$ .

**Property 1** Given N, X follows discrete uniform distribution then,

1. 
$$E[X|N] = \sum_{i=1}^{N} P(X=i)i = \frac{N+1}{2}$$

2. 
$$Var(X|N) = E[X^2|N] - E[X|N]^2 = \frac{N^2 - 1}{12}$$

## Intuition (Usage in statistics)

How can we test the two given data group X,Y follows the same distribution?

 $X_1, X_2...X_n \stackrel{iid}{\sim} F_1(x)$  and  $Y_1, Y_2...Y_m \stackrel{iid}{\sim} F_1(x)$  want to test  $H_0: F_1(x) = F_2(x) \ \forall x$ 

**Kolmogrov statistics**: Using empirical distribution of  $F_1(x), F_2(x)$ 

$$\hat{F}_1(x) = \frac{1}{n} \sum_{i=1}^n \mathbb{1}(X_i \le x)$$

$$\hat{F}_2(x) = \frac{1}{m} \sum_{i=1}^m \mathbb{1}(Y_i \le x)$$

We have Kolmogrov statistics:

$$\sup_{x} |\hat{F}_1(x) - \hat{F}_2(x)|$$

which can not be too big if X and Y following same distributions.

Rank statistics(Wilcoxon test): Instead of using true value as the predictor. We use the order i.e. rank of the data in the group. We combine X and Y and sort them to give rank

$$W = \frac{1}{n} \sum_{i=1}^{n} Rank(X_i)$$

To prevent the issue that extreme values dominated the statistics. And  $X \sim Y$  if W is not too big or too small.

**Definition 2 (Bernoulli Distribution)** X follows Bernoulli distribution then it has density

$$f_X(x|p) = p^x (1-p)^{1-x} \mathbb{1}_{\{0,1\}}(x)$$

where  $0 \le p \le 1$  denoting as  $X \sim Bernoulli(p)$ .

**Property 2** Given p, X follows binomial distribution then,

- 1.  $E[X^m|p] = E[X|p] = p$
- 2.  $Var(X|p) = p p^2 = p(1-p)$
- 3.  $F_X(x) = P(X \le x) = E[I(X \le x)] = E[N(x)]$

**Definition 3 (Binomial Distribution)**  $X_1, X_2...X_n$  *i.i.d* follows Bernoulli(p),  $let X = \sum_{i=1}^n X_i$ , X follows binomial distribution having density

$$f_X(x) = \binom{n}{x} p^x (1-p)^x \mathbb{1}_{\{0,1,2,3...n\}}(x)$$

## Intuition (Independent)

 $X_1, X_2...X_n$  are independent iff

$$P(X_1 = x_1, X_2 = x_2...X_n = x_n) = \prod_{i=1}^{n} f(x_i|p)$$

The mutually independent property automatically satisfied since we can think of  $\Omega = \Omega_1 \times \Omega_2... \times \Omega_n$  where  $\Omega_i = \{0,1\}$  for the *i*th Bernoulli trial. And  $\bigcup (A_i \in \Omega_1)(\Omega_2... \times \Omega_n)$  augmented  $\Omega$ .

**Remark 1** In reality,  $X_1, X_2...X_n$  are not i.i.d.. Since we sometimes sample population with common factors. They may affect each other, within positive or negative relation.

- 1.  $\rho > 0$ : over-dispersion binomial distribution.
- 2.  $\rho < 0$ : under-dispersion binomial distribution.