

# STAT 516

## Multivariate case: multinomial distribution

Prof. Michael Levine

April 20, 2020

- ▶ Let  $X_1, X_2, \dots, X_n$  be the discrete random variables defined on a common sample space  $\Omega$ .
- ▶ Each  $X_i$  takes values in some countable set  $\aleph_i$
- ▶ The joint pmf of  $(X_1, X_2, \dots, X_n)$  is defined as

$$p(x_1, x_2, \dots, x_n) = P(X_1 = x_1, \dots, X_n = x_n)$$

whenever  $x_i \in \aleph_i$  and 0 otherwise

# Requirements of joint pmf

1.

$$p(x_1, \dots, x_n) \geq 0$$

for any  $x_1, \dots, x_n \in R$

2.

$$\sum_{x_1 \in \mathcal{X}_1, \dots, x_n \in \mathcal{X}_n} p(x_1, \dots, x_n) = 1$$

# Multinomial distribution

- ▶ Let  $n$  balls be distributed to  $k$  cells independently
- ▶ Each ball has the probability  $p_i$  of being dropped into the  $i$ th cell
- ▶ We consider  $X_i$ ,  $i = 1, \dots, k$  as the number of balls that gets dropped into the  $i$ th cell
- ▶ Their joint pmf with parameters  $n, p_1, \dots, p_k$  is

$$P(X_1 = x_1, \dots, X_k = x_k) = \frac{n!}{x_1! \dots x_k!} p_1^{x_1} \dots p_k^{x_k}$$

where  $x_i \geq 0$ ,  $\sum_{i=1}^k x_i = n$ ,  $p_i \geq 0$  and  $\sum_{i=1}^k p_i = 1$

- ▶ We write  $(X_1, \dots, X_k) \sim \text{Mult}(n, p_1, \dots, p_k)$  - a random vector with a multinomial distribution

# Example 1

1. Suppose a fair die is rolled 30 times
2. We have  $(X_1, \dots, X_6) \sim \text{Mult}(30, p_1, \dots, p_6)$  where each  $p_i = \frac{1}{6}$
3. The probability

$$\begin{aligned} P(X_1 = 5, \dots, X_6 = 5) &= \frac{30!}{(5!)^6} \left(\frac{1}{6}\right)^5 \cdots \left(\frac{1}{6}\right)^5 \\ &= 0.0004 \end{aligned}$$

## Example I

1. Note that each of the thirty rolls is either six or not
2. The probability of getting 6 on each roll is  $\frac{1}{6}$  and the rolls are independent
3. Since  $X_6 \sim \text{Bin}(30, 1/6)$ ,

$$\begin{aligned} P(X_6 \geq 5) &= 1 - P(X_6 \leq 4) = 1 - \sum_{x=0}^4 \binom{30}{x} \left(\frac{1}{6}\right)^x \left(\frac{5}{6}\right)^{30-x} \\ &= .5757 \end{aligned}$$

# Basic properties of the multinomial distribution

► Let  $(X_1, \dots, X_k) \sim \text{Mult}(n, p_1, \dots, p_k)$ . Then

1.  $E(X_i) = np_i$ ;  $\text{Var}(X_i) = np_i(1 - p_i)$
2. For any  $i$ ,  $X_i \sim \text{Bin}(n, p_i)$
3.  $\text{Cov}(X_i, X_j) = -np_i p_j$  for any  $i \neq j$
4.  $\rho_{X_i, X_j} = -\sqrt{\frac{p_i p_j}{(1-p_i)(1-p_j)}}$  for any  $i \neq j$
5. For any  $m$  such that  $1 \leq m \leq k$  the conditional distribution

$$(X_1, \dots, X_m) | (X_{m+1} + X_{m+2} + \dots + X_k) = s \sim \text{Mult}(n-s, \theta_1, \dots, \theta_m)$$

$$\text{where } \theta_i = \frac{p_i}{p_1 + \dots + p_m}$$

# Basic properties of the multinomial distribution

- ▶ (1) Denote  $W_{ir}$  as the indicator of the event that  $r$ th ball lands in the  $i$ th cell
- ▶ For given  $i$ , variables  $W_{ir}$  are independent; thus,  
$$X_i = \sum_{r=1}^n W_{ir}$$
- ▶ Therefore,  $E(X_i) = E(\sum_{r=1}^n W_{ir}) = np_i$  and  
$$\text{Var}(X_i) = \text{Var}(\sum_{r=1}^n W_{ir}) = np_i(1 - p_i)$$
- ▶ (2) follows from the definition of a multinomial experiment