

Probability 2 - Notes 6

The Trinomial Distribution

Consider a sequence of n independent trials of an experiment. The binomial distribution arises if each trial can result in 2 outcomes, success or failure, with fixed probability of success p at each trial. If X counts the number of successes, then $X \sim \text{Binomial}(n, p)$.

Now suppose that at each trial there are 3 possibilities, say “success”, “failure”, or “neither” of the two, with corresponding probabilities $p, \theta, 1 - p - \theta$, which are the same for all trials. If we write 1 for “success”, 0 for “failure”, and -1 for “neither”, then the outcome of n trials can be described as a sequence of n numbers

$$\omega = (i_1, i_2, \dots, i_n), \text{ where each } i_j \text{ takes values } 1, 0, \text{ or } -1$$

Obviously, $P(i_j = 1) = p, P(i_j = 0) = \theta, P(i_j = -1) = 1 - p - \theta$.

Definition. Let X be the number of trials where 1 occurs, and Y be the number of trials where 0 occurs. The joint distribution of the pair (X, Y) is called the trinomial distribution.

The following statement provides us with .

Theorem. The joint p.m.f. for (X, Y) is given by

$$f_{X,Y}(k, l) = P(X = k, Y = l) = \frac{n!}{k!l!(n-k-l)!} p^k \theta^l (1-p-\theta)^{n-k-l},$$

where $k, l \geq 0$ and $k+l \leq n$.

Proof. The sample space consists of all sequences of length n described above. If a specific sequence ω has k “successes” (1’s) and l “failures” (0’s) then $P(\omega) = p^k \theta^l (1-p-\theta)^{n-k-l}$. There are $\binom{n}{k} \binom{n-k}{l} = \frac{n!}{k!l!(n-k-l)!}$ different sequences with k “successes” (1’s) and l “failures” (0’s). Hence $P(X = k, Y = l) = \frac{n!}{k!l!(n-k-l)!} p^k \theta^l (1-p-\theta)^{n-k-l}$. \square

The name of the distribution comes from the trinomial expansion

$$\begin{aligned} (a+b+c)^n &= (a+(b+c))^n = \sum_{k=0}^n \binom{n}{k} a^k (b+c)^{n-k} \\ &= \sum_{k=0}^n \sum_{l=0}^{n-k} \binom{n}{k} \binom{n-k}{l} a^k b^l c^{n-k-l} = \sum_{k=0}^n \sum_{l=0}^{n-k} \frac{n!}{k!l!(n-k-l)!} a^k b^l c^{n-k-l} \end{aligned}$$

Properties of the trinomial distribution

1) The marginal distributions of X and Y are just $X \sim \text{Binomial}(n, p)$ and $Y \sim \text{Binomial}(n, \theta)$. This follows the fact that X is the number of “successes” in n independent trials with p being the probability of ‘successes’ in each trial. Similar argument works for Y .

Note that therefore $E[X] = np, E[Y] = n\theta$ and $E[Y^2] = \text{Var}(Y) + (E[Y])^2 = n\theta(1-\theta) + n^2\theta^2$

2) If $Y = l$, then the conditional distribution of $X|(Y = l)$ is *Binomial* $(n - l, \frac{p}{1-\theta})$.

Proof.

$$\begin{aligned} P(X = k|Y = l) &= \frac{P(X = k, Y = l)}{P(Y = l)} = \frac{\frac{n!}{k!(n-k-l)!} p^k \theta^l (1-p-\theta)^{n-k-l}}{\frac{n!}{l!(n-l)!} \theta^l (1-\theta)^{n-l}} \\ &= \binom{n-l}{k} \left(\frac{p}{1-\theta}\right)^k \left(1 - \frac{p}{1-\theta}\right)^{n-l-k} \end{aligned}$$

for $x = 0, 1, \dots, (n - y)$. Hence $(X|Y = y) \sim \text{Binomial}(n - y, \frac{p}{1-\theta})$. \square

This is intuitively obvious. Consider those trials for which “failure” (or 0) did not occur. There are $(n - l)$ such trials, for each of which the probability that 1 occurs is actually the conditional probability of 1 given that 0 has not occurred, i.e. $\frac{p}{1-\theta}$. So you have the standard binomial set-up.

3) We shall now use the results on conditional distributions (Notes 5) and the above properties to find $Cov(X, Y)$ and the coefficient of correlation $\rho(X, Y)$.

We proved that $E[XY] = E[YE[X|Y]]$ (see the last page of Notes 5). According to property 2), $E[X|Y = l] = (n - l) \frac{p}{1-\theta}$ and thus $E[X|Y] = (n - Y) \frac{p}{1-\theta}$. Hence

$$\begin{aligned} E[XY] &= E\left[Y \times (n - Y) \frac{p}{(1 - \theta)}\right] = \frac{p}{1 - \theta} E(nY - Y^2) = \frac{p}{1 - \theta} (n^2\theta - n\theta(1 - \theta) - n^2\theta^2) \\ &= \frac{p}{(1 - \theta)} [n(n - 1)\theta(1 - \theta)] = n(n - 1)p\theta \end{aligned}$$

Therefore $Cov(X, Y) = E[XY] - E[X]E[Y] = n(n - 1)p\theta - n^2p\theta = -np\theta$ and hence

$$\rho(X, Y) = \frac{Cov(X, Y)}{\sqrt{Var(X)Var(Y)}} = \frac{-np\theta}{\sqrt{n^2p(1-p)\theta(1-\theta)}} = -\left(\frac{p\theta}{(1-p)(1-\theta)}\right)^{\frac{1}{2}}$$

Note that if $p + \theta = 1$ then $Y = n - X$ and there is an exact linear relation between Y and X . In this case it is easily seen that $\rho(X, Y) = -1$.

Definition of the multinomial distribution

Now suppose that there are k outcomes possible at each of the n independent trials. Denote the outcomes A_1, A_2, \dots, A_k and the corresponding probabilities p_1, \dots, p_k where $\sum_{j=1}^k p_j = 1$. Let X_j count the number of times A_j occurs. Then

$$P(X_1 = x_1, \dots, X_{k-1} = x_{k-1}) = \frac{n!}{x_1!x_2!\dots x_{k-1}!(n - \sum_{j=1}^{k-1} x_j)!} p_1^{x_1} p_2^{x_2} \dots p_{k-1}^{x_{k-1}} p_k^{n - \sum_{j=1}^{k-1} x_j}$$

where x_1, x_2, \dots, x_{k-1} are non-negative integers with $\sum_{j=1}^{k-1} x_j \leq n$.