

M358K: November 12th, 2021.

χ^2 : test of Independence.

Independence of categorical random variables.

Let X and Y be two categorical r.v.s.

Their joint probability mass function can be represented as:

$X \setminus Y$	y_1	y_2	\dots	y_j	\dots	y_c
x_1						
x_2						
\vdots						
x_i				p_{ij}		
\vdots						
x_r					$p_{Y(y_j)}$	
				$\sum_{i=1}^r p_{ij}$		

$$P_X(x_i) := \sum_{j=1}^c p_{ij}$$

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marginal distributions

$$p_{ij} := P[X=x_i, Y=y_j] \text{ for all } i, j$$

X and Y are independent

\Leftrightarrow

$$p_{ij} = P_X(x_i) \cdot P_Y(y_j)$$

for all i, j

Two-way Tables

... empirical counterparts of the joint pmf table

		y_j	
x_i	n_{ij}	r_i	\leftarrow total count for row i
	c_j	n	... total sample size

↑
total count for column j

n_{ij} ... the # of observed cases w/ the combination x_i, y_j

Empirically:

Q: What is the probability that you land in row i ?

$$\frac{r_i}{n}$$

Q: What is the probability that you land in column j ?

$$\frac{c_j}{n}$$

Q: If the row and column r.v.s were independent, what would be the probability of landing in the cell (i, j) ?

$$\frac{r_i}{n} \cdot \frac{c_j}{n} = \frac{r_i \cdot c_j}{n^2}$$

=> The expected count for the cell (i, j) if the two effects are independent is:

$$n \cdot \frac{r_i \cdot c_j}{n^2} = \frac{r_i \cdot c_j}{n}$$