

# Chi-Square Part 2: Independence

STAT-S520

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↓  
(A particular  
case if  
Goodness-of-fit)

# Independence: Setting

- 2 partition
- ▶ Let  $S$  be the sample space of our experiment and
    - ▶  $A_1, \dots, A_r$  partition  $S$  into  $r$  cells
    - ▶  $B_1, \dots, B_c$  also partition  $S$  into  $c$  cells
  - ▶ Think of  $A$ s and  $B$ s as two variables with different categories (partition)

# Independence: Setting

- ▶ We care about the relationship between As and Bs
- ▶ We define a third partition by

*rows*  $E_{ij} = A_i \cap B_j$

- ▶ Partitions  $A_1, \dots, A_r$  and  $B_1, \dots, B_c$  are mutually independent if and only if

*columns*

$$P(E_{ij}) = P(A_i) \cdot P(B_j)$$

for each  $ij$  pair.

$$P_{ij} = P_i \cdot P_j$$

- ▶ We use the chi-squared methods developed above to check if independence holds

## Example 2

Two partitions of criminals, one by type of crime (arson, rape, violence, stealing, coining, fraud) and the other by alcohol consumption (drinker, abstainer). Here is the sample (counts) observed:

observed counts

Expected Counts

##	drink	abstain	
## arson	<u>50</u>	43	93 arson
## rape	<u>88</u>	<u>62</u>	150 rape
## violence	155	110	
## stealing	379	300	
## coining	18	14	
## fraud	63	144	
	<u>753</u>	<u>673</u>	

$$\begin{array}{r} \text{drink} \\ 753 \times 93 \\ \hline 1426 \end{array}$$

$$\begin{array}{r} \text{abstain} \\ 673 \times 93 \\ \hline 1425 \end{array}$$

## Example 2 (continued)

The expected counts using the outer product (%o%)

```
exp = rowSums(obs)%o%colSums(obs)/sum(obs)
exp
```

observed counts (contingency table)

outer product

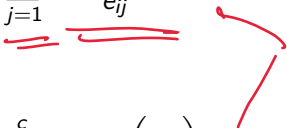
	drink	abstain
## arson	49.10870	43.89130
## rape	79.20757	70.79243
## violence	139.93338	125.06662
## stealing	358.54628	320.45372
## coining	16.89762	15.10238
## fraud	109.30645	97.69355

Expected  
counts  
under  $H_0$

$H_0$ : The variables are independent  
 $H_1$ : " " " are related

# Test statistics

As before, the test statistics are

$$\chi^2 = \sum_{i=1}^r \sum_{j=1}^c \frac{(o_{ij} - e_{ij})^2}{e_{ij}}$$


or

$$G^2 = 2 \sum_{i=1}^r \sum_{j=1}^c o_{ij} \log \left( \frac{o_{ij}}{e_{ij}} \right)$$

Same  
statistics  
as before

||

# Degrees of Freedom

- ▶ Unrestricted set:  $rc - 1$
- ▶ Restricted set:  $(r - 1) + (c - 1)$
- ▶ Degrees of freedom:

$$(rc - 1) - [(r - 1) + (c - 1)] = (r - 1)(c - 1)$$

## Example 2 (continued)

```
X2 = sum((obs - exp)^2/exp)
```

```
X2
```

```
## [1] 49.73061
```

```
G2 = sum(2*obs*log(obs/exp))
```

```
G2
```

```
## [1] 50.51729
```

```
df = (6 - 1)*(2 - 1)
```

```
1 - pchisq(X2, df)
```

```
## [1] 1.573317e-09
```

```
1 - pchisq(G2, df)
```

```
## [1] 1.085962e-09
```

$\chi^2_{(5)}$   
df

So



## Example 2 Simulation-Based

```
library(tidyverse)
df.obs = as.data.frame(obs)
data2 <- df.obs |>
  rownames_to_column("crime") |>
  pivot_longer(cols=c('drink', 'abstain'),
               names_to='alcohol',
               values_to='count') |>
  rowwise() |>
  mutate(count = list(1:count)) |>
  unnest(count) |>
  select(-count)
```

Contingency table.

Name of first variable (you choose)

Categories of 2nd variables.

Name of 2nd variable

DO NOT CHANGE

Use it only if you need to convert a contingency table into a long-form data frame (one row per individual)

## Example 2 Simulation-Based (continued)

```
null_dist <- data2 |>  
  specify(alcohol ~ crime) |>  
  hypothesize(null = "independence") |>  
  generate(reps = 1000, type = "permute") |>  
  calculate(stat = "Chisq")
```

```
null_dist |>  
  get_p_value(obs_stat = G2, direction = "greater")
```

```
## Warning: Please be cautious in reporting a p-value of 0.  
## approximation based on the number of 'reps' chosen in the  
## '?get_p_value()' for more information.
```

```
## # A tibble: 1 x 1  
##   p_value  
##   <dbl>  
## 1      0
```

Example:

Crime(c)	Alcohol(A)	
	D	A
a	$p_{11}$	$p_{12}$
r	$p_{21}$	$p_{22}$
v	$p_{31}$	$p_{32}$
s	$p_{41}$	$p_{42}$
c	$p_{51}$	$p_{52}$
f	$p_{61}$	$p_{62}$

$$H_0: \begin{aligned} p_{11} &= p_{1c} \cdot p_{1A} \\ p_{21} &= p_{2c} \cdot p_{1A} \\ &\vdots \end{aligned}$$

$$| p_{ij} = p_{ic} \cdot p_{jA} |$$

$$\begin{aligned} i &= 1, \dots, 6 \\ j &= 1, 2 \end{aligned}$$