# **Describing Categorical Variables**

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- Two-way tables.
- Conditional distributions.
- Bar Charts (and pie-charts)
- Section 1.7 of DBC

These data represent incoming emails for the first three months of 2012 for an email account.

Some variables:

- spam Indicator for whether the email was spam.
- to\_multiple Indicator for whether the email was addressed to more than one recipient.
- viagra The number of times "viagra" appeared in the email.
- num\_car The number of characters in the email, in thousands.
- number Factor variable saying whether there was no number, a small number (under 1 million), or a big number.

	$\operatorname{spam}$	to_multiple	viagra	num_char	number
1	0	0	0	11.370	big
2	0	0	0	10.504	small
3	0	0	0	7.773	small
4	0	0	0	13.256	small
5	0	0	0	1.231	none
6	0	0	0	1.091	none

#### Distribution of categorical variable

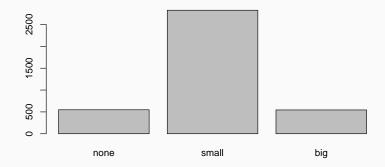
- Recall: The distribution of a variable tells us what values it takes and how often it takes these values
- In terms of categorical variables, the distribution is just the counts of cases/proportions/percents in each category.
- A table of counts for a single variable is a frequency table.

table(email\$number)

none small big 549 2827 545 • A table of proportions/percentages for a single variable is a relative frequency table.

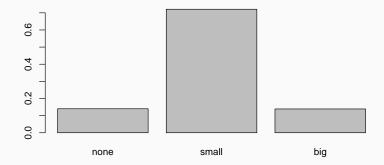
prop.table(table(email\$number))

none small big 0.140 0.721 0.139 barplot(table(email\$number)) ## need table

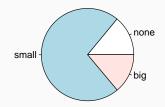


#### **Barchart of proportions**

barplot(prop.table(table(email\$number))) ## need table



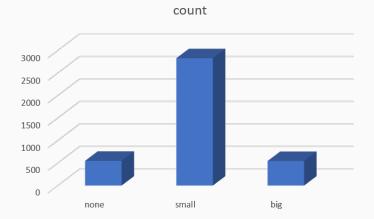
#### pie(table(email\$number))



- Humans find it easier to distinguish height rather than distinguish area.
- Which category has more emails: "big" or "none".
- In which plot is it easier to see which category has more emails?

#### Never use 3D graphics to plot 2D data

They tend to distort/obscure the view of the data and are distracting.



- What about the *joint* distribution of two categorical variables?
- The distribution of a variable tells us what values it takes and how often it takes these values.
- The joint distribution is just the counts of cases/proportions/percents in each possible combination of categories.
- A table of these counts is a contingency table, also called a two-way table.

```
tabdat <- table(email$spam, email$number)
rownames(tabdat) <- c("Not Spam", "Spam")
tabdat</pre>
```

	none	small	big
Not Spam	400	2659	495
Spam	149	168	50

	none	small	big	total
Not Spam	400	2659	495	3554
Spam	149	168	50	367
total	549	2827	545	3921

- What does 2659 represent?
- What does 495 represent?
- What does 3554 represent?
- What does 2827 represent?
- What does 3921 represent?

More informative: joint distribution in proportions:

```
prop.table(tabdat)
```

	none	small	big
Not Spam	0.10201	0.67814	0.12624
Spam	0.03800	0.04285	0.01275

- What does 0.6781 represent?
- What does 0.1262 represent?

#### **Row Proportions**

#### row proportions

The row proportions are computed as the counts divided by the row totals.

```
prop.table(tabdat, margin = 1)
```

none small big Not Spam 0.1125 0.7482 0.1393 Spam 0.4060 0.4578 0.1362

- What does 0.7482 represent?
- What does 0.1393 represent?

### **Column Proportions**

#### column proportions

The column proportions are computed as the counts divided by the column totals.

```
prop.table(tabdat, margin = 2)
```

none small big Not Spam 0.72860 0.94057 0.90826 Spam 0.27140 0.05943 0.09174

- What does 0.9406 represent?
- What does 0.9083 represent?

- Row/column proportions help us determine if two categorical variables are associated.
- E.g. Is the distribution of spam conditioned on seeing no numbers different from the distribution of spam conditioned on seeing small numbers? If so, then number and spam are associated.
- Would these be row or column proportions?
- Can also look for associations by checking the distribution of number conditioned on an email being spam and the distribution of number contioned on an email not being spam.
- Would these be row or column proportions?

```
prop.table(tabdat, margin = 2)
```

	none	small	big
Not Spam	0.72860	0.94057	0.90826
Spam	0.27140	0.05943	0.09174

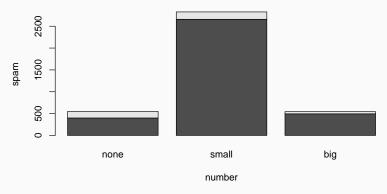
- The row/column proportions represent conditional distributions.
- Each column is the distribution of spam conditioned on either no big number (column 1), a small number (column 2), or a big number (column 3).

```
prop.table(tabdat, margin = 1)
```

none small big Not Spam 0.1125 0.7482 0.1393 Spam 0.4060 0.4578 0.1362

- The row/column proportions represent conditional distributions.
- Each row is the distribution of number conditioned on either an email being not spam (first row) or spam (second row).

### Visualizing row proportions: segmented barplot



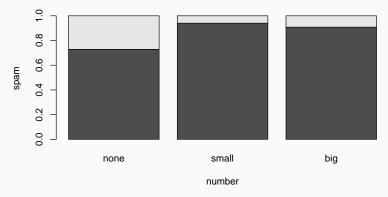
What does the bottom left box represent?

### Visualizing row proportions: standardized segmented barplot

barplot(prop.table(table(email\$spam, email\$number),

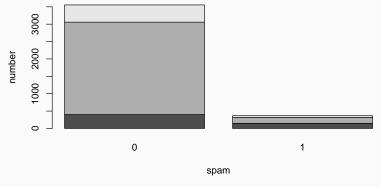
margin = 2),

xlab = "number", ylab = "spam")



What does the bottom left box represent?

#### Visualizing row proportions: segmented barplot



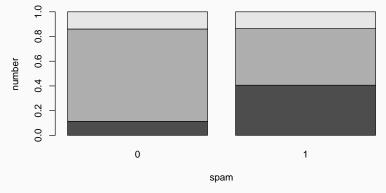
What does the bottom left box represent?

#### Visualizing row proportions: standardized segmented barplot

barplot(prop.table(table(email\$number, email\$spam),

margin = 2),

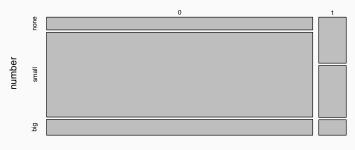
xlab = "spam", ylab = "number")



What does the bottom left box represent?

## Visualizing row proportions: mosaic plot

#### table(email\$spam, email\$number)

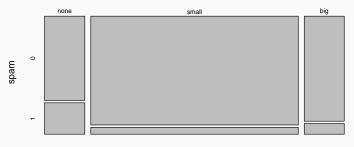


spam

Width proportional to the counts in each spam category. What does the bottom left box represent?

## Visualizing row proportions: mosaic plot

#### table(email\$number, email\$spam)



number

Width proportional to the counts in each number category. What does the bottom left box represent? • What in a mosaic plot are we looking for to see if two variables are associated?