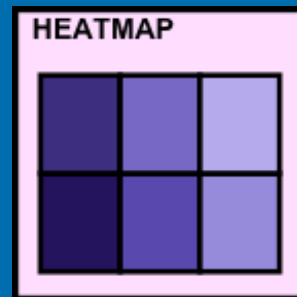
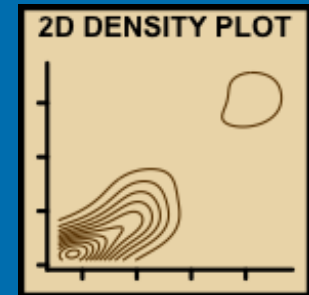
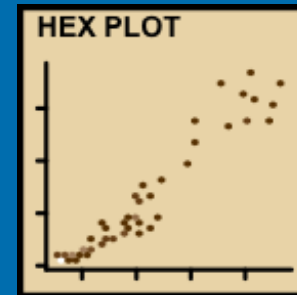
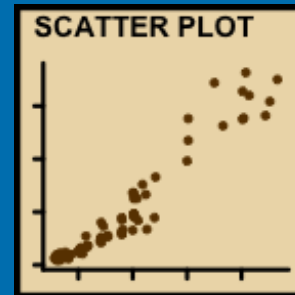
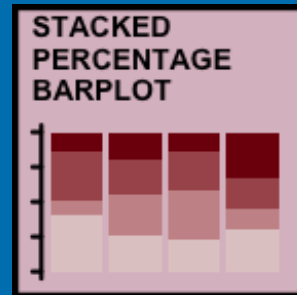
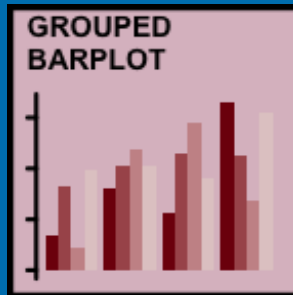
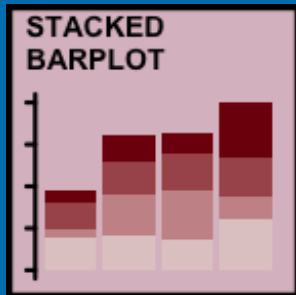
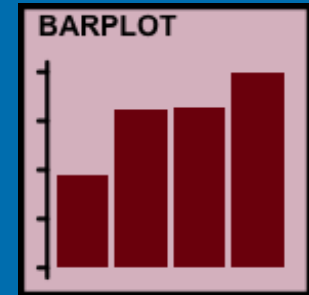
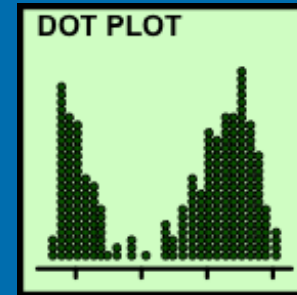
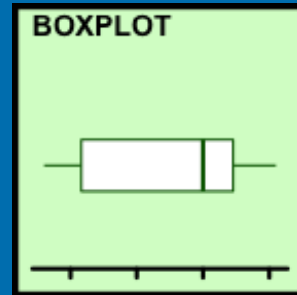
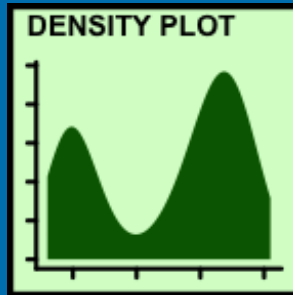
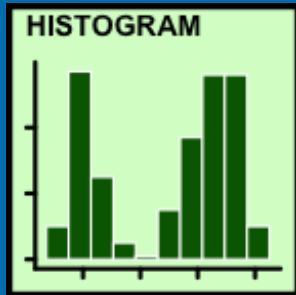


A cluster of purple flowers, possibly carnations, is positioned in the top-left corner of the slide. The flowers are in various shades of purple, from deep magenta to lighter lavender, and are partially cut off by the left edge of the frame.

Data Visualisation with R

Workshop Part 1

Catalogue



Illustrative data 💰 Wages

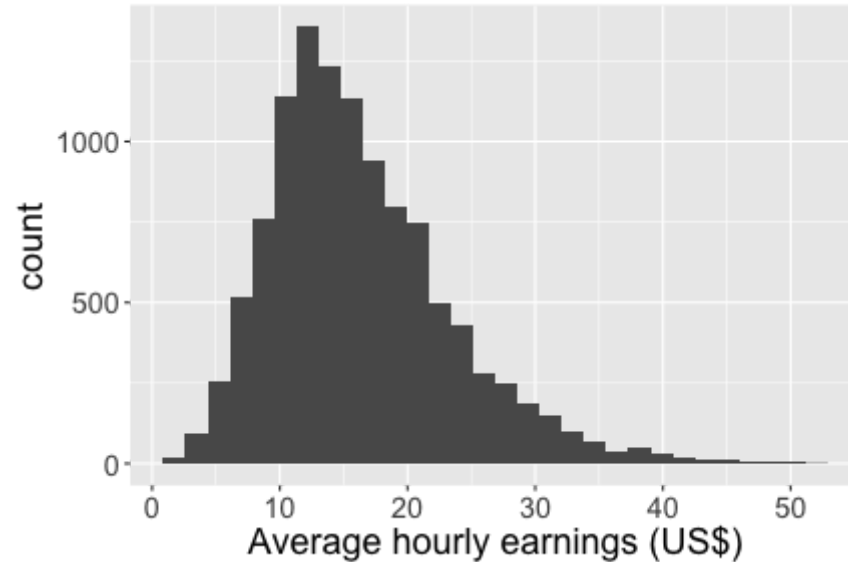
- Let's look at a survey of average hourly earnings in United States where each observation is an individual.

```
data(CPSch3, package = "Ecdat")  
str(CPSch3)
```

```
## 'data.frame':    11130 obs. of  3 variables:  
## $ year: int  1992 1992 1992 1992 1992 1992 1992 1992 1992 1992  
## $ ahe : num  13 11.6 17.4 10.1 16.8 ...  
## $ sex : Factor w/ 2 levels "male","female": 1 1 1 2 1 2
```

A histogram with `geom_histogram()`

```
ggplot(data = CPSch3,  
       mapping = aes(x = ahe)) +  
  geom_histogram() +  
  labs(x = "Average hourly earnings (US$)")
```



Layer data

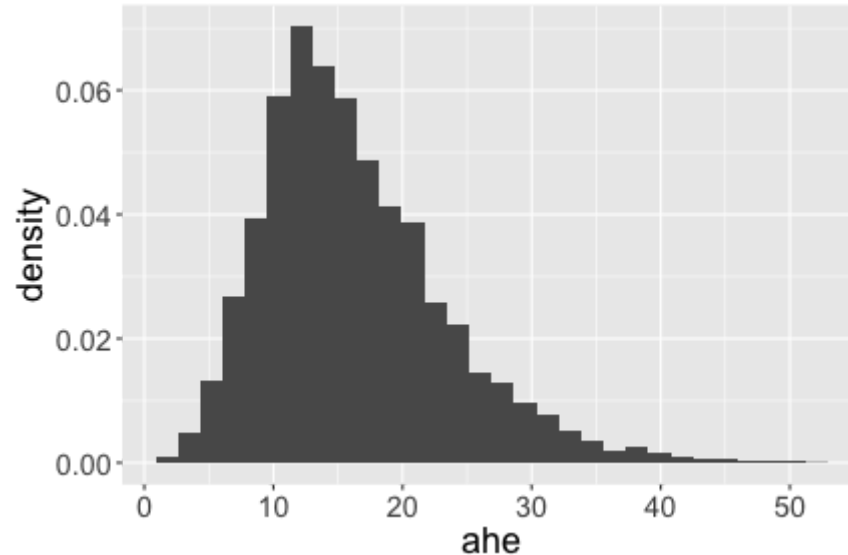
```
g <- ggplot(data = CPSch3,  
            mapping = aes(x = ahe)) +  
  geom_histogram()
```

```
layer_data(g, 1)
```

##	y	count	x	xmin	xmax	density	ncount	ndensity	flipped_aes
## 1	19	19	1.734720	0.8673601	2.602080	0.0009840768	0.014011799	0.014011799	FALSE
## 2	91	91	3.469440	2.6020802	4.336800	0.0047132101	0.067109145	0.067109145	FALSE
## 3	255	255	5.204160	4.3368004	6.071521	0.0132073470	0.188053097	0.188053097	FALSE
## 4	514	514	6.938881	6.0715206	7.806241	0.0266218681	0.379056047	0.379056047	FALSE
## 5	760	760	8.673601	7.8062407	9.540961	0.0393630735	0.560471976	0.560471976	FALSE
## 6	1143	1143	10.408321	9.5409609	11.275681	0.0591999908	0.842920354	0.842920354	FALSE
## 7	1356	1356	12.143041	11.2756811	13.010401	0.0702320101	1.000000000	1.000000000	FALSE
## 8	1232	1232	13.877761	13.0104012	14.745121	0.0638096139	0.908554572	0.908554572	FALSE
## 9	1137	1137	15.612481	14.7451214	16.479842	0.0588892297	0.838495575	0.838495575	FALSE
## 10	941	941	17.347202	16.4798416	18.214562	0.0487377002	0.693952802	0.693952802	FALSE

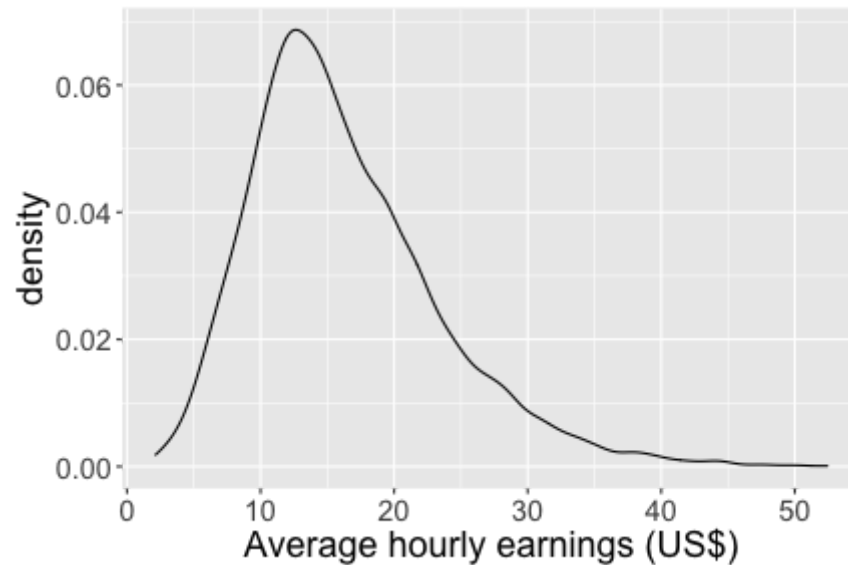
Layer data

```
ggplot(data = CPSch3,  
       mapping = aes(x = ahe)) +  
  geom_histogram(aes(y = after_stat(density)))
```



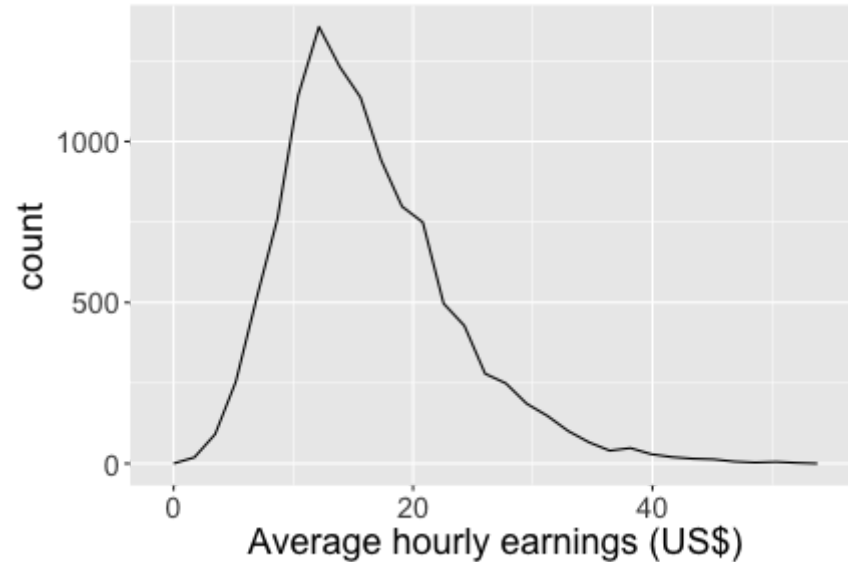
A density plot with `geom_density()`

```
ggplot(data = CPSch3,  
       mapping = aes(x = ahe)) +  
  geom_density() +  
  labs(x = "Average hourly earnings (US$)")
```



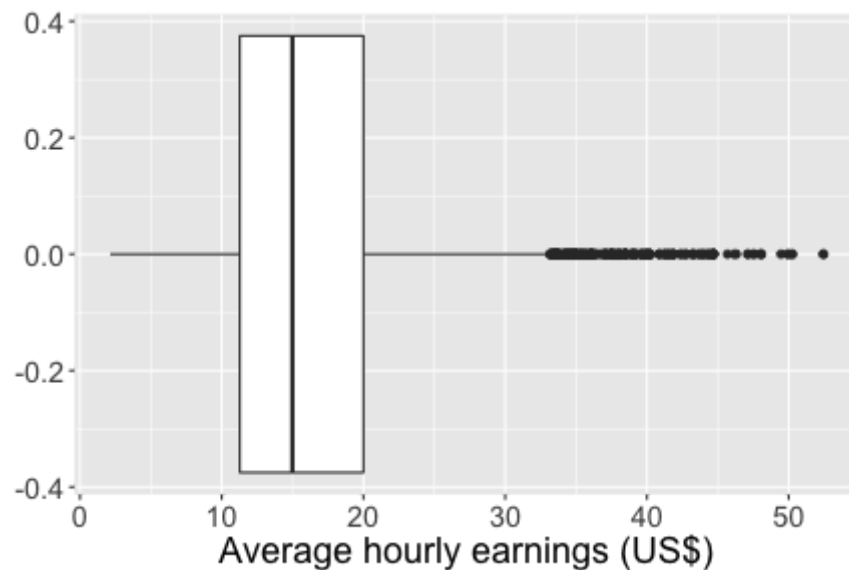
A frequency polygon with `geom_freqpoly()`

```
ggplot(data = CPSch3,  
       mapping = aes(x = ahe)) +  
  geom_freqpoly() +  
  labs(x = "Average hourly earnings (US$)")
```



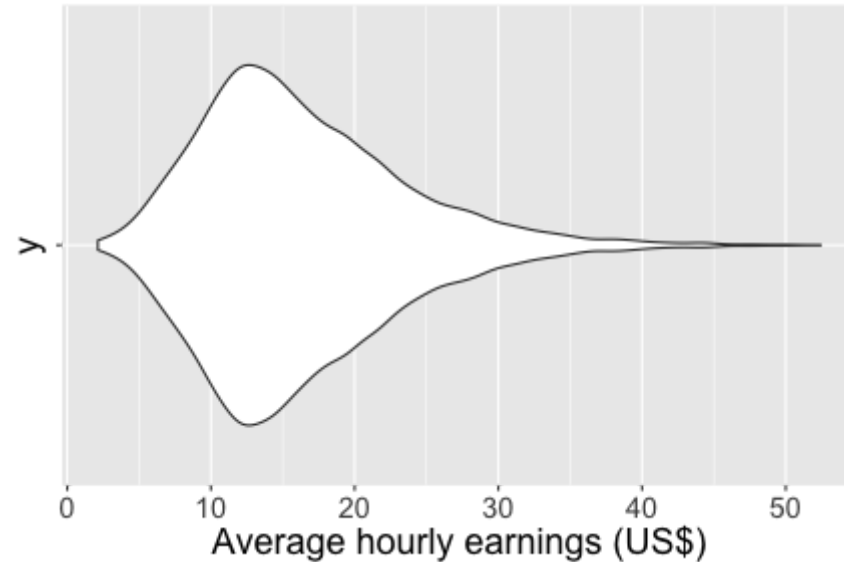
A boxplot with `geom_boxplot()`

```
ggplot(data = CPSch3,  
       mapping = aes(x = ahe)) +  
  geom_boxplot() +  
  labs(x = "Average hourly earnings (US$)")
```



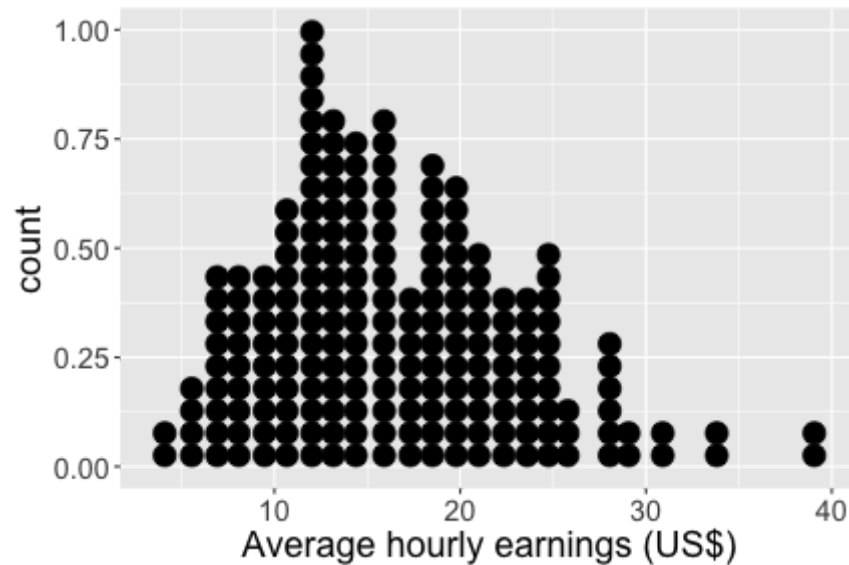
A violin plot with `geom_violin()`

```
ggplot(data = CPSch3,  
       mapping = aes(x = ahe, y = "")) +  
  geom_violin() +  
  labs(x = "Average hourly earnings (US$)")
```



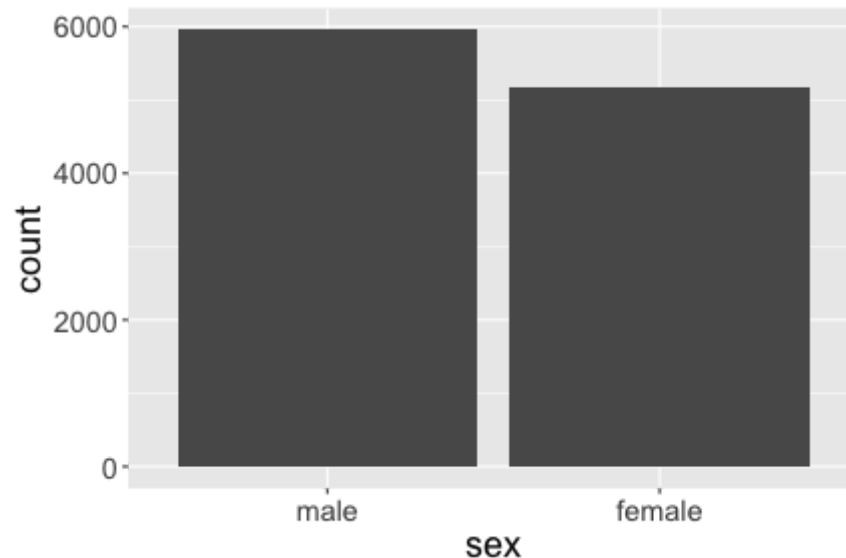
A dotplot with `geom_dotplot()`

```
ggplot(data = dplyr::sample_n(CPSch3, 200),  
       mapping = aes(x = ahe)) +  
  geom_dotplot() +  
  labs(x = "Average hourly earnings (US$)")
```



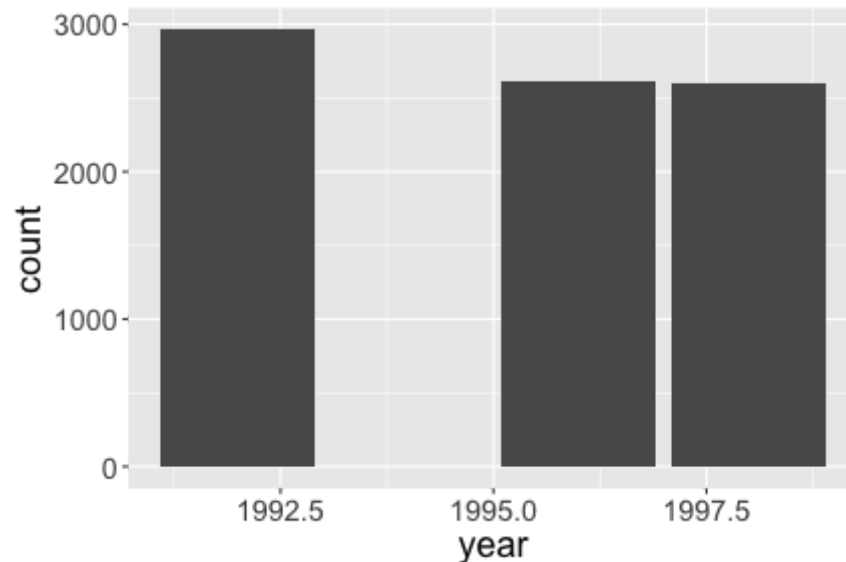
A barplot with `geom_bar()` with categorical variables

```
ggplot(data = CPSch3,  
       mapping = aes(x = sex)) +  
  geom_bar()
```



A barplot with `geom_bar()` with discrete variables

```
ggplot(data = filter(CPSch3, year != 1994),  
       mapping = aes(x = year)) +  
  geom_bar()
```



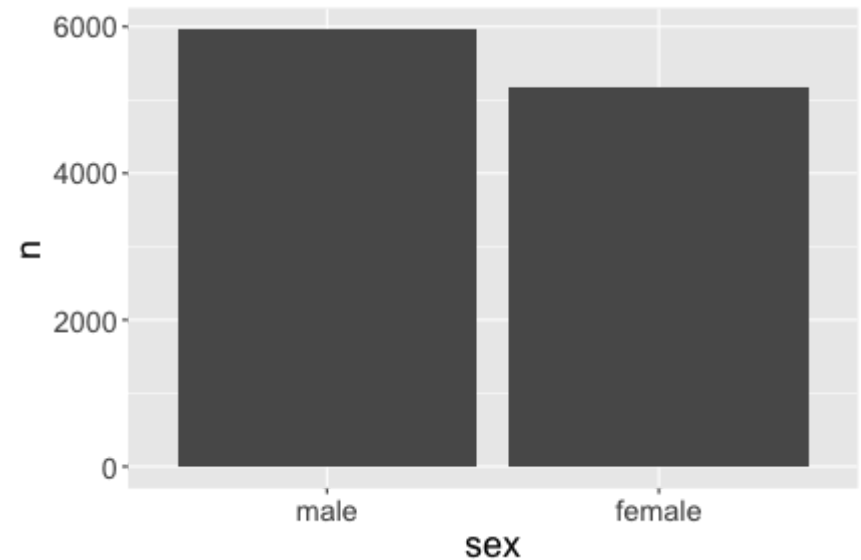
A barplot with `geom_col()`

```
dfsum <- CPSch3 %>%  
  group_by(sex) %>%  
  tally()
```

```
dfsum
```

```
## # A tibble: 2 × 2  
##   sex      n  
##   <fct> <int>  
## 1 male   5956  
## 2 female 5174
```

```
ggplot(data = dfsum,  
       mapping = aes(x = sex,  
                     y = n)) +  
  geom_col()
```



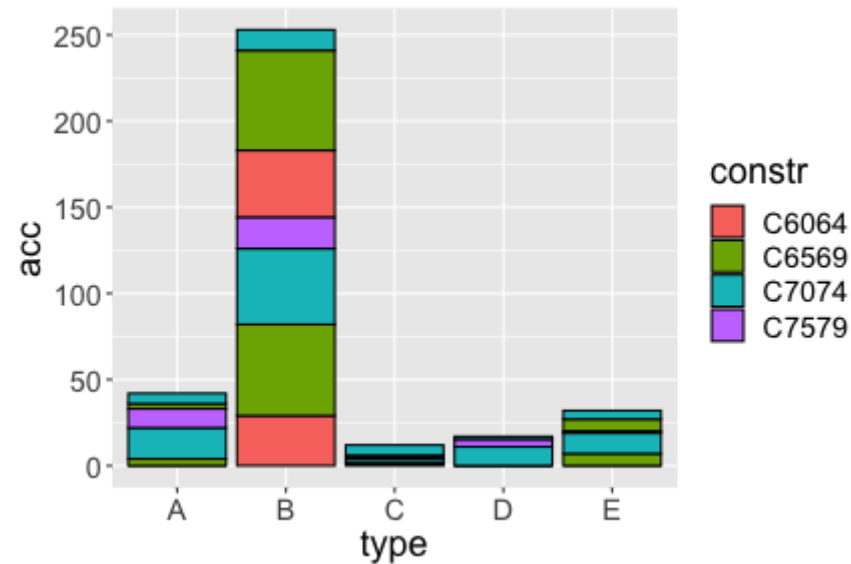
Illustrative data ship accidents

```
data(Accident, package = "Ecdat")  
str(Accident)
```

```
## 'data.frame':    40 obs. of  5 variables:  
## $ type      : Factor w/ 5 levels "A","B","C","D",...: 1 1 1  
## $ constr    : Factor w/ 4 levels "C6064","C6569",...: 1 1 2  
## $ operate   : Factor w/ 2 levels "06074","07579": 1 2 1 2 1  
## $ months    : int   127 63 1095 1095 1512 3353 NA 2244 44882  
## $ acc       : int    0 0 3 4 6 18 NA 11 39 29 ...
```

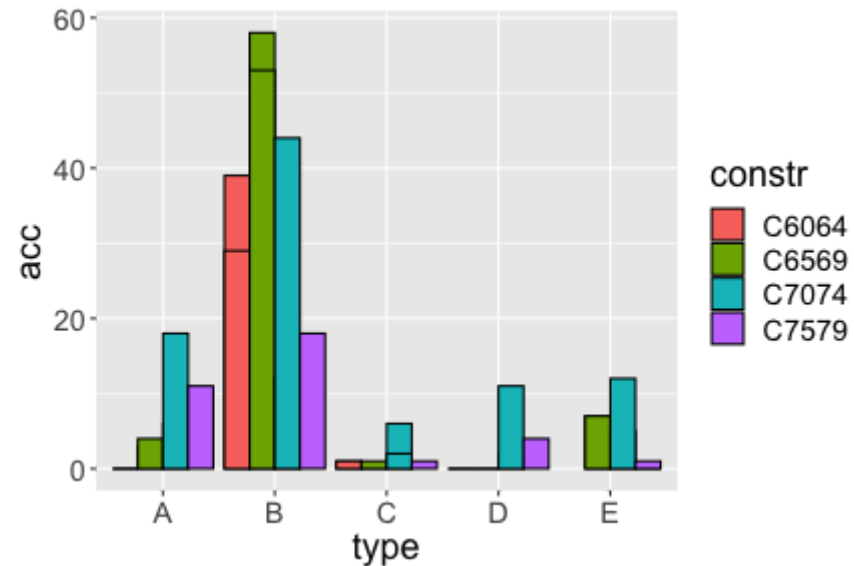
A stacked barplot with `geom_col()`

```
ggplot(Accident, aes(type, acc, fill = constr, group = opera  
  geom_col(color = "black", position = "stack"))
```



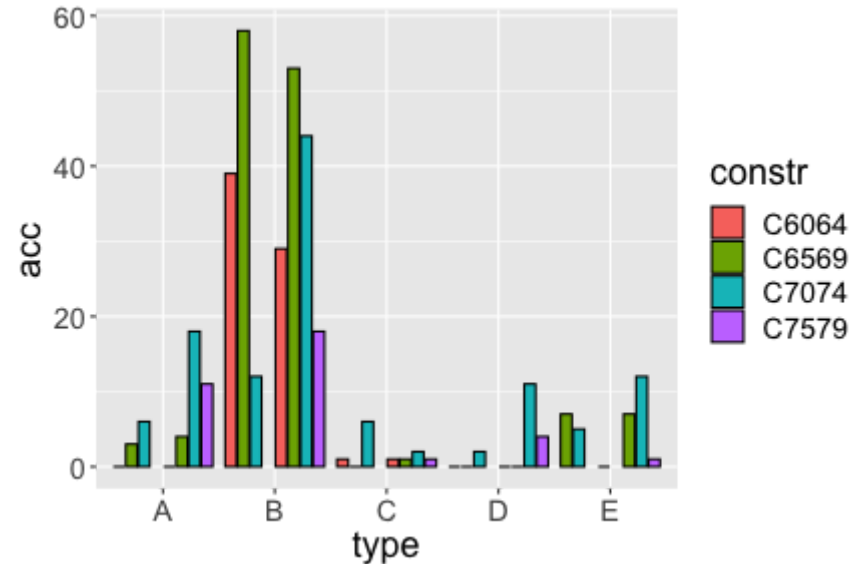
A grouped barplot with `geom_col()`

```
ggplot(Accident, aes(type, acc, fill = constr)) +  
  geom_col(color = "black", position = "dodge")
```



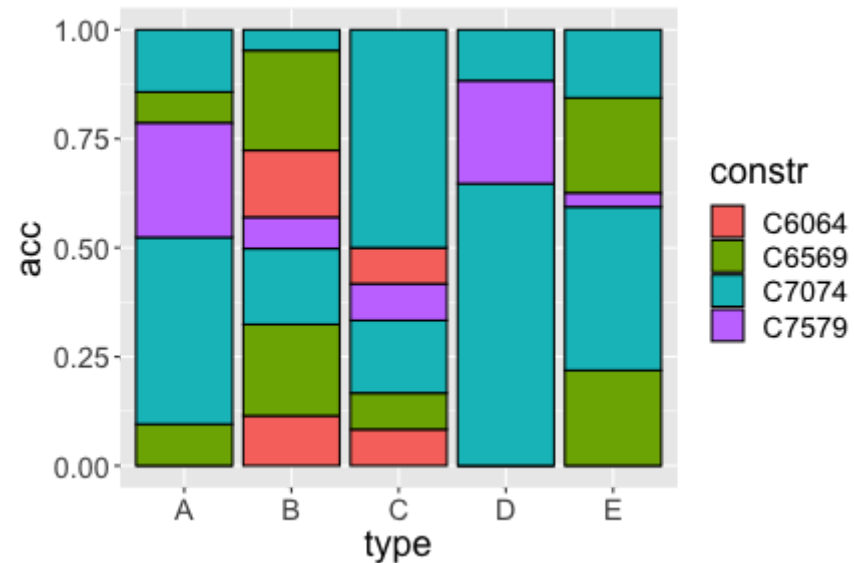
Another grouped barplot with `geom_col()`

```
ggplot(Accident, aes(type, acc, fill = constr, group = operation)) +  
  geom_col(color = "black", position = "dodge2")
```



Stacked percentage barplot with geom_col()

```
ggplot(Accident, aes(type, acc, fill = constr, group = operation)) +  
  geom_col(color = "black", position = "fill")
```



Illustrative data diamonds

- Let's have a look at the diamonds data which contains information about different attributes of diamonds

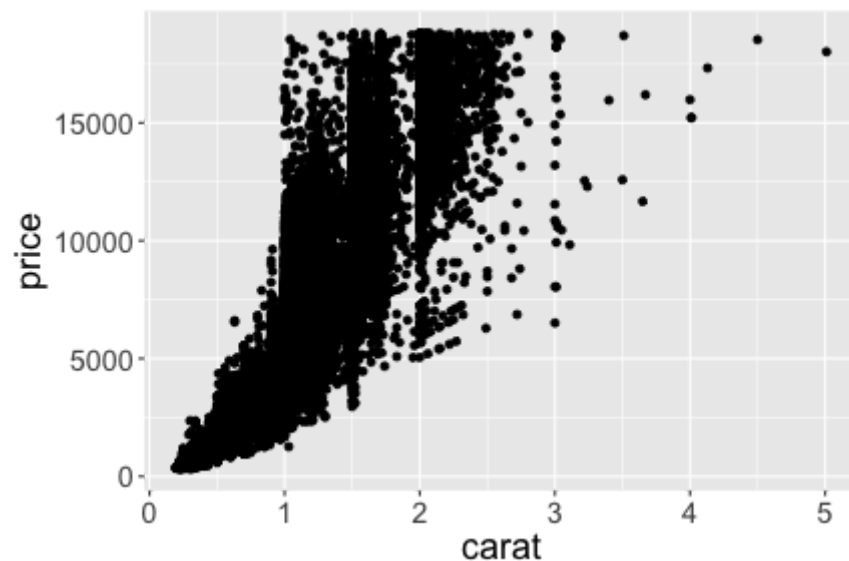
```
data(diamonds, package = "ggplot2")  
diamonds
```

```
## # A tibble: 53,940 × 10
```

```
##      carat cut      color clarity depth table price      x  
##      <dbl> <ord>      <ord> <ord>    <dbl> <dbl> <int> <dbl>  
##  1  0.23 Ideal      E      SI2     61.5    55    326  3.95  
##  2  0.21 Premium    E      SI1     59.8    61    326  3.89  
##  3  0.23 Good      E      VS1     56.9    65    327  4.05  
##  4  0.29 Premium    I      VS2     62.4    58
```

A scatterplot with `geom_point()`

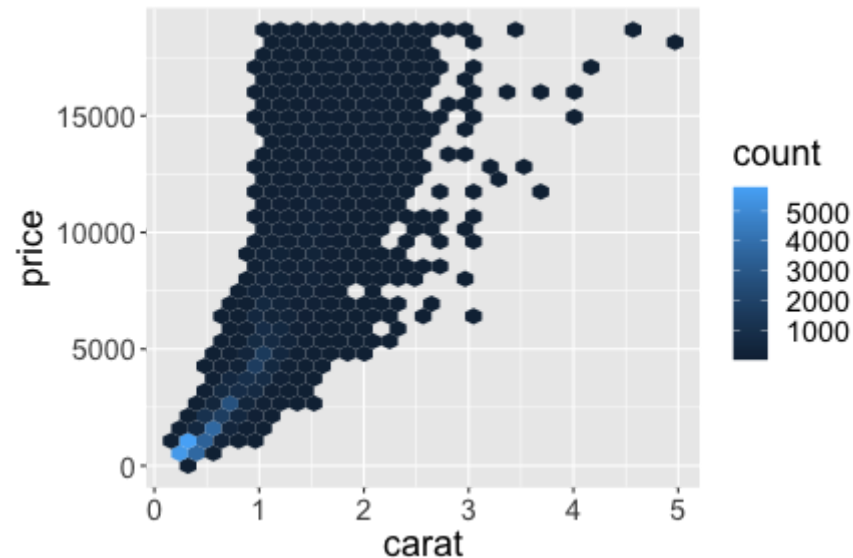
```
ggplot(data = diamonds,  
       mapping = aes(x = carat, y = price)) +  
  geom_point()
```



- Scatter plot of price vs carat of diamonds
- Each point corresponds to a diamond

A hexagonal 2D heatmap with `geom_hex()`

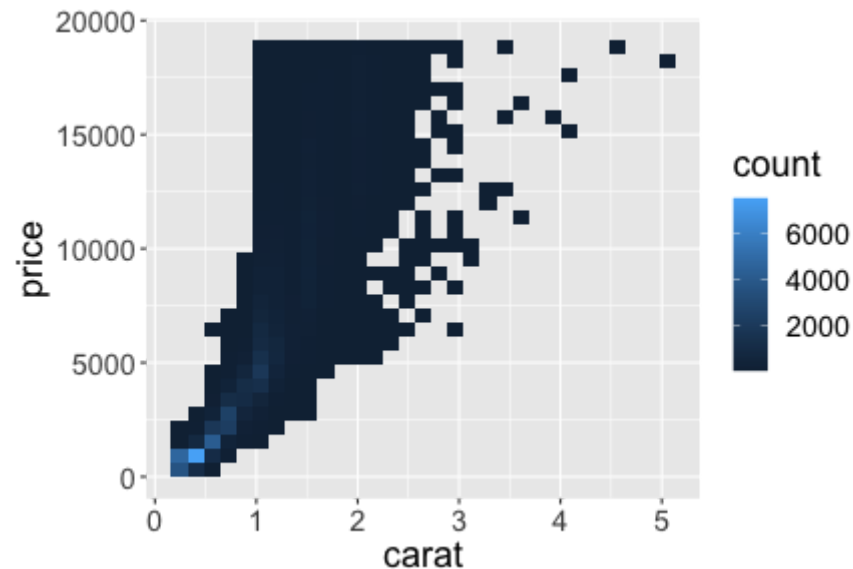
```
ggplot(data = diamonds,  
       mapping = aes(x = carat, y = price)) +  
  geom_hex()
```



- A hexagon shows the count of observations within the region.

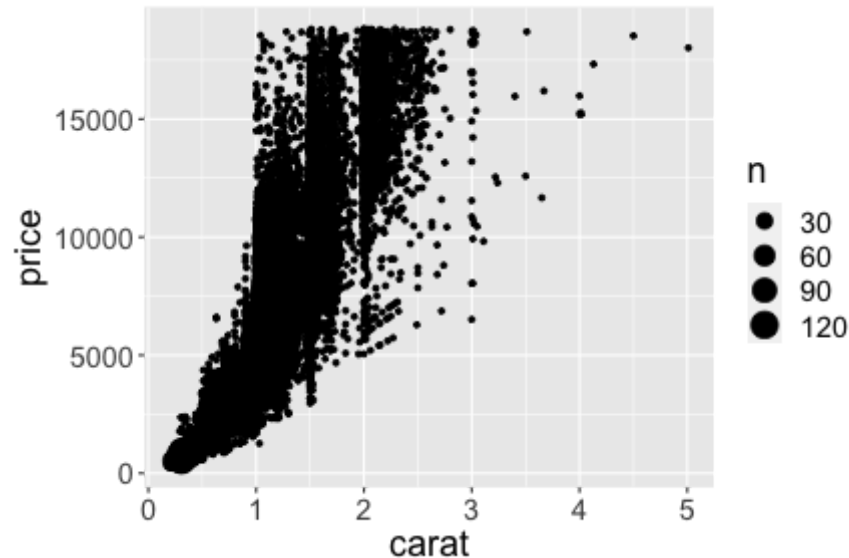
A rectangular 2D heatmap with `geom_bin_2d()`

```
ggplot(data = diamonds,  
       mapping = aes(x = carat, y = price)) +  
  geom_bin_2d()
```



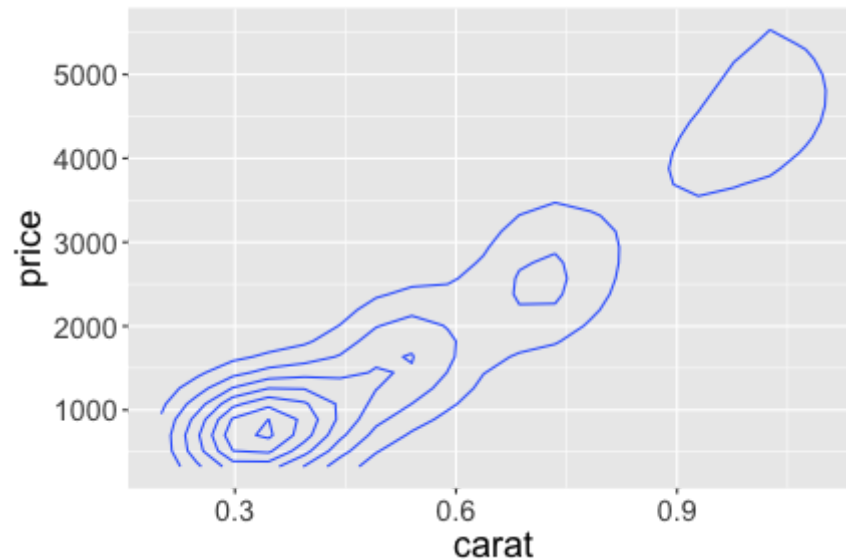
Count of overlapping points with `geom_count()`

```
ggplot(data = diamonds,  
       mapping = aes(x = carat, y = price)) +  
  geom_count()
```



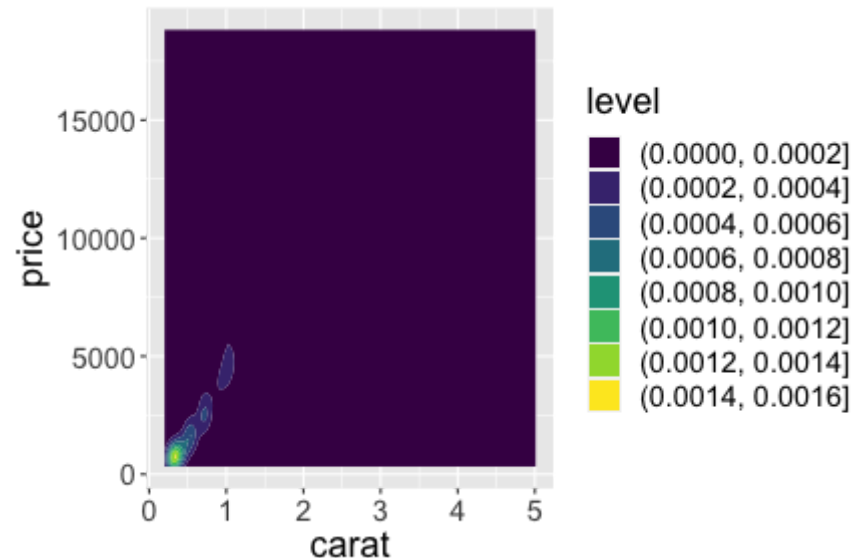
Contour of a 2D density with `geom_density_2d()`

```
ggplot(data = diamonds,  
       mapping = aes(x = carat, y = price)) +  
  geom_density_2d()
```



Contour of a 2D density with `geom_density_2d_filled()`

```
ggplot(data = diamonds,  
       mapping = aes(x = carat, y = price)) +  
  geom_density_2d_filled()
```



Illustrative data volcano

- The data contains topographic information on Auckland's Maunga Whau Volcano

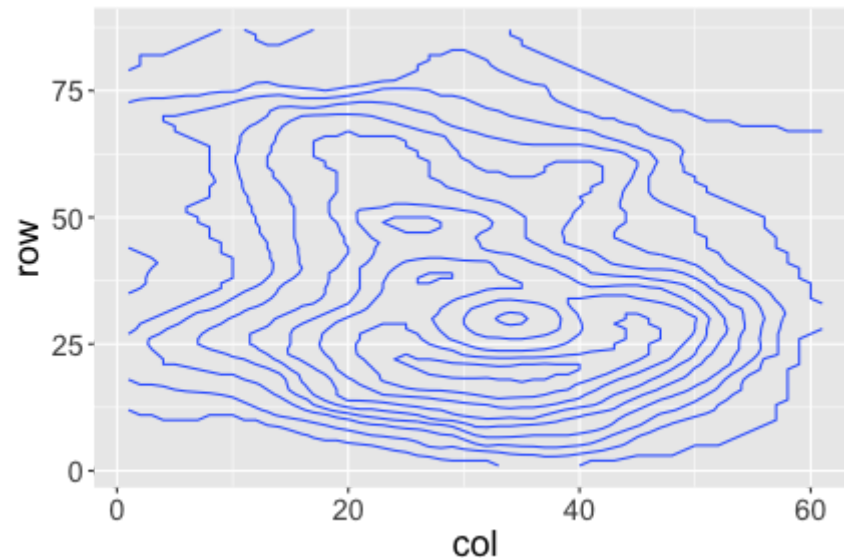
```
volcanod <- data.frame(row = as.vector(row(volcano)),  
                        col = as.vector(col(volcano)),  
                        value = as.vector(volcano))
```

volcanod

```
##      row col value  
## 1      1  1  100  
## 2      2  1  101  
## 3      3  1  102
```

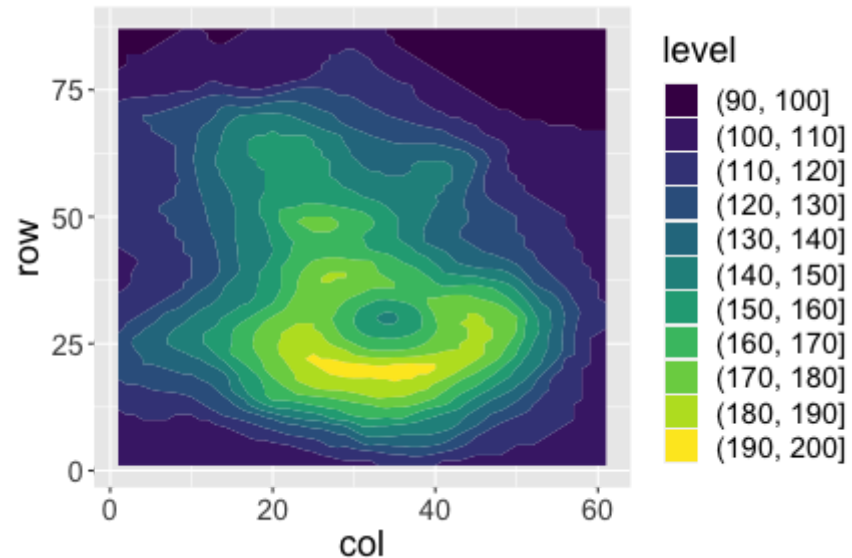
A 2D contour plot with `geom_contour()`

```
ggplot(data = volcanod,  
       mapping = aes(x = col, y = row, z = value)) +  
  geom_contour()
```



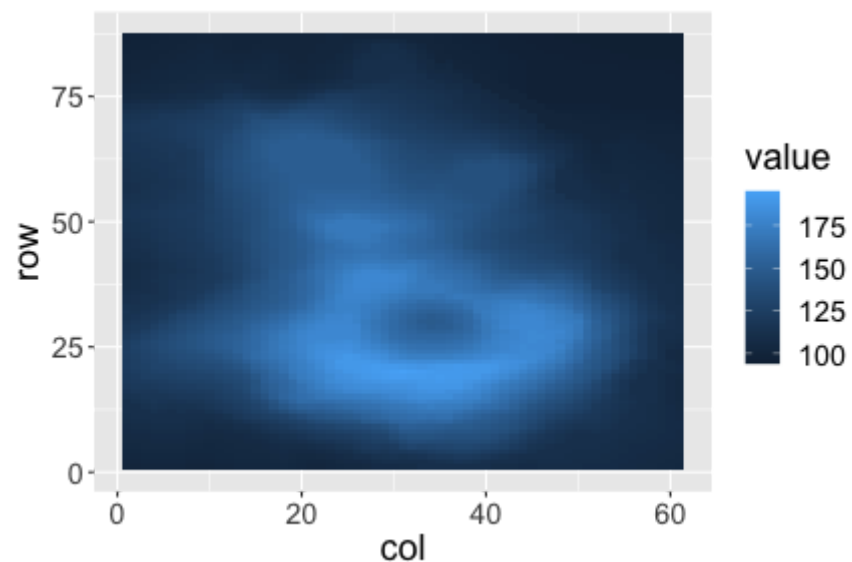
A 2D contour plot with `geom_contour_filled()`

```
ggplot(data = volcanod,  
       mapping = aes(x = col, y = row, z = value)) +  
  geom_contour_filled()
```



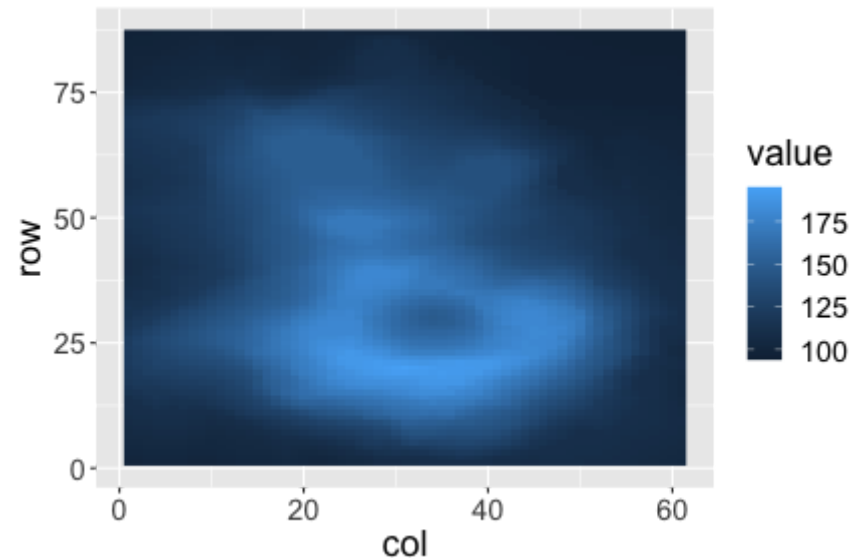
Tile plots with `geom_tile()`

```
ggplot(volcanod, aes(col, row, fill = value)) +  
  geom_tile()
```



Raster plots with `geom_raster()`

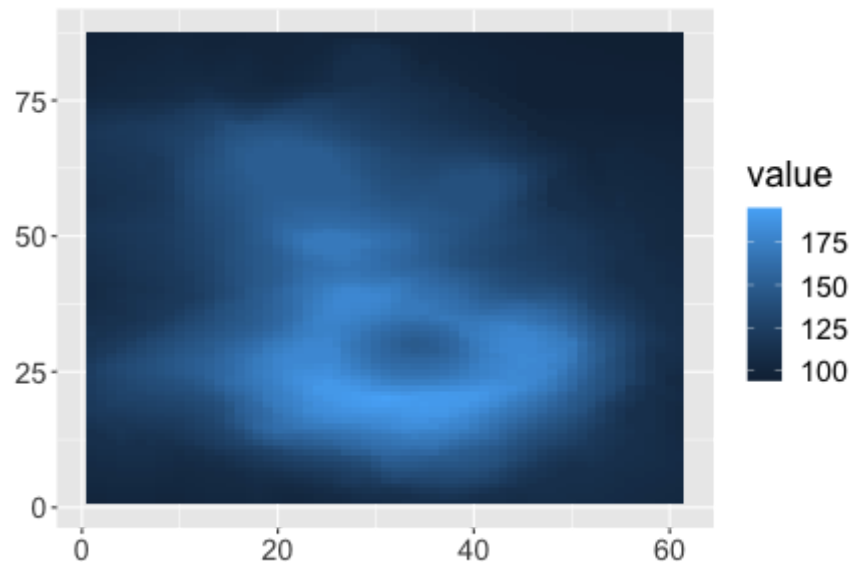
```
ggplot(volcanod, aes(col, row, fill = value)) +  
  geom_raster()
```



- A high performance special case of `geom_tile()` for when the tiles are the same size

Rectangular plots with `geom_rect()`

```
ggplot(volcanod, aes(xmin = col - 0.5, xmax = col + 0.5,  
                    ymin = row - 0.5, ymax = row + 0.5, fill = value))  
  geom_rect()
```



- A reparameterised version of `geom_tile()`

Illustrative data economics

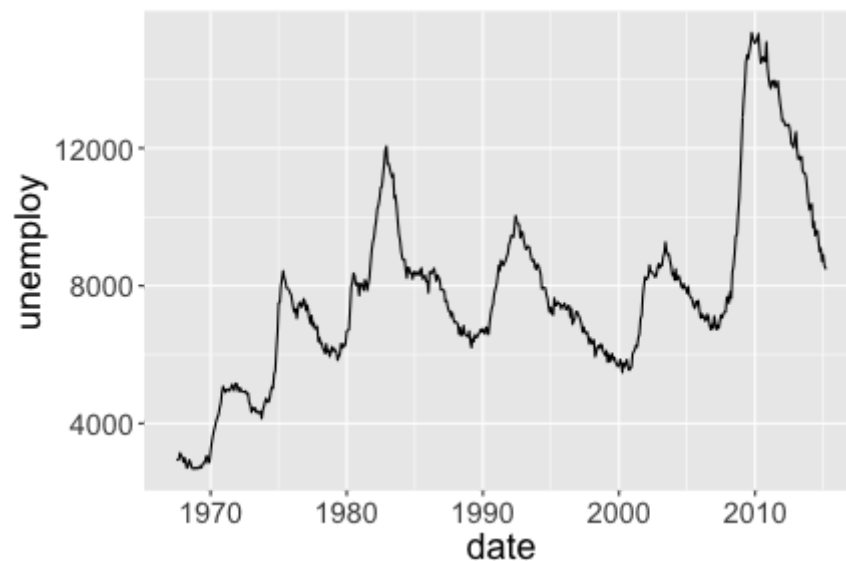
```
data(economics, package = "ggplot2")  
economics
```

```
## # A tibble: 574 × 6
```

##	date	pce	pop	psavert	uempmed	unemploy
##	<date>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>
##	1 1967-07-01	507.	198712	12.6	4.5	2944
##	2 1967-08-01	510.	198911	12.6	4.7	2945
##	3 1967-09-01	516.	199113	11.9	4.6	2958
##	4 1967-10-01	512.	199311	12.9	4.9	3143
##	5 1967-11-01	517.	199498	12.8	4.7	3066
##	6 1967-12-01	525.	199657	11.8	4.8	3018

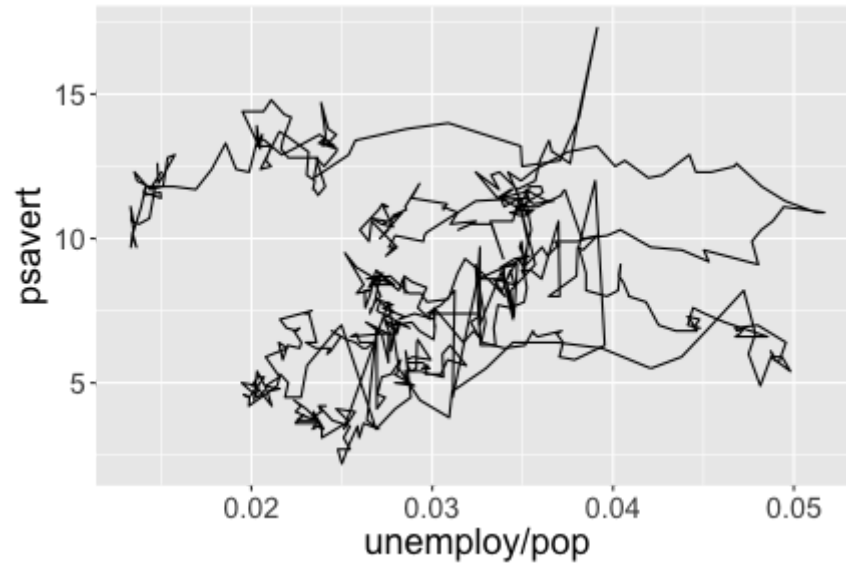
Line plot with `geom_line()`

```
ggplot(economics, aes(date, unemploy)) +  
  geom_line()
```



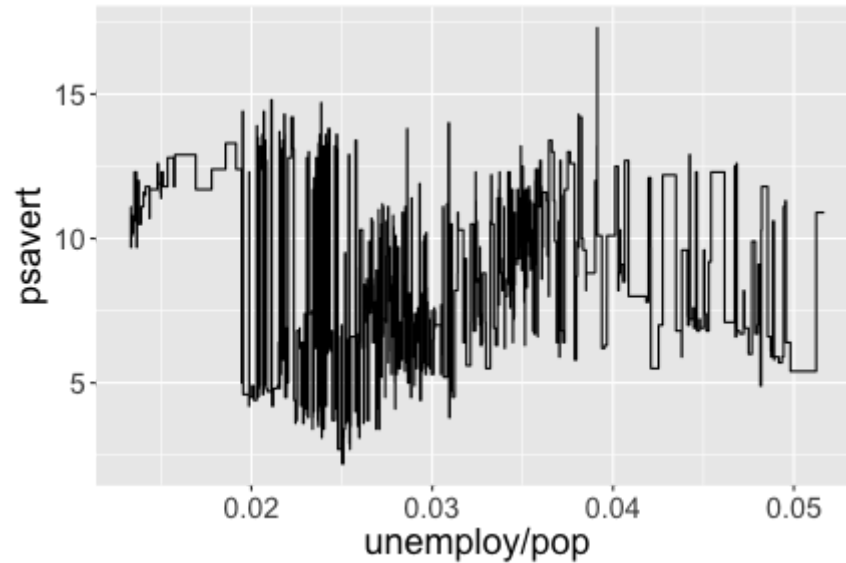
geom_path()

```
ggplot(economics, aes(unemploy/pop, psavert)) +  
  geom_path()
```



geom_step()

```
ggplot(economics, aes(unemploy/pop, psavert)) +  
  geom_step()
```



Drawing maps

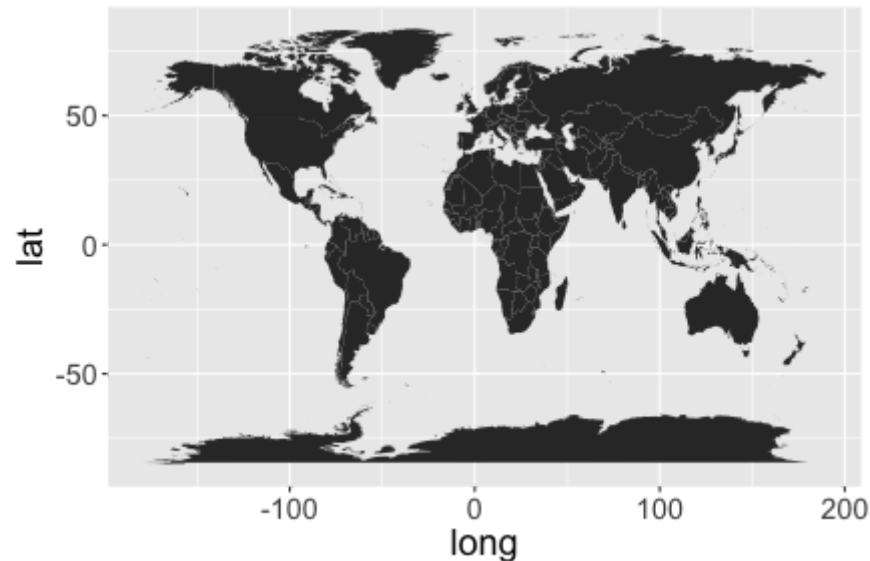
- Drawing maps require the map data

```
world <- map_data("world")  
world
```

##		<i>long</i>	<i>lat</i>	<i>group</i>	<i>order</i>	<i>region</i>	<i>subregion</i>
##	1	-69.89912	12.45200	1	1	Aruba	<NA>
##	2	-69.89571	12.42300	1	2	Aruba	<NA>
##	3	-69.94219	12.43853	1	3	Aruba	<NA>
##	4	-70.00415	12.50049	1	4	Aruba	<NA>
##	5	-70.06612	12.54697	1	5	Aruba	<NA>
##	6	-70.05088	12.59707	1	6	Aruba	<NA>
##	7	-70.03511	12.61411	1	7	Aruba	<NA>

Drawing world map with `geom_polygon()`

```
world <- map_data("world")  
ggplot(world, aes(long, lat, group = group)) +  
  geom_polygon()
```



Other layers

- There are more layers in `ggplot`!
- There are layers also beginning with `stat_`.
- There are also many extension packages to draw more plots:
<https://exts.ggplot2.tidyverse.org/gallery/>
- It is normal to google and copy-and-paste code that you find on the web.
- `ggplot2` is also **continually updated** so some new features may be added in the future.

Session Information

```
devtools::session_info()
```

```
## - Session info   
## hash: keycap: 3, vampire: light skin tone, man: medium-dark skin tone
##
## setting value
## version R version 4.1.2 (2021-11-01)
## os macOS Big Sur 11.6.2
## system x86_64, darwin17.0
## ui RStudio
## language (EN)
## collate en_AU.UTF-8
## ctype en_AU.UTF-8
## tz Australia/Melbourne
## date 2022-02-20
```

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