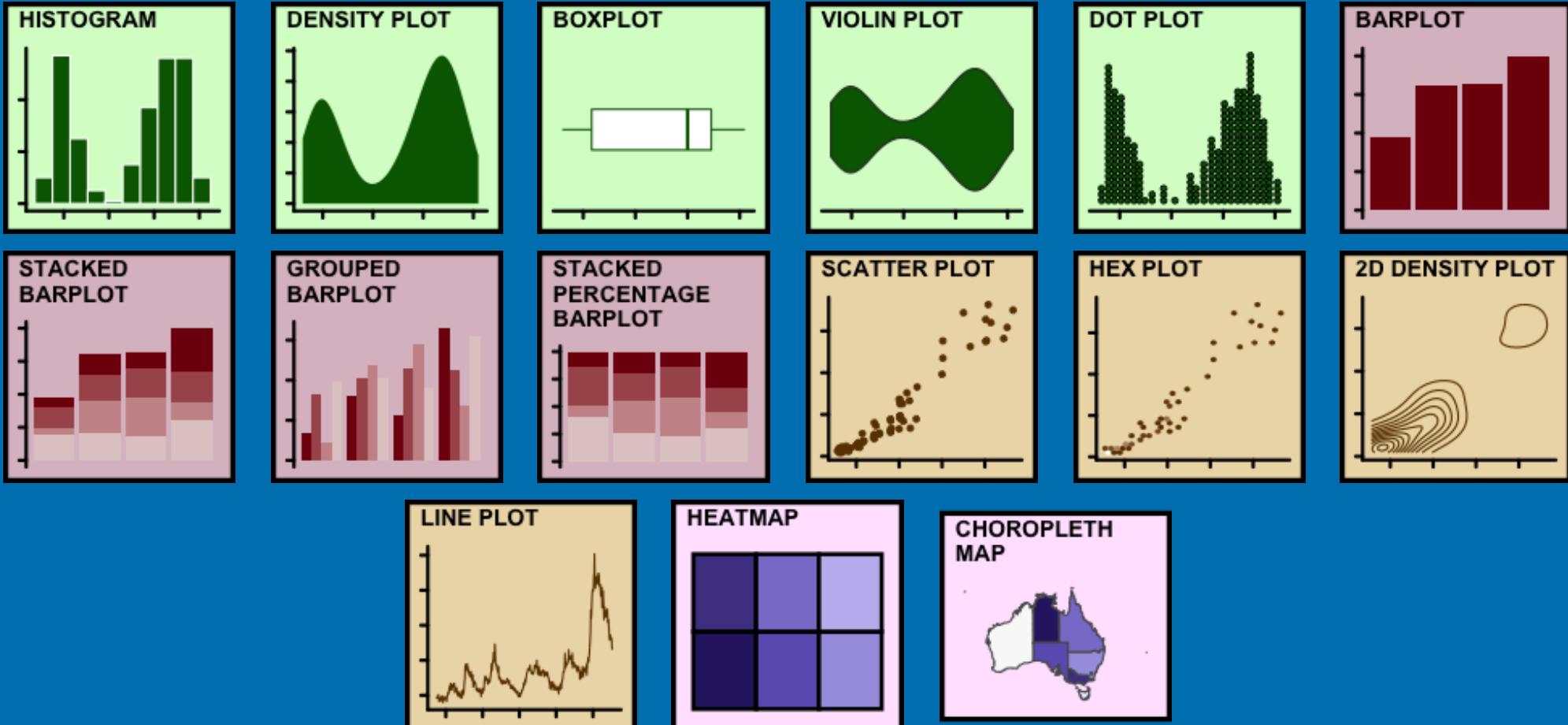




# Data Visualisation with R

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:
```

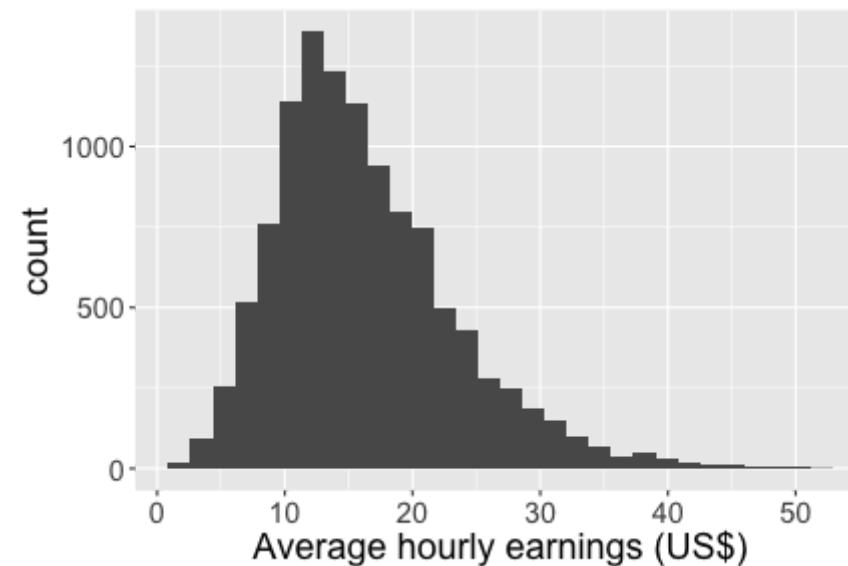
```
## $ year: int 1992 1992 1992 1992 1992 1992 1992 1992 1992 1992 ...
```

```
## $ 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()
```

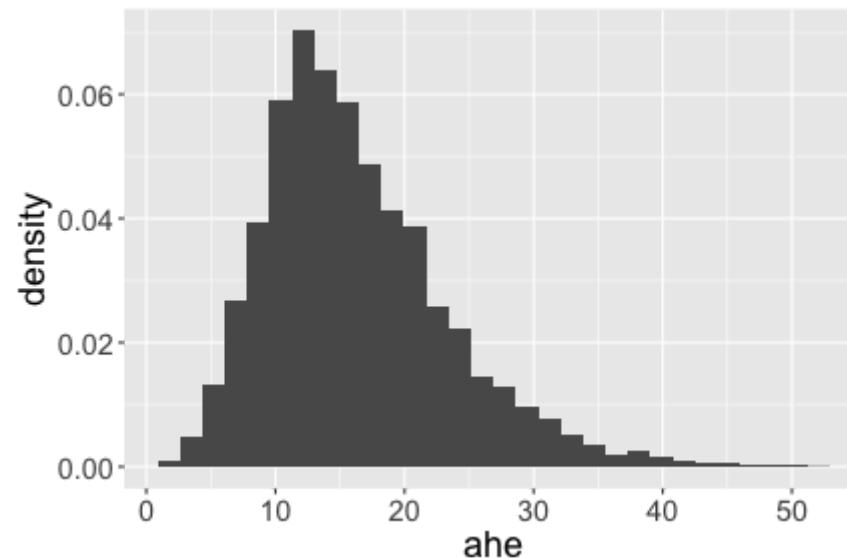
```
layer_data(g, 1)
```

##	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

▶ Back to catalogue

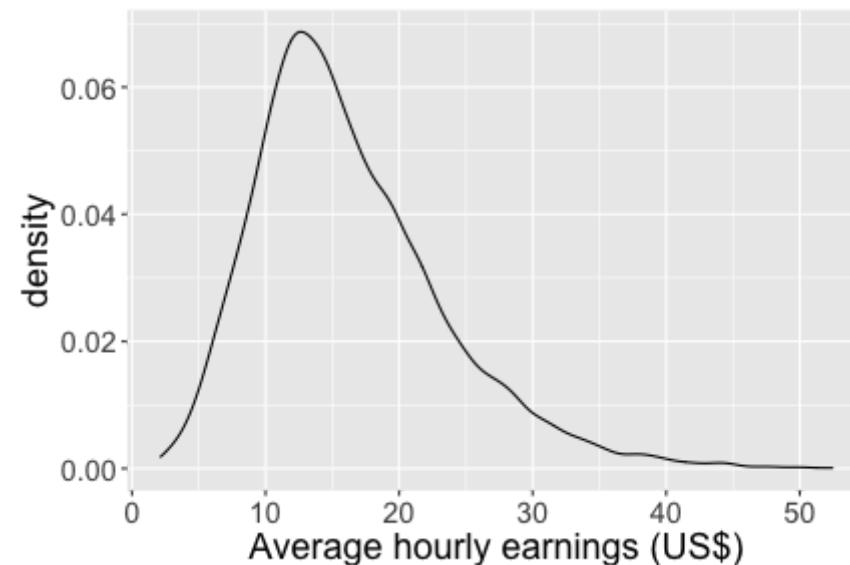
# Layer data

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



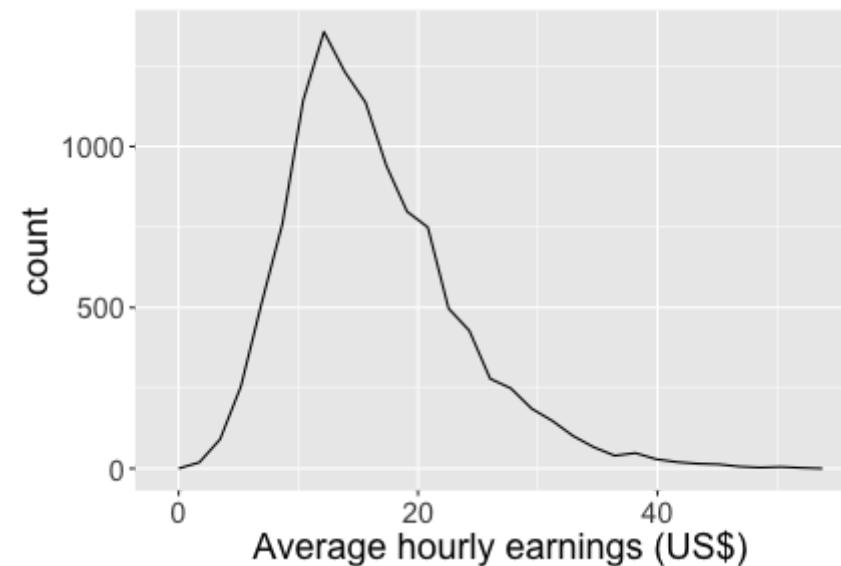
# A density plot with geom\_density()

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



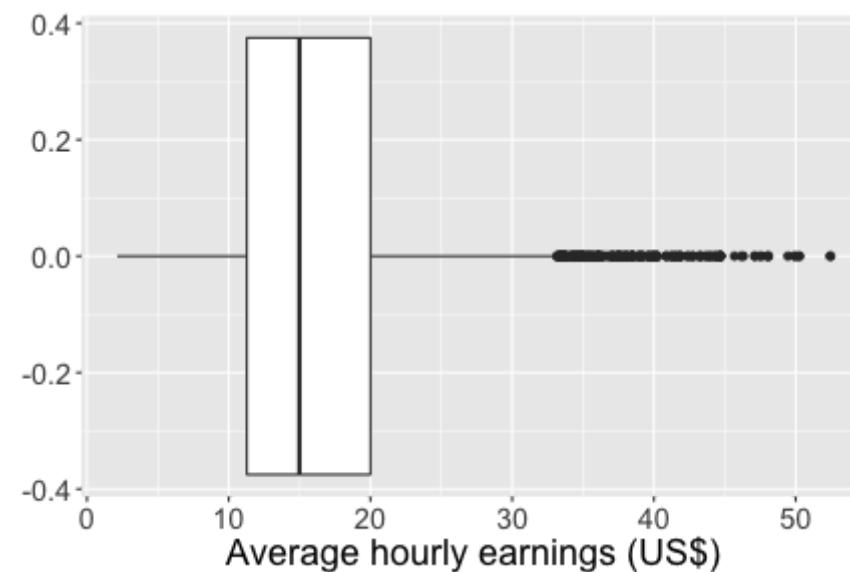
# A frequency polygon with geom\_freqpoly()

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



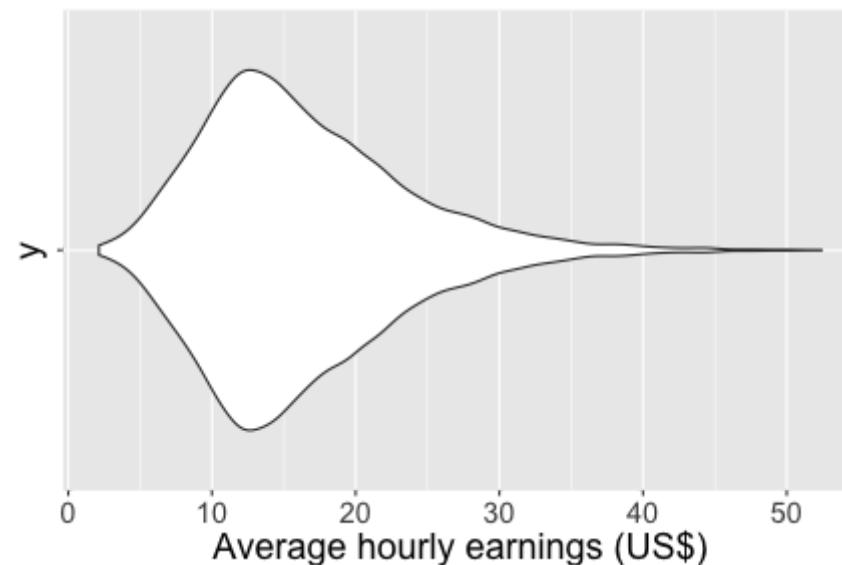
# A boxplot with geom\_boxplot()

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



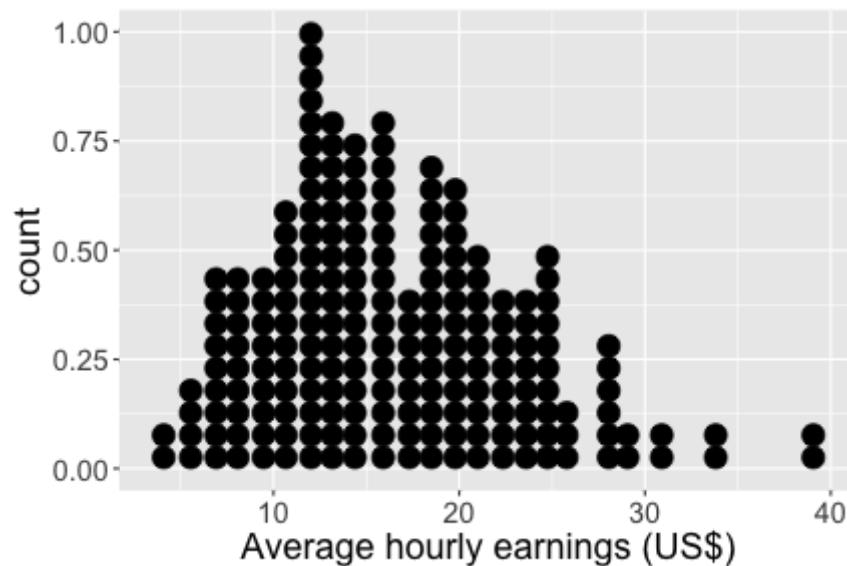
# A violin plot with geom\_violin()

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



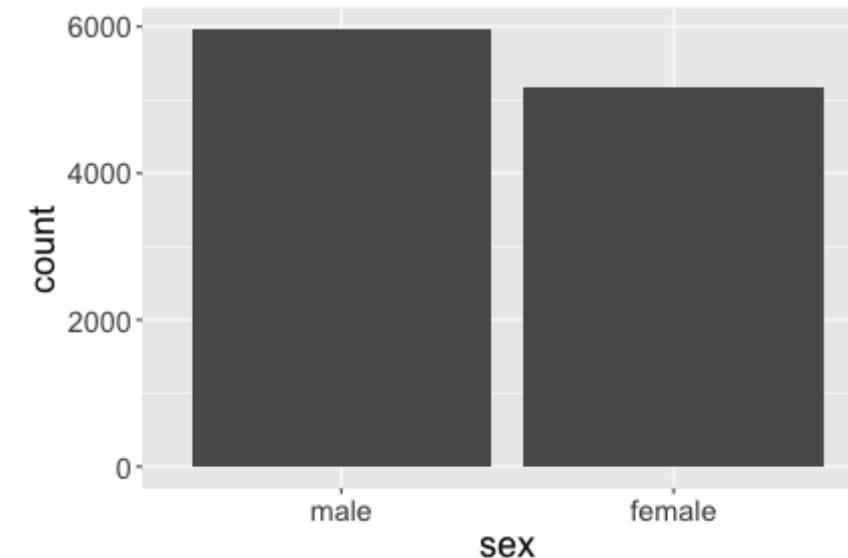
# A dotplot with geom\_dotplot()

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



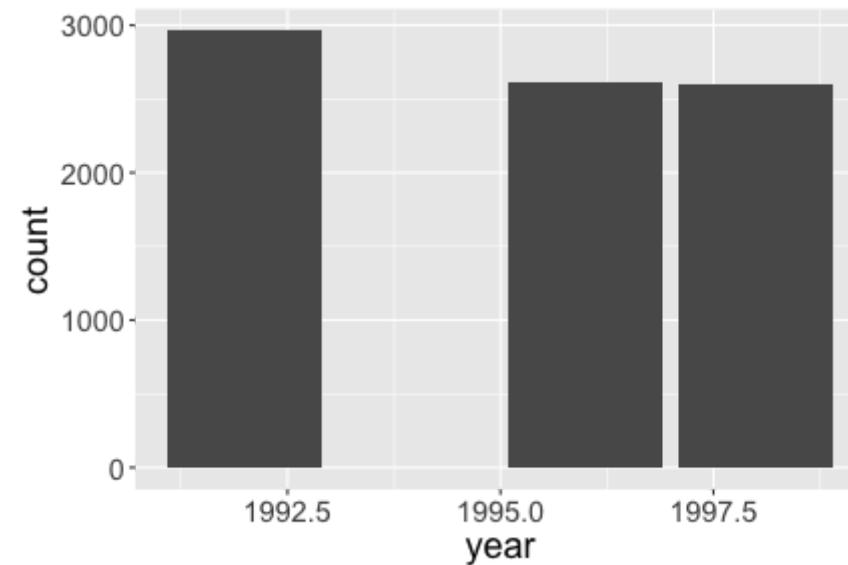
# A barplot with geom\_bar() with categorical variables

```
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()
```



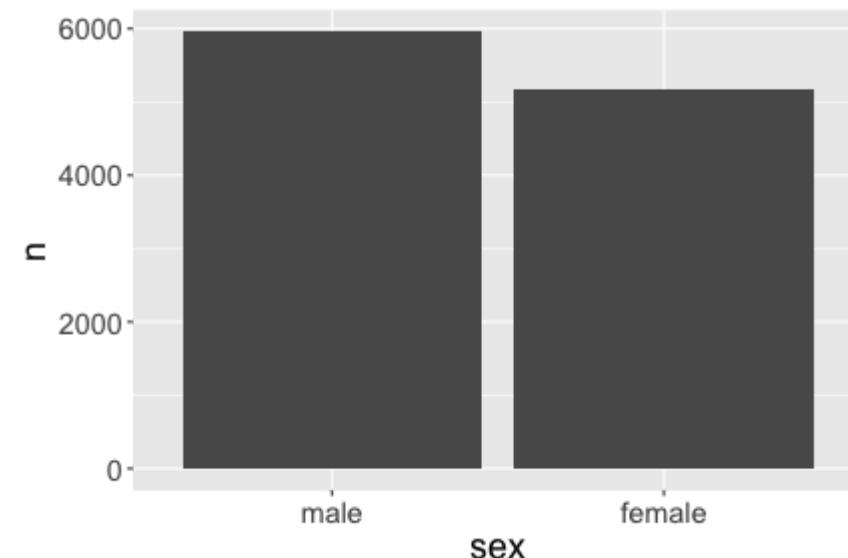
# A barplot with geom\_col()

```
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()



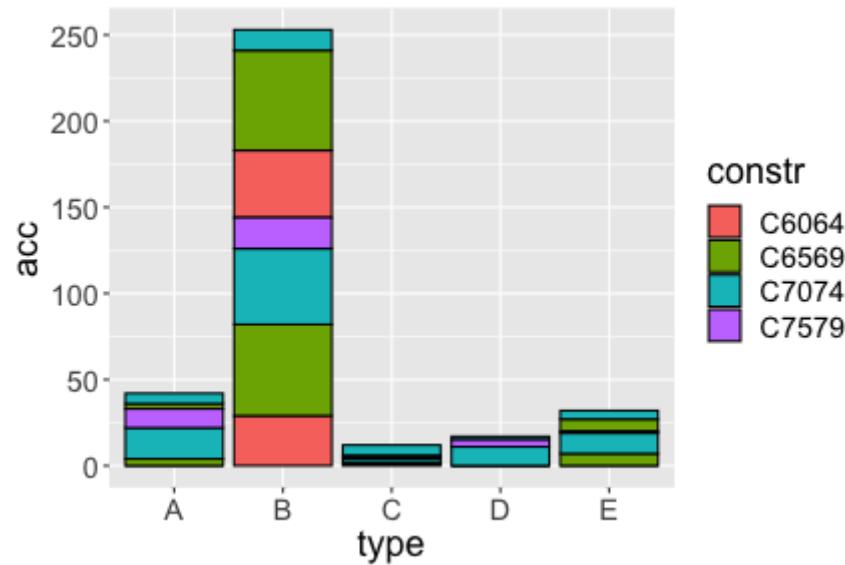
# Illustrative data 🚢 ship accidents

```
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 ...
```

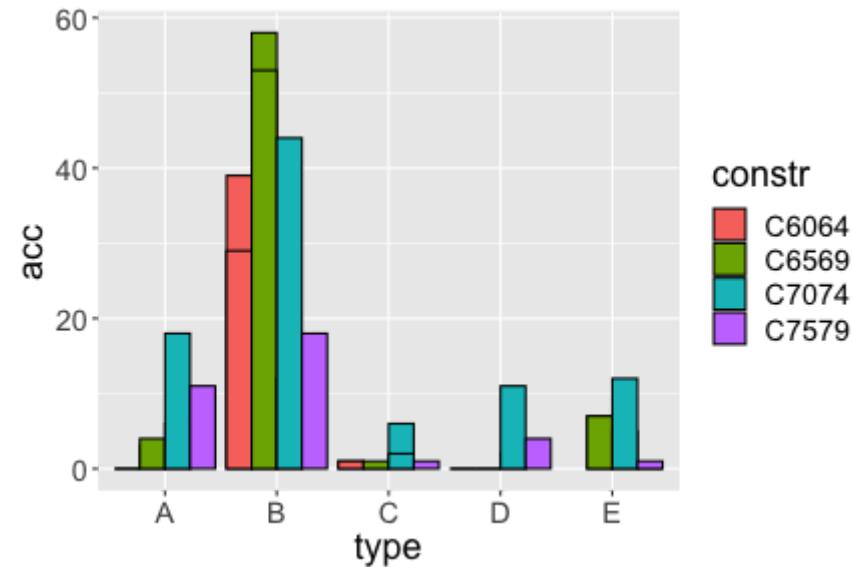
# A stacked barplot with geom\_col()

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



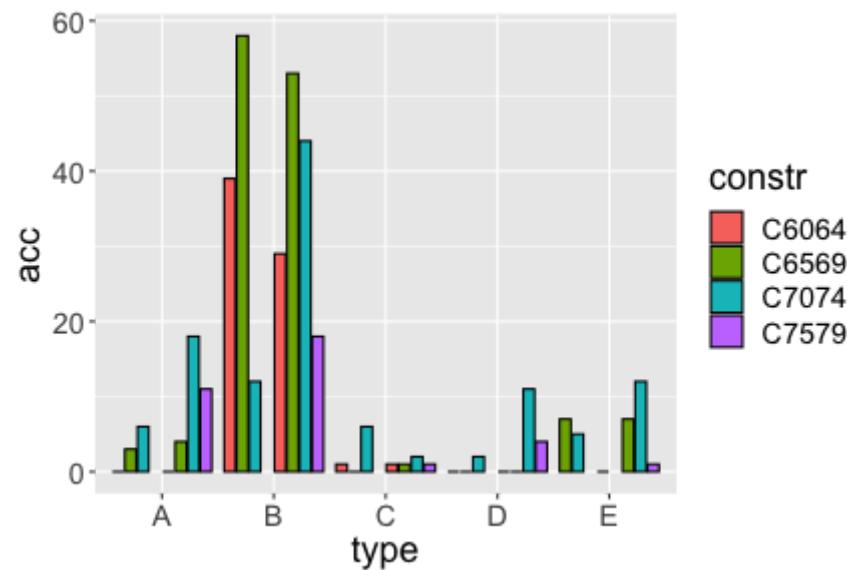
# A grouped barplot with geom\_col()

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



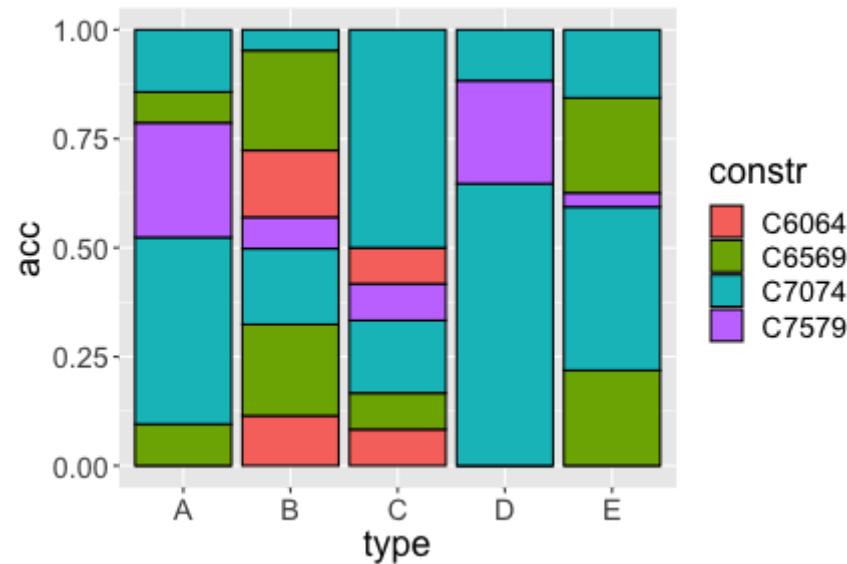
# Another grouped barplot with geom\_col()

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



# Stacked percentage barplot with geom\_col()

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



# Illustrative data 💎 diamonds

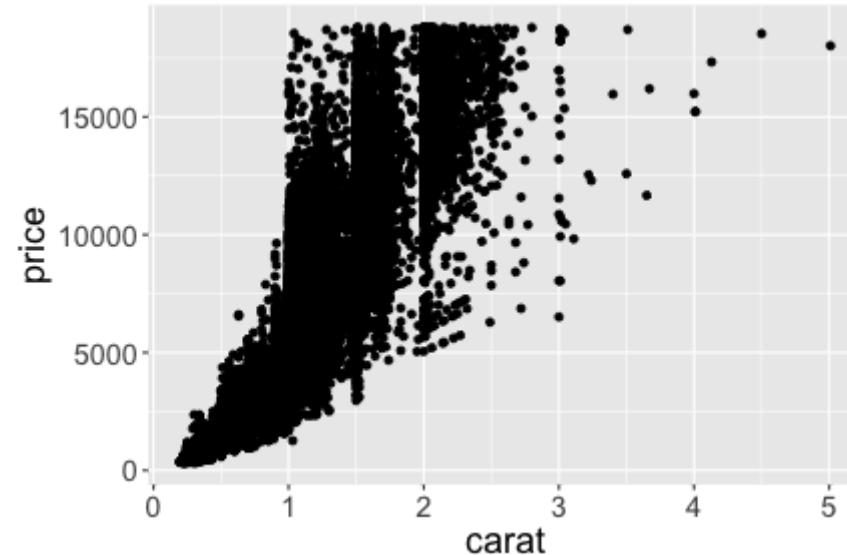
- Let's have a look at the diamonds data which contains information about different attributes of diamonds

```
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()

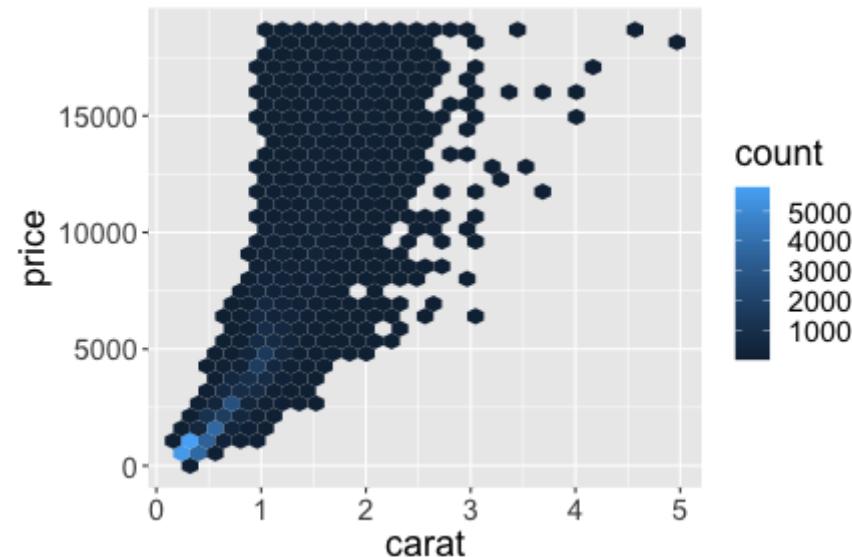
```
ggplot(data = diamonds,  
       mapping = aes(x = carat, y = price)) +  
geom_point()
```



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

# A hexagonal 2D heatmap with geom\_hex()

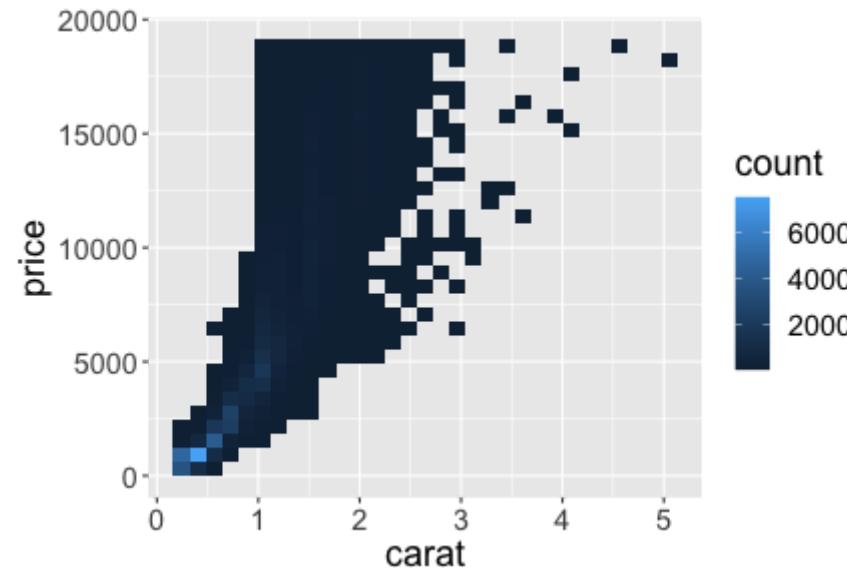
```
ggplot(data = diamonds,  
       mapping = aes(x = carat, y = price)) +  
  geom_hex()
```



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

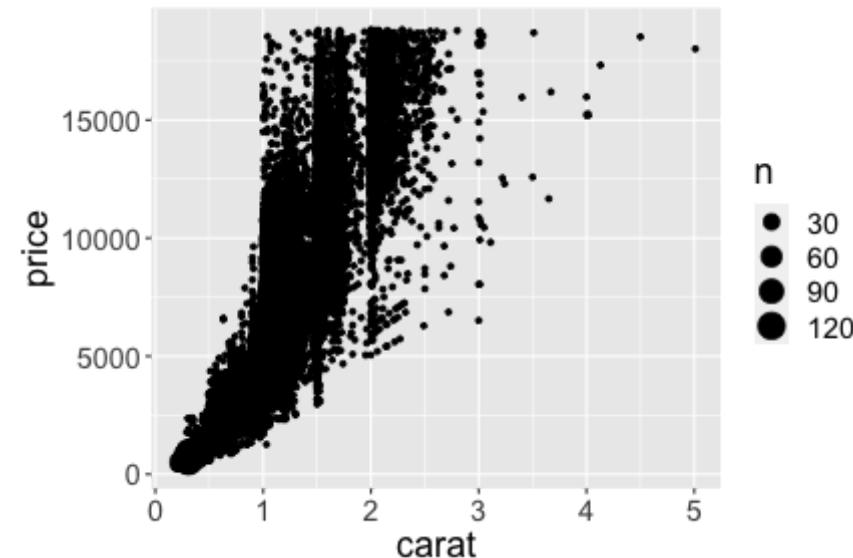
# A rectangular 2D heatmap with geom\_bin\_2d()

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



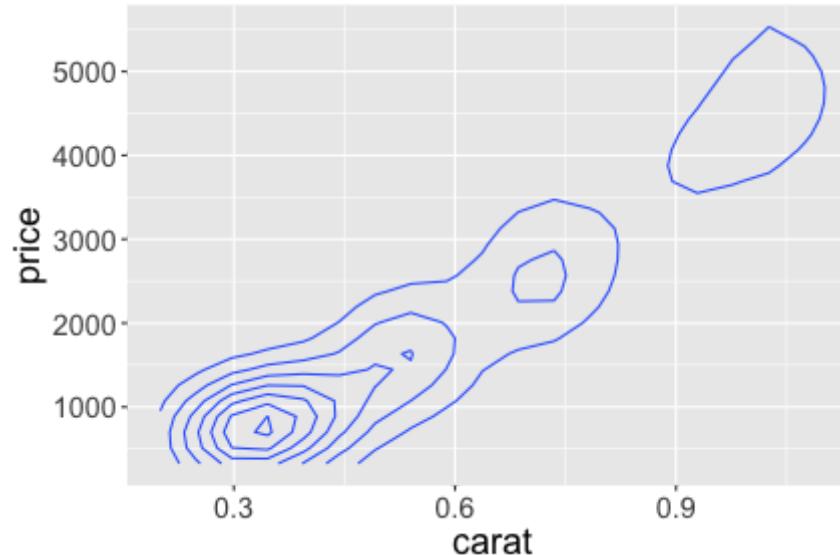
# Count of overlapping points with geom\_count()

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



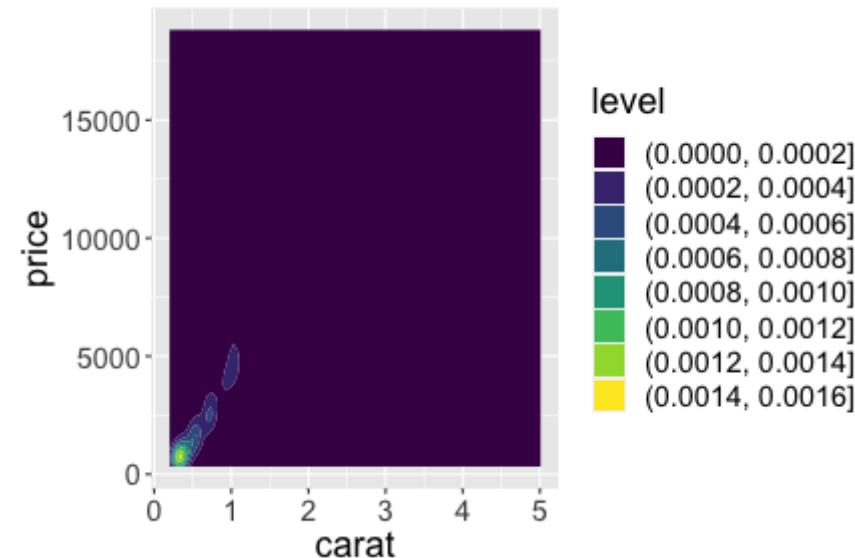
# Contour of a 2D density with geom\_density\_2d()

```
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

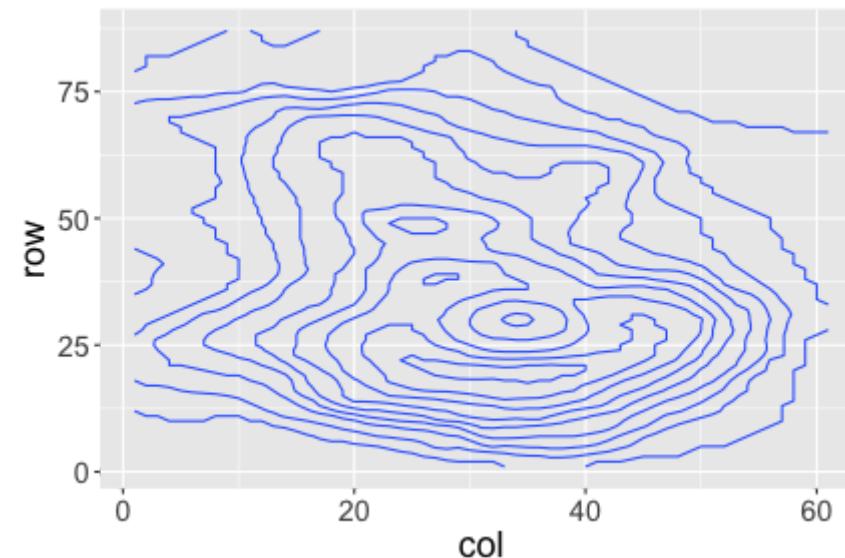
```
volcanod <- data.frame(row = as.vector(row(volcano)),  
                        col = as.vector(col(volcano)),  
                        value = as.vector(volcano))
```

```
volcanod
```

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

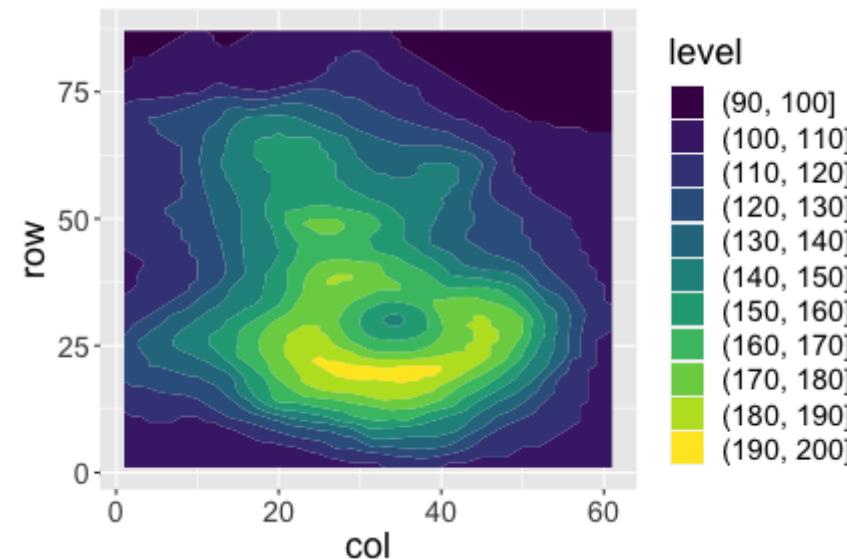
# A 2D contour plot with geom\_contour()

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



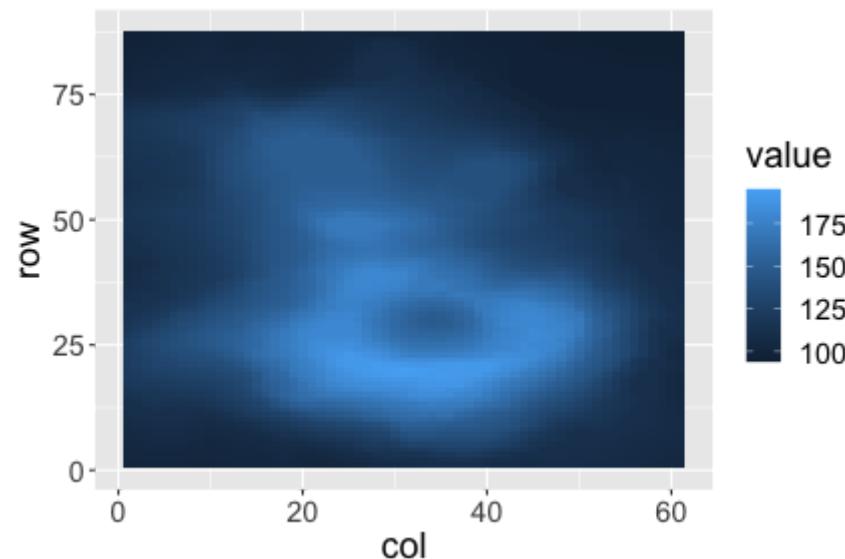
# A 2D contour plot with geom\_contour\_filled()

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



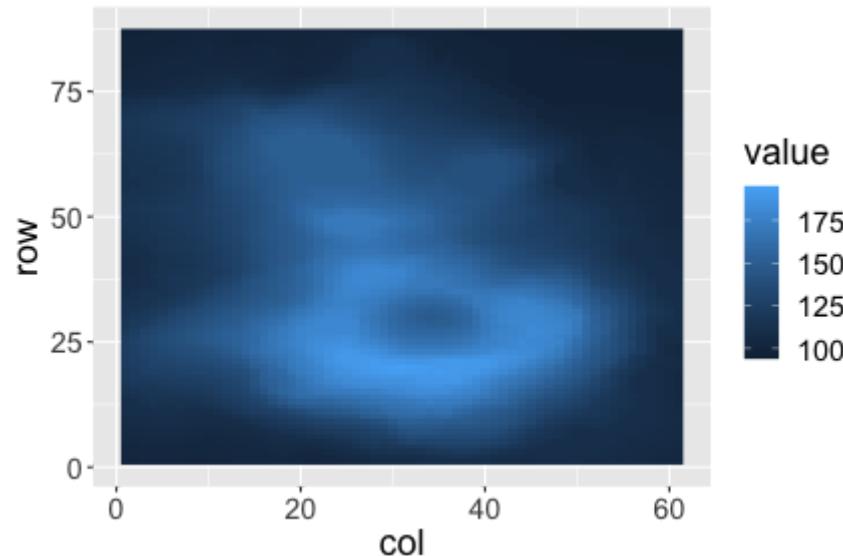
# Tile plots with geom\_tile()

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



# Raster plots with geom\_raster()

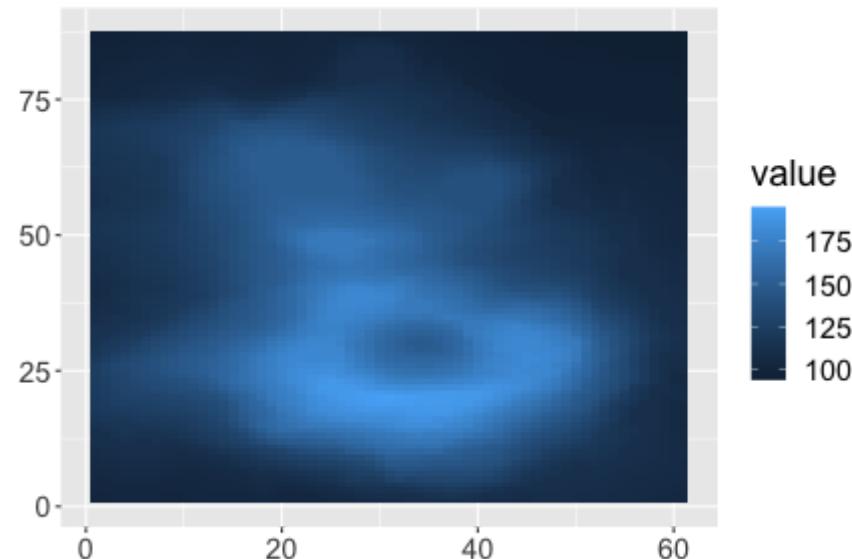
```
ggplot(volcanod, aes(col, row, fill = value)) +  
  geom_raster()
```



- 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, fil  
geom_rect()
```



- A reparameterised version of geom\_tile()

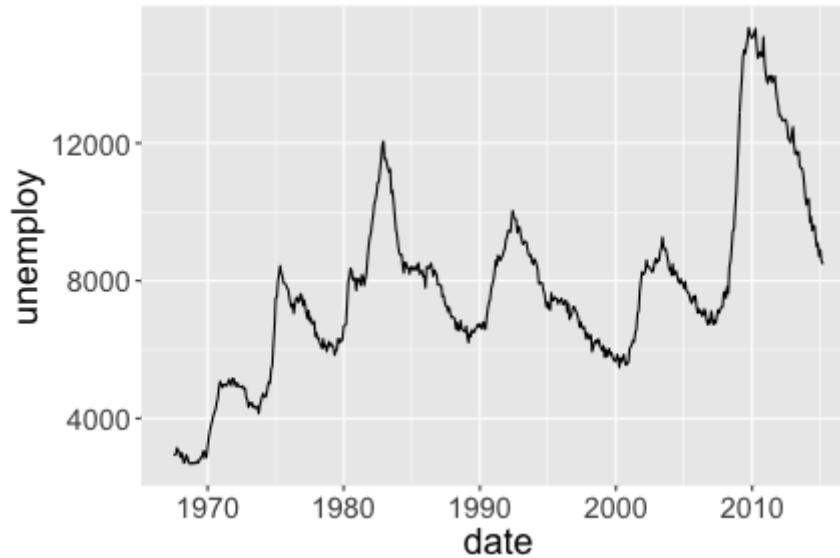
# Illustrative data economics

```
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       3066
```

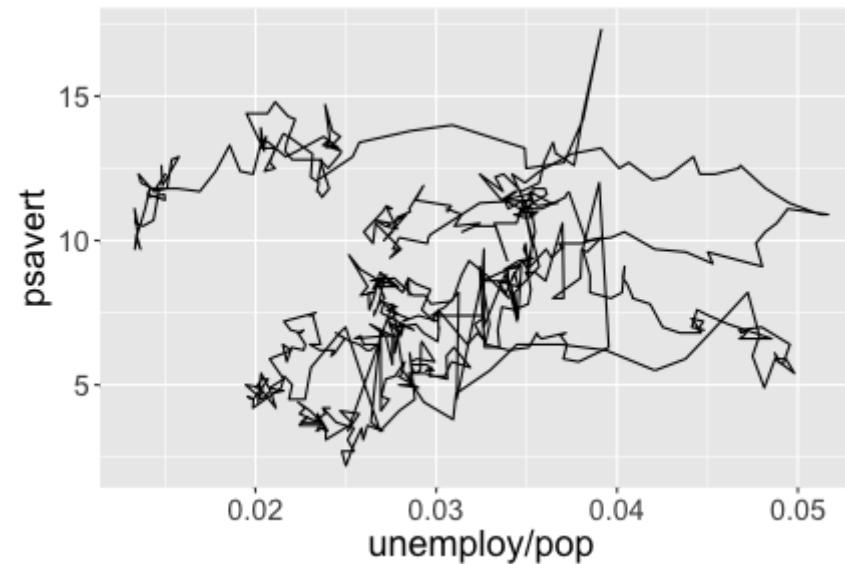
# Line plot with geom\_line()

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



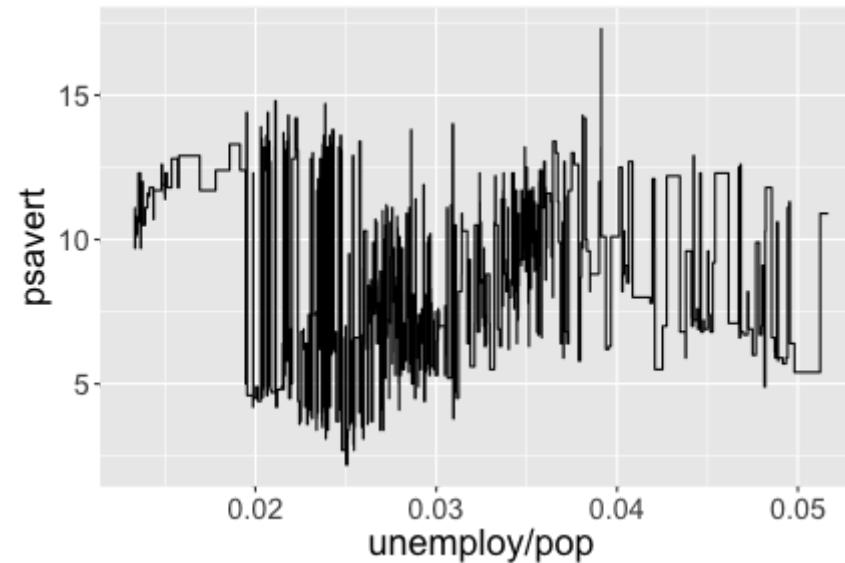
# geom\_path()

```
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

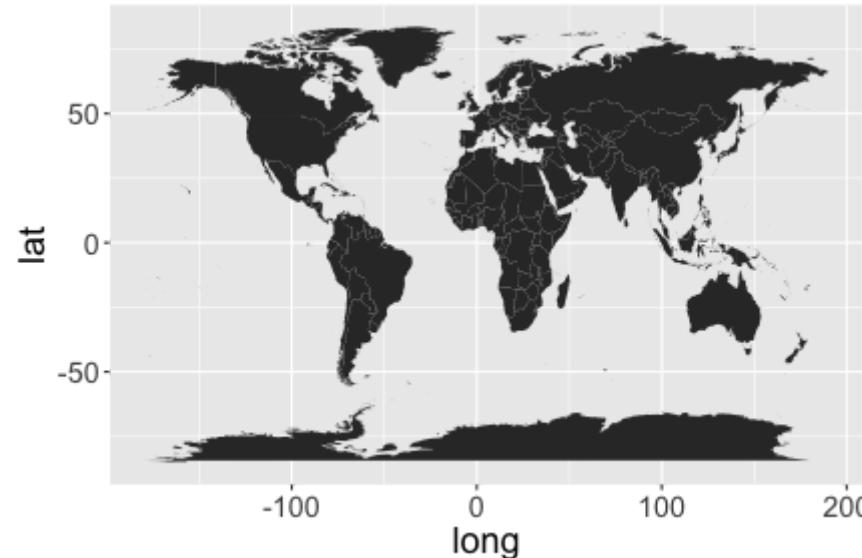
```
world <- map_data("world")
```

```
world
```

<i>##</i>	<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 3 🧑🏻‍💻 🧑🏿  
## 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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