

Graphics with ggplot2

STAT 220

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Layered Grammar of Graphics

- Essentials

- Data
- **Aesthetic mappings**
- **Geometric objects**

- Additional elements

- Facets
- **Coordinate system**
- **Statistical transformations**
- **Position adjustments**
- Scales
- Theme

Common **ggplot2** options

```
ggplot(data) +      # data
  <geometry_funs>(aes(<variables>)) + # aesthetic variable mapping
  <label_funs> +     # add context
  <facet_funs> +     # add facets (optional)
  <coordinate_funs> + # play with coords (optional)
  <scale_funs> +     # play with scales (optional)
  <theme_funs> # play with axes, colors, etc (optional)
```

- See the [Rstudio cheatsheets](#) for more details

Hate Crime and income inequality

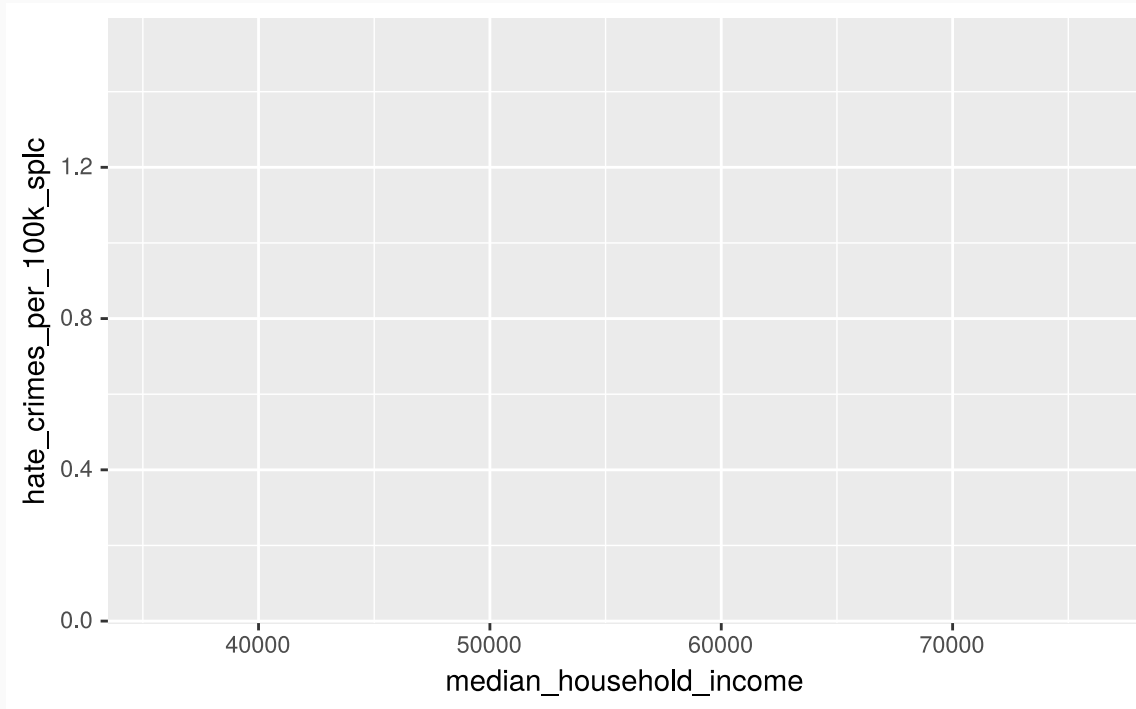
A FiveThirtyEight article published in 2017 claimed that higher rates of hate crimes were tied to greater income inequality.

```
glimpse(hate_crimes)
Rows: 51
Columns: 17
$ X               <int> 1, 2, 3, 4, 5, 6, 7, 8, 9, 10...
$ state           <chr> "Alabama", "Alaska", "Arizona..."
$ median_household_income <int> 42278, 67629, 49254, 44922, 6...
$ share_unemployed_seasonal <dbl> 0.060, 0.064, 0.063, 0.052, 0...
$ share_population_in_metro_areas <dbl> 0.64, 0.63, 0.90, 0.69, 0.97,...
$ share_population_with_high_school_degree <dbl> 0.821, 0.914, 0.842, 0.824, 0...
$ share_non_citizen <dbl> 0.02, 0.04, 0.10, 0.04, 0.13,...
$ share_white_poverty <dbl> 0.12, 0.06, 0.09, 0.12, 0.09,...
$ gini_index       <dbl> 0.472, 0.422, 0.455, 0.458, 0...
$ share_non_white   <dbl> 0.35, 0.42, 0.49, 0.26, 0.61,...
$ share_voters_voted_trump <dbl> 0.63, 0.53, 0.50, 0.60, 0.33,...
$ hate_crimes_per_100k_splc <dbl> 0.12583893, 0.14374012, 0.225...
$ avg_hatecrimes_per_100k_fbi <dbl> 1.8064105, 1.6567001, 3.41392...
$ state_code        <chr> "AL", "AK", "AZ", "AR", "CA",...
$ region            <chr> "South", "West", "West", "Sou...
$ division          <chr> "East South Central", "Pacifi...
$ support           <chr> "Trump", "Trump", "Split", "T...
```

Layering geoms

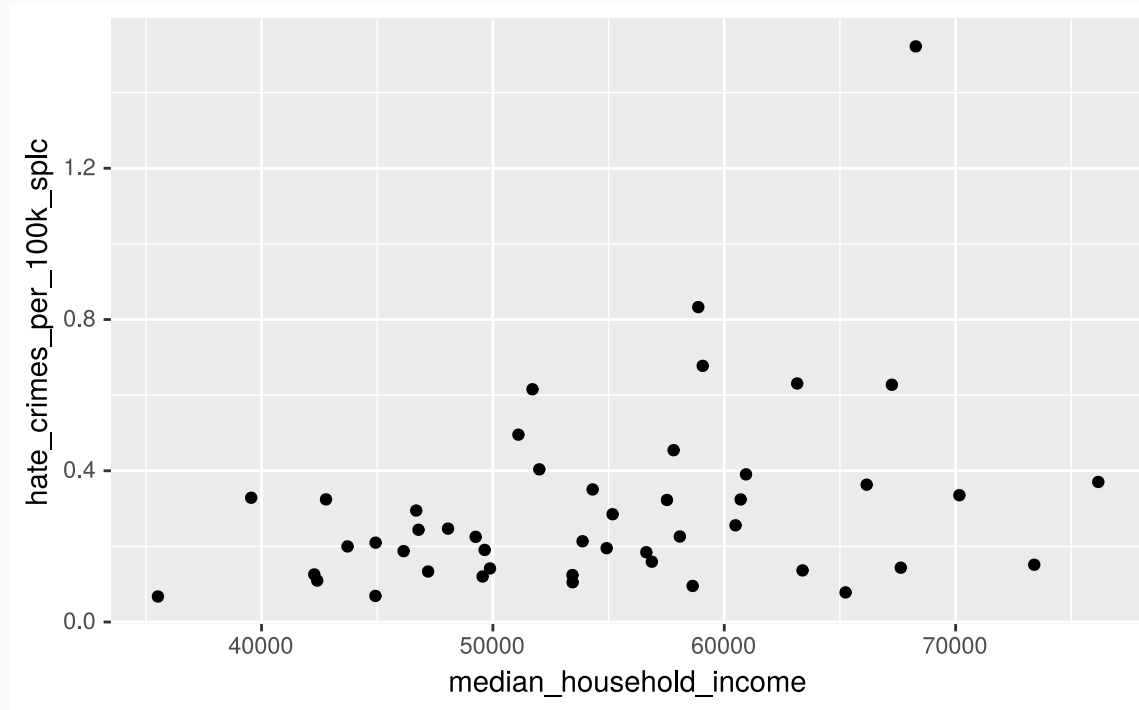
When you are iteratively building plots, it's useful to store the `base plot` as an object

```
base <- ggplot(hate_crimes, aes(x=median_household_income, y=hate_crimes_per_100k_splc))  
base
```



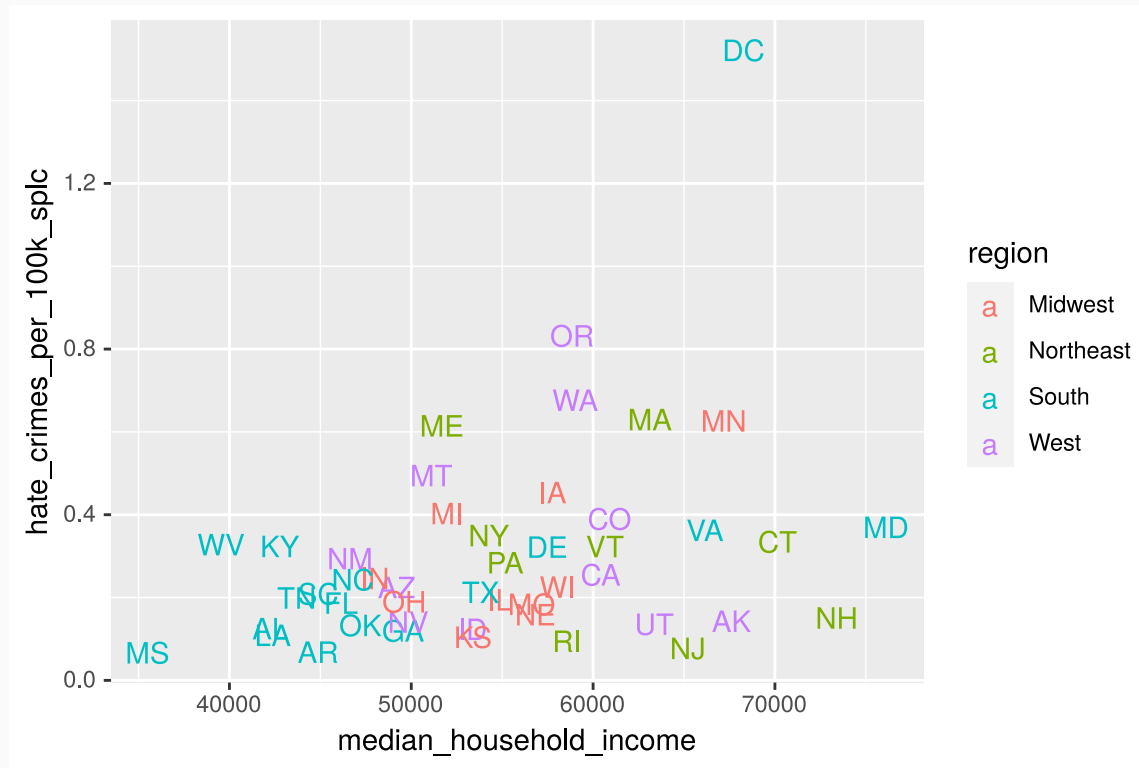
Layering geoms

```
base +  
  geom_point()
```



A better plot

```
base +  
  geom_text(aes(label=state_code, color=region))
```



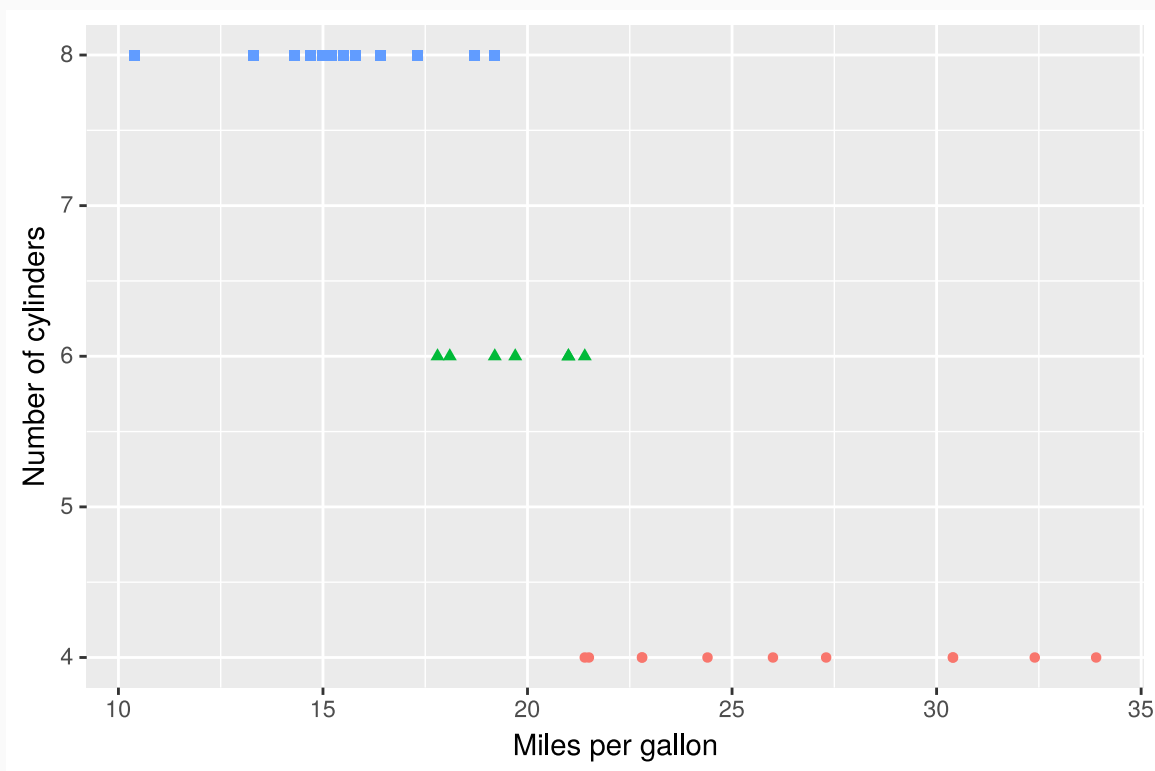
Labeling your graphics

In `ggplot2` you can add/change the title, subtitle, caption, and x- and y-axis labels by adding a `labs()` layer. Below is an example illustrating it's use:

```
ggplot(data = mpg) +  
  geom_point(mapping = aes(x = displ, y = hwy)) + labs(  
    title = "Put your informative title here",  
    subtitle = "and your subtitle here",  
    x = "New x label",  
    y = "New y label",  
    caption = "Put a caption here"  
  )
```


Plotting variables

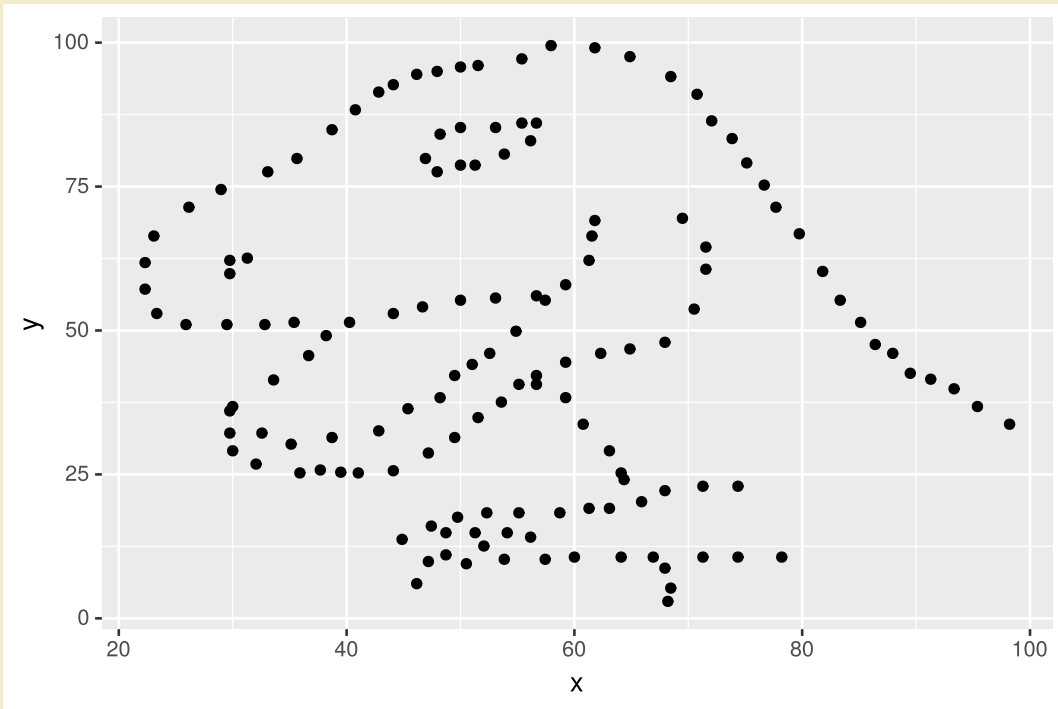
```
ggplot(data = mtcars, aes(x=mpg, y=cyl)) +  
  geom_point(aes(col=as.factor(cyl), pch=as.factor(cyl))) +  
  labs(x='Miles per gallon', y='Number of cylinders') +  
  guides(col=FALSE, pch=FALSE)
```



Your Turn 1

Please git clone the GitHub repository [03-visualizations](#).

The data frame in this exercise is called `datasaurus_dozen` and it's in the `datasauRus` package. This single data frame contains 13 datasets.



05:00

Statistical transformations: Default stats

Common geom	stat
<code>geom_histogram()</code>	<code>stat_bin()</code>
<code>geom_bar()</code>	<code>stat_count()</code>
<code>geom_smooth()</code>	<code>stat_smooth()</code>
<code>geom_boxplot()</code>	<code>stat_boxplot()</code>
<code>geom_density()</code>	<code>stat_density()</code>

- Every geom has a default stat
- Often no need to explicitly specify the stat
- Check help files: `?geom_bar`

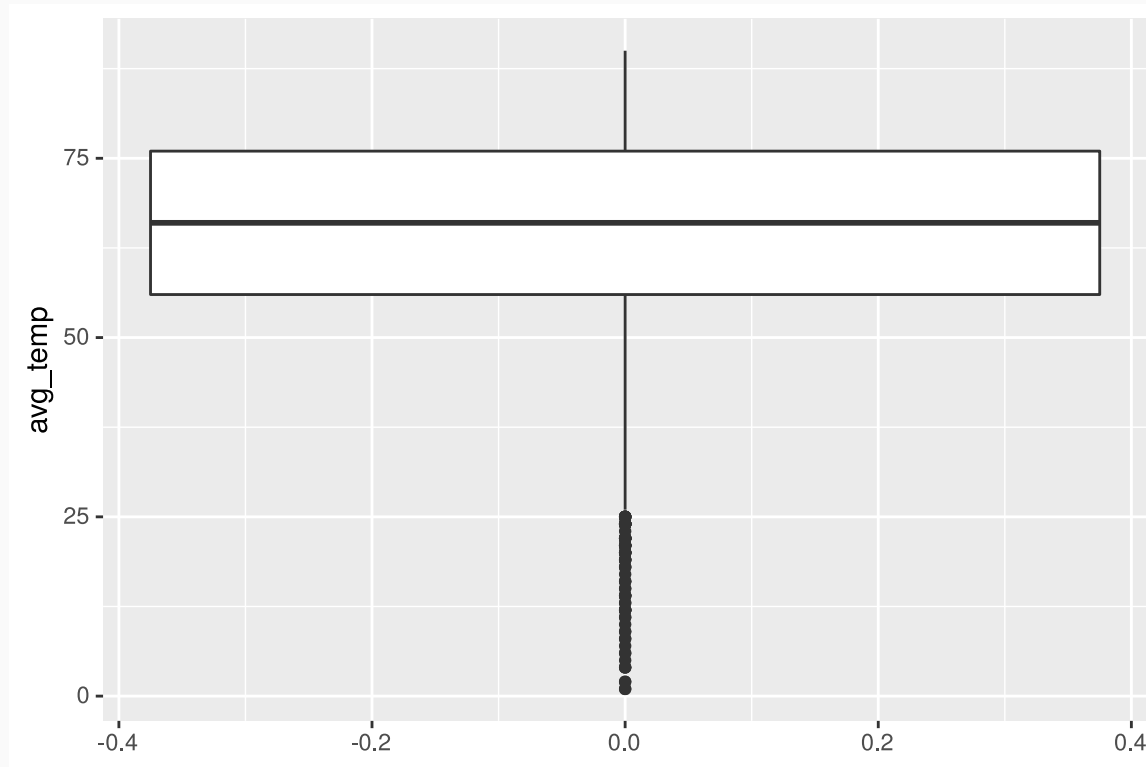
Weather Dataset

The data set `Weather` contains data on weather-related variables for several world cities.

```
#install.packages(mosaicData)
library(mosaicData)
data(Weather)
glimpse(Weather)
Rows: 3,655
Columns: 25
$ city      <chr> "Auckland", "Auckland", "Auckland", "Auckland", "Aucklan...
$ date      <date> 2016-01-01, 2016-01-02, 2016-01-03, 2016-01-04, 2016-01...
$ year      <dbl> 2016, 2016, 2016, 2016, 2016, 2016, 2016, 2016, 2016, 20...
$ month     <int> 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, ...
$ day       <dbl> 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 1...
$ high_temp <dbl> 68, 68, 77, 73, 69, 69, 71, 77, 69, 71, 75, 69, 71, 75, ...
$ avg_temp  <dbl> 65, 66, 72, 66, 62, 63, 66, 70, 66, 66, 67, 66, 66, 68, ...
$ low_temp  <dbl> 62, 64, 66, 60, 55, 57, 60, 64, 64, 62, 59, 62, 62, 62, ...
$ high_dewpt <dbl> 64, 64, 70, 66, 55, 54, 59, 72, 68, 63, 61, 66, 61, 63, ...
$ avg_dewpt <dbl> 60, 63, 67, 60, 52, 51, 54, 67, 61, 58, 58, 62, 57, 61, ...
$ low_dewpt <dbl> 55, 61, 64, 54, 48, 46, 50, 59, 55, 55, 54, 59, 54, 59, ...
$ high_humidity <dbl> 100, 100, 100, 100, 82, 88, 83, 100, 100, 88, 94, 100, 8...
$ avg_humidity <dbl> 82, 94, 91, 76, 69, 65, 65, 92, 81, 76, 72, 87, 73, 80, ...
$ low_humidity <dbl> 68, 88, 74, 53, 56, 46, 53, 83, 64, 64, 53, 78, 64, 65, ...
$ high_hg    <dbl> 30.15, 30.04, 29.80, 30.12, 30.21, 30.24, 30.24, 30.01, ...
```

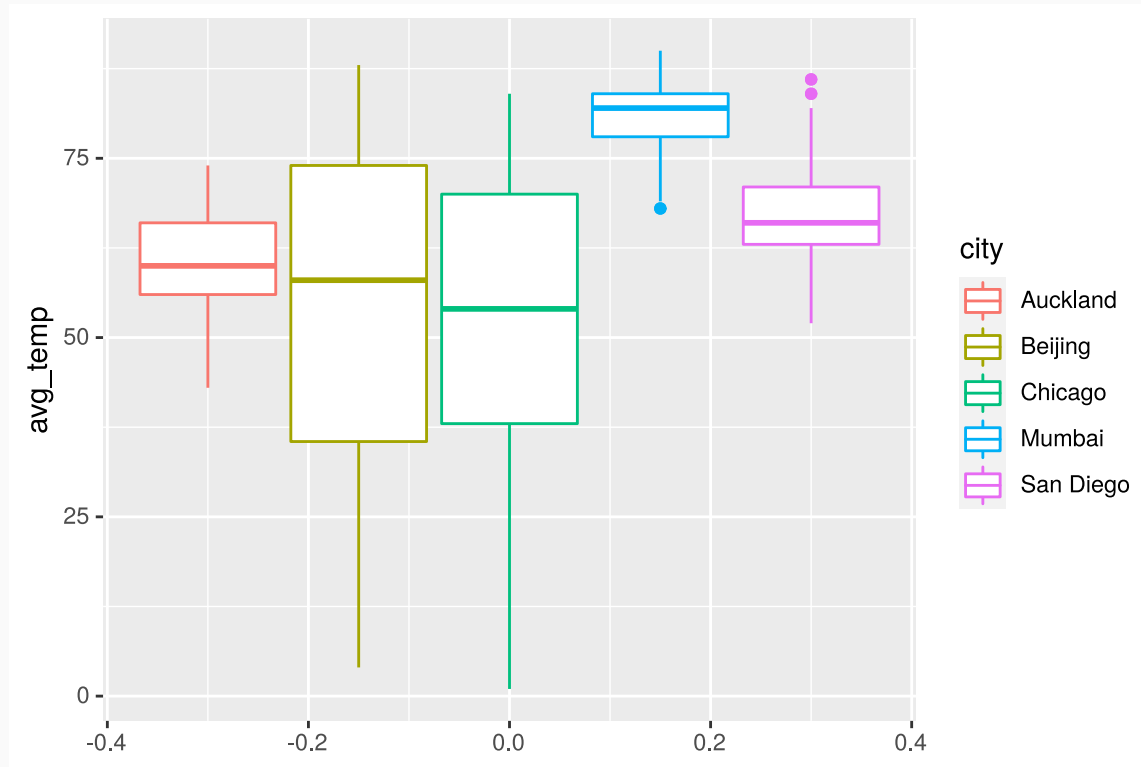
Weather patterns

```
ggplot(Weather, aes(y=avg_temp)) + geom_boxplot()
```



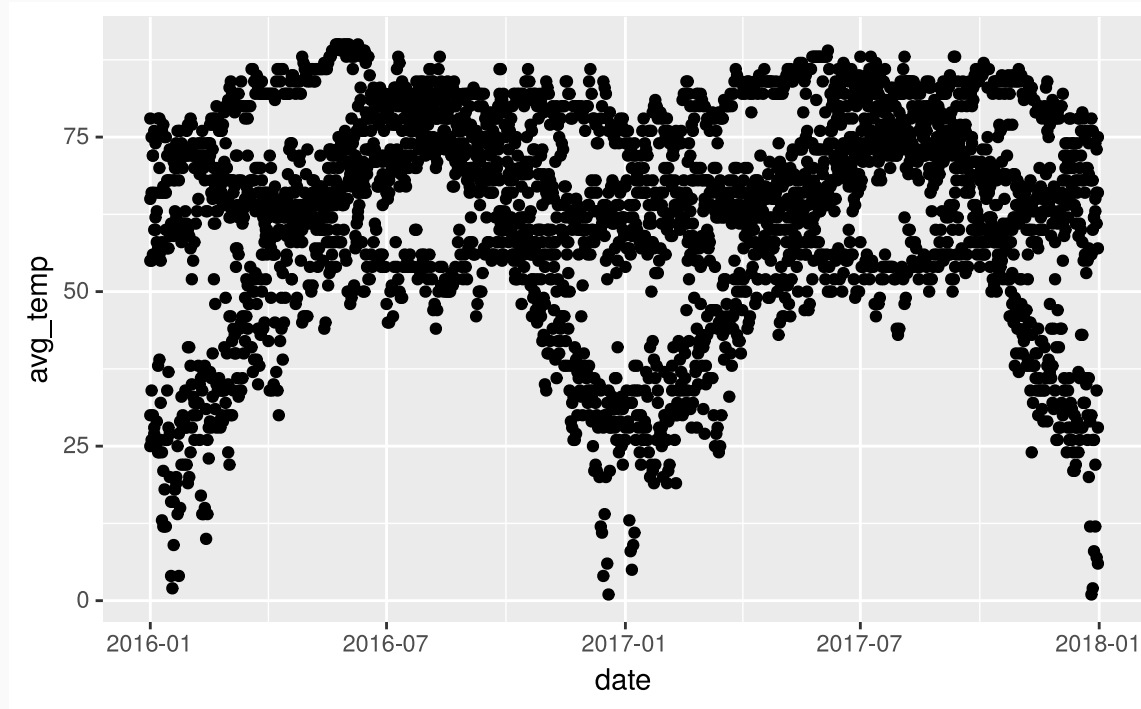
Weather patterns

```
ggplot(Weather, aes(y=avg_temp, group=city)) + geom_boxplot(aes(color=city))
```



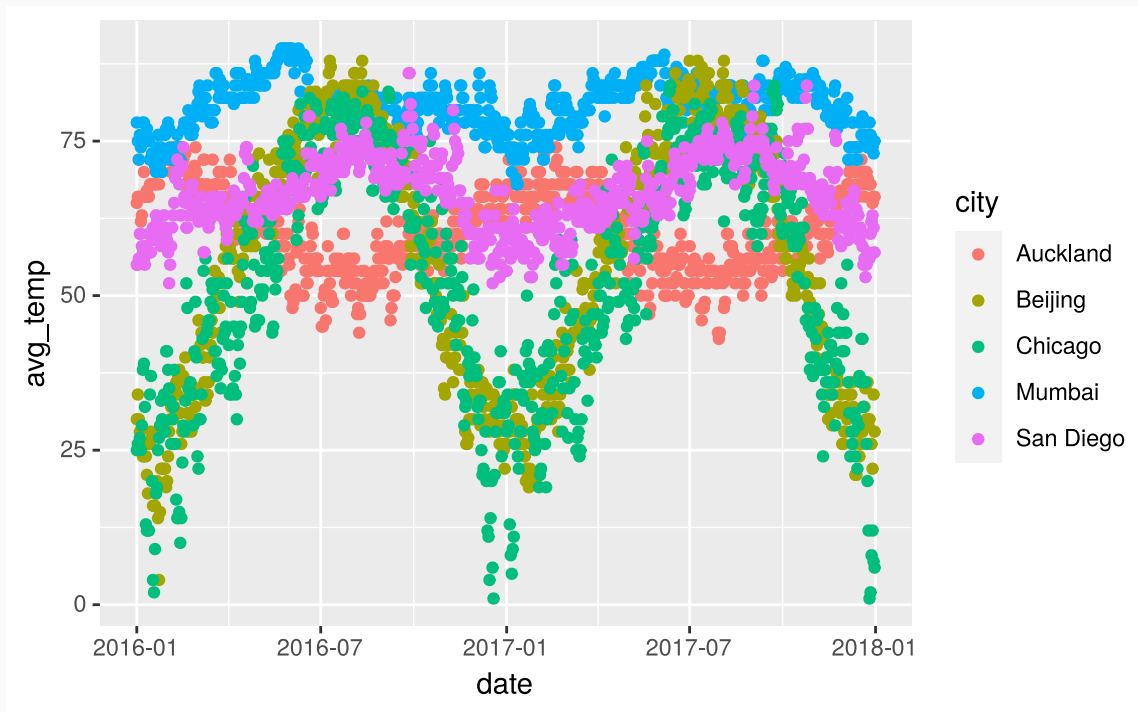
Weather patterns

```
ggplot(Weather, aes(x=date, y=avg_temp))+geom_point()
```



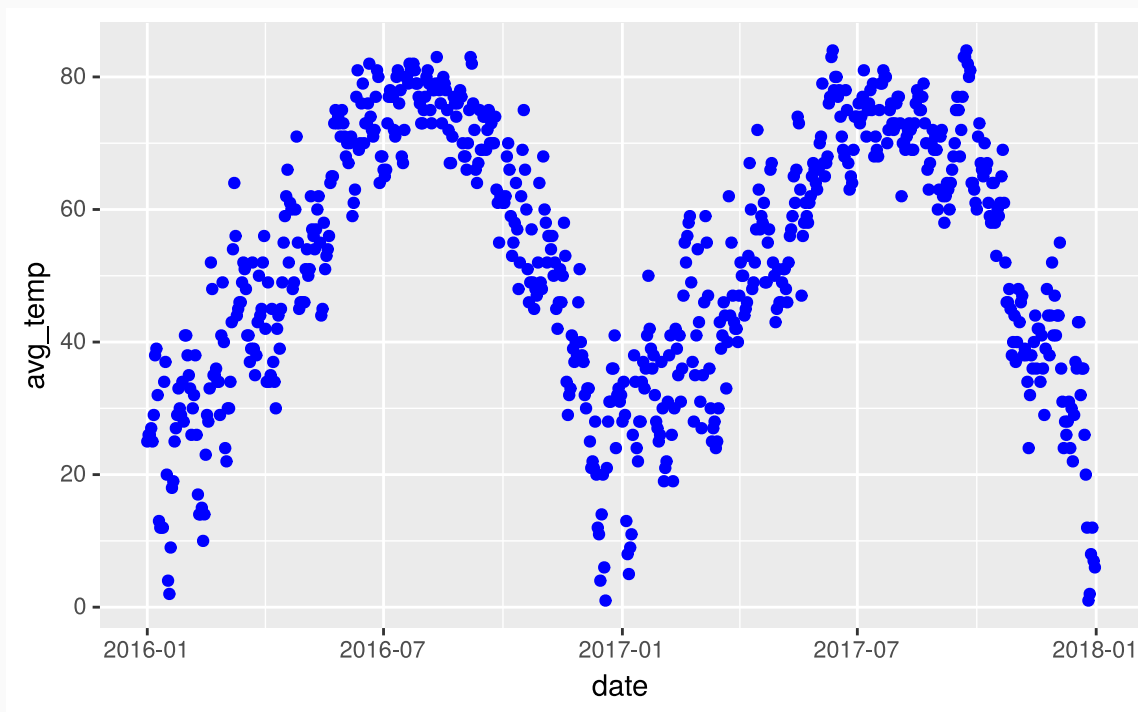
Weather patterns

```
ggplot(Weather, aes(x=date, y=avg_temp)) +  
  geom_point(aes(color=city))
```



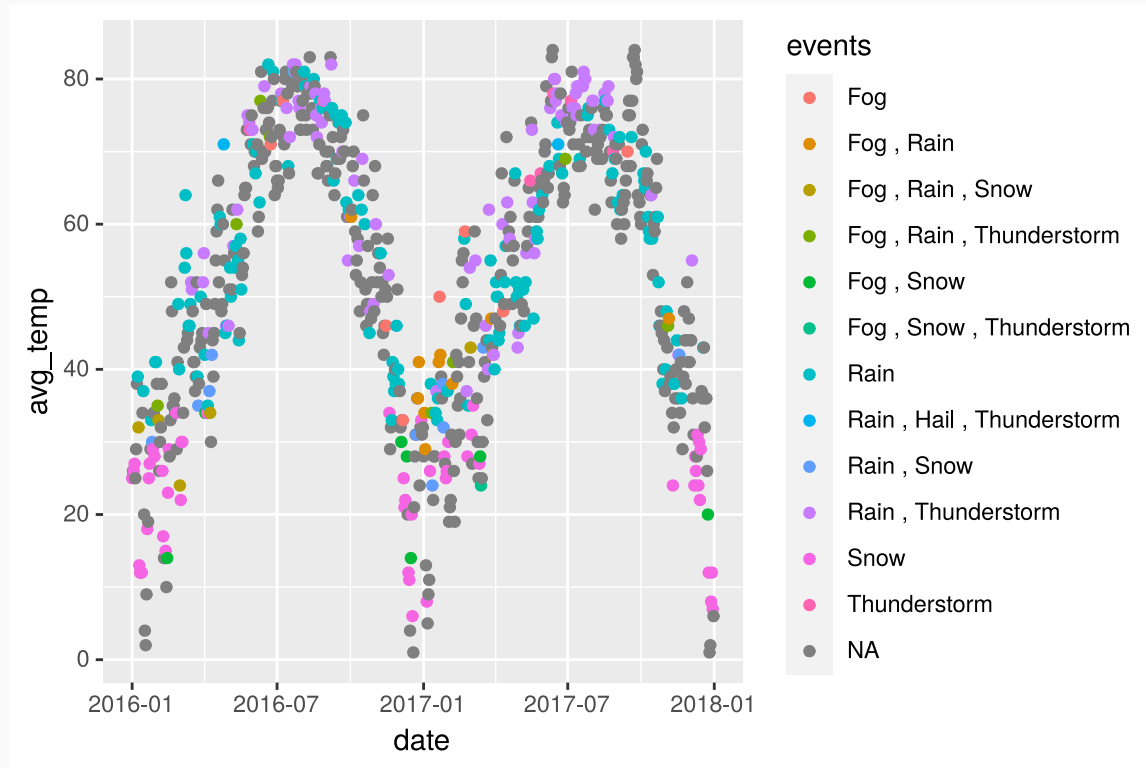
Weather patterns in Chicago

```
Chicago <- Weather %>% filter(city=='Chicago')  
ggplot(Chicago, aes(x=date, y=avg_temp)) +  
  geom_point(color='blue')
```



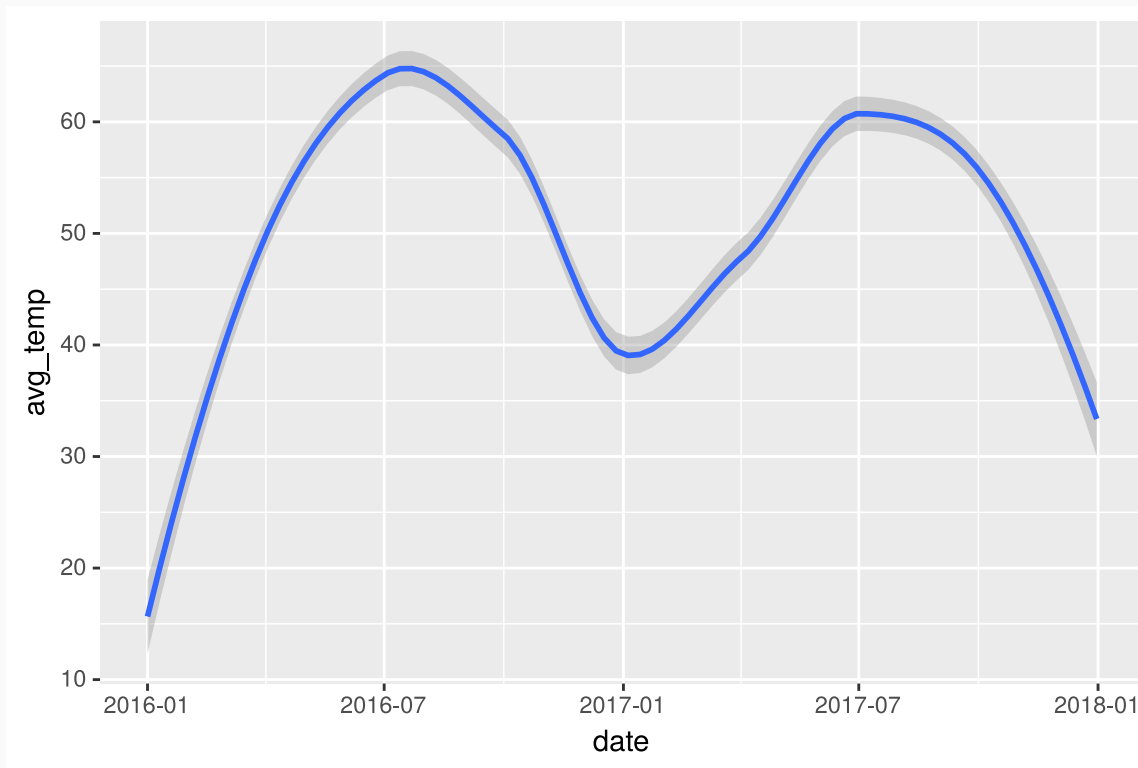
Weather patterns in Chicago

```
ggplot(Chicago, aes(x=date, y=avg_temp)) +  
  geom_point(aes(color=events))
```



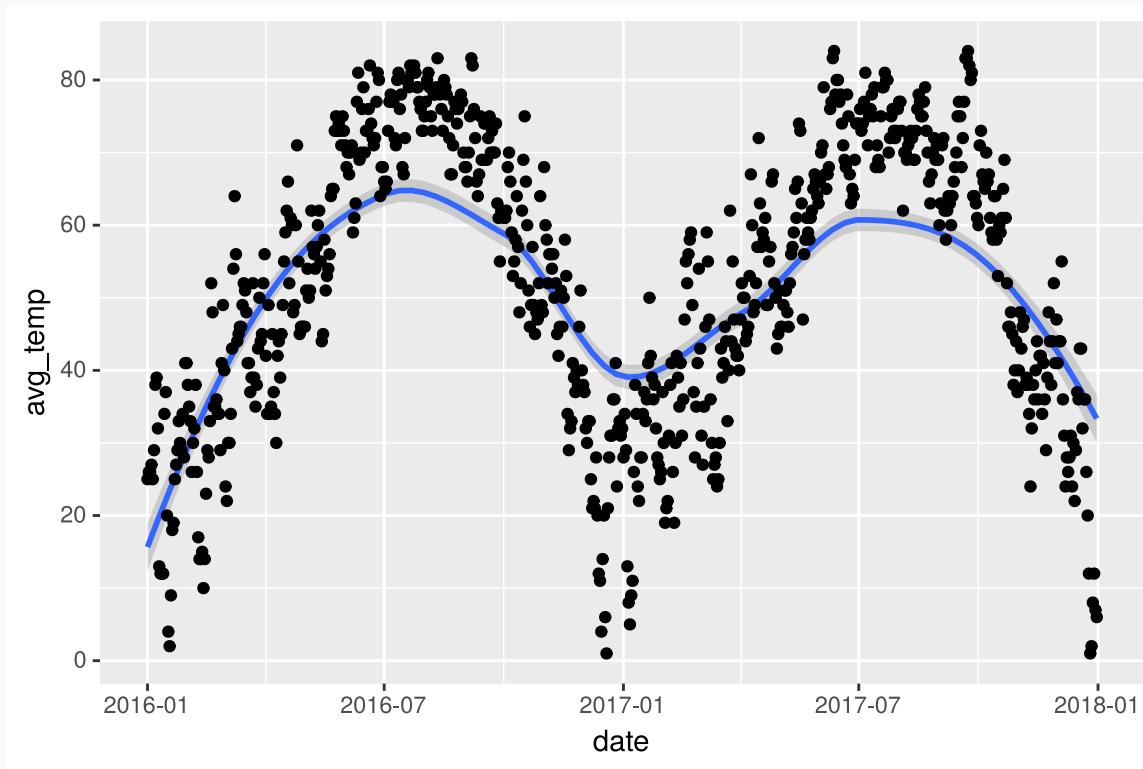
Weather patterns in Chicago

```
ggplot(Chicago, aes(x=date, y=avg_temp)) +  
  geom_smooth()
```



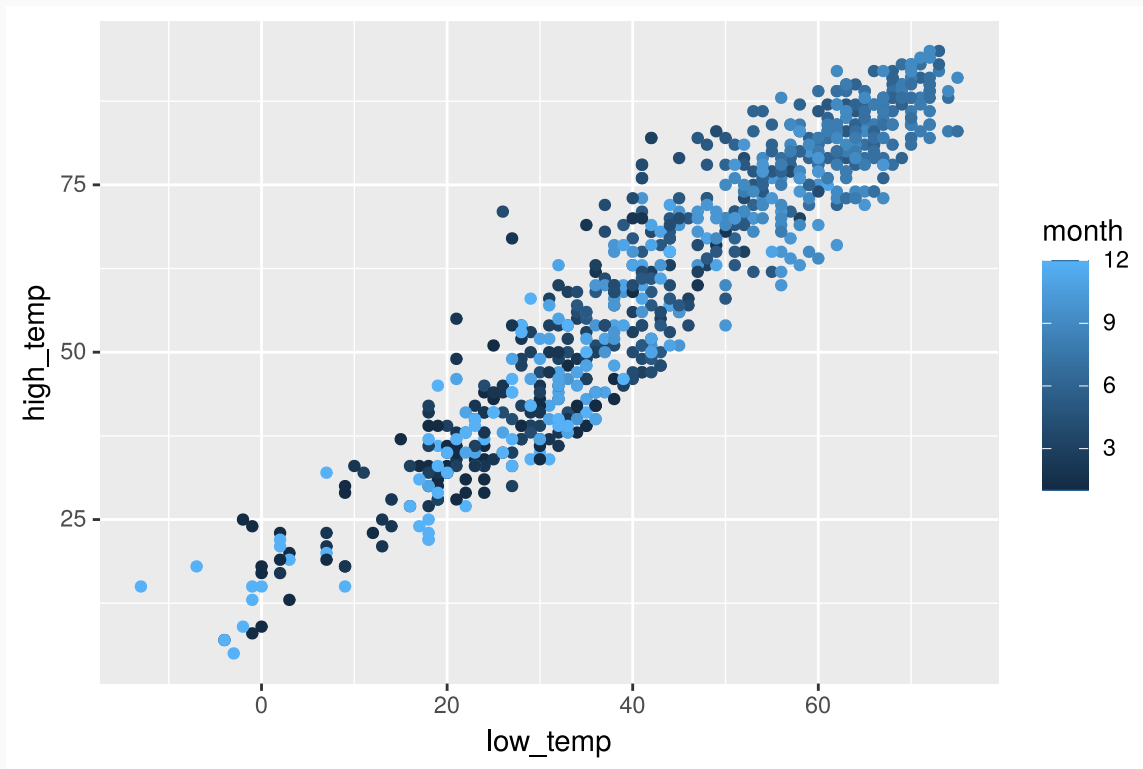
Weather patterns in Chicago

```
ggplot(Chicago, aes(x=date, y=avg_temp)) +  
  geom_smooth()+ geom_point()
```



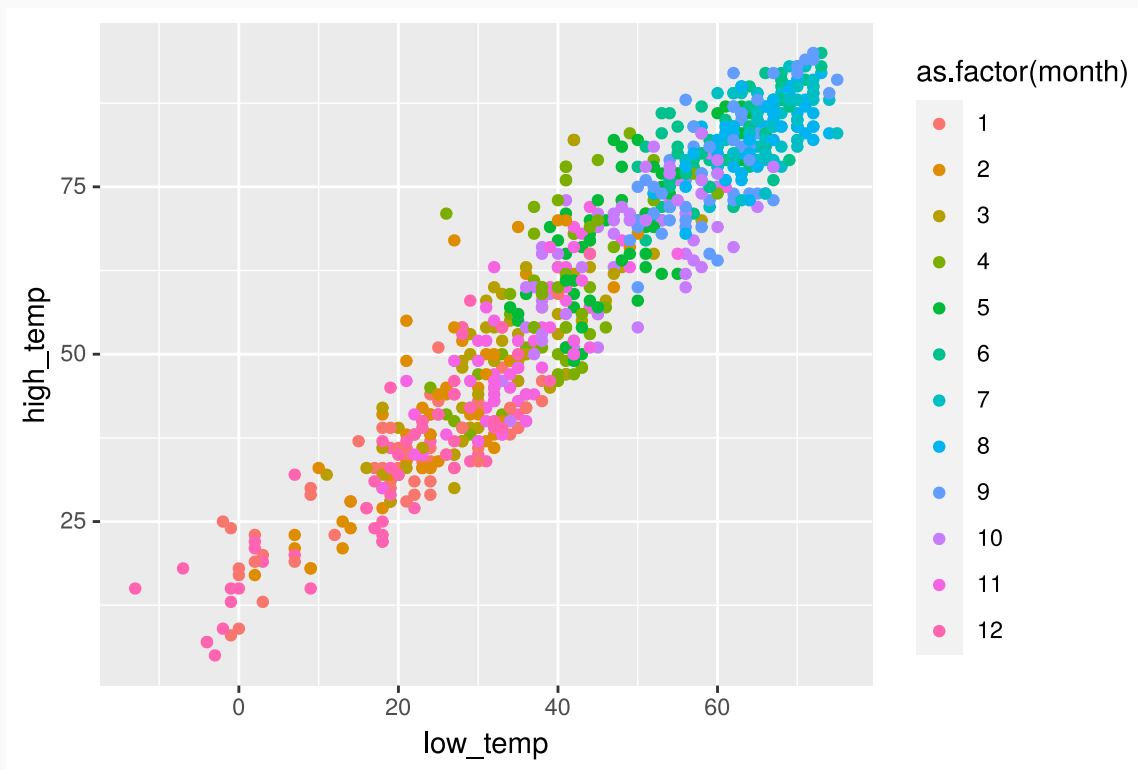
Weather patterns in Chicago

```
ggplot(Chicago, aes(x=low_temp, y=high_temp)) +  
  geom_point(aes(color=month))
```



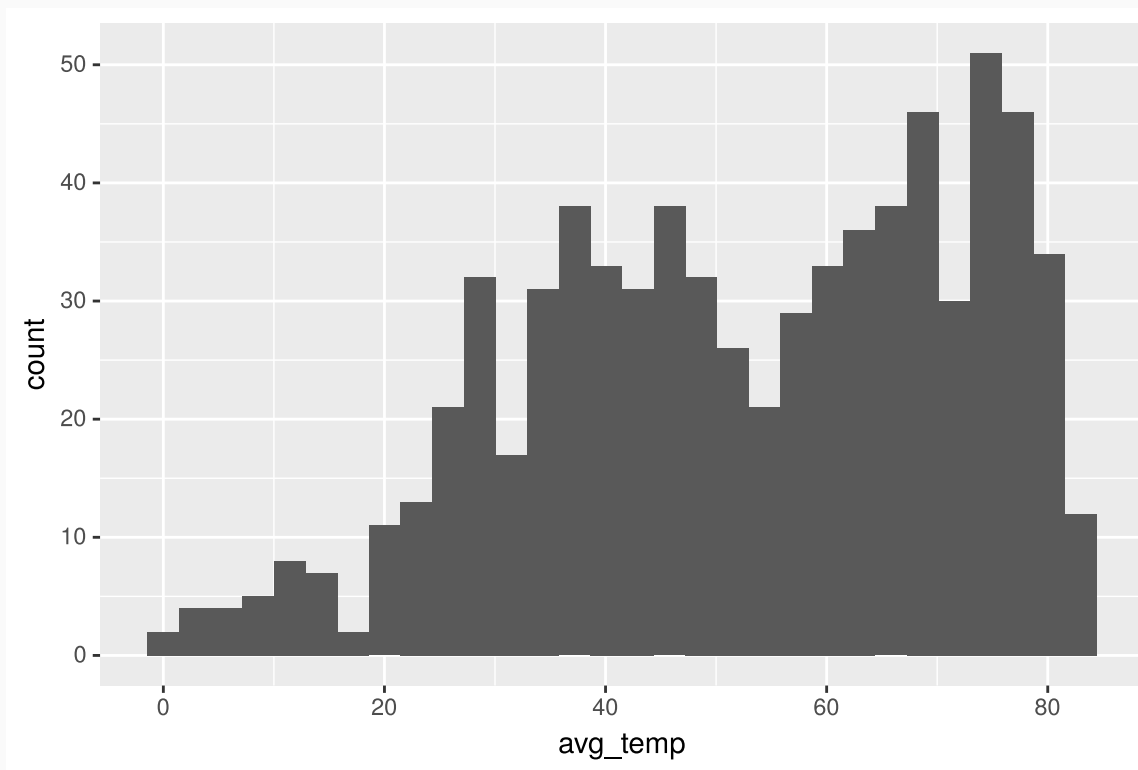
Weather patterns in Chicago

```
ggplot(Chicago, aes(x=low_temp, y=high_temp)) +  
  geom_point(aes(color=as.factor(month)))
```



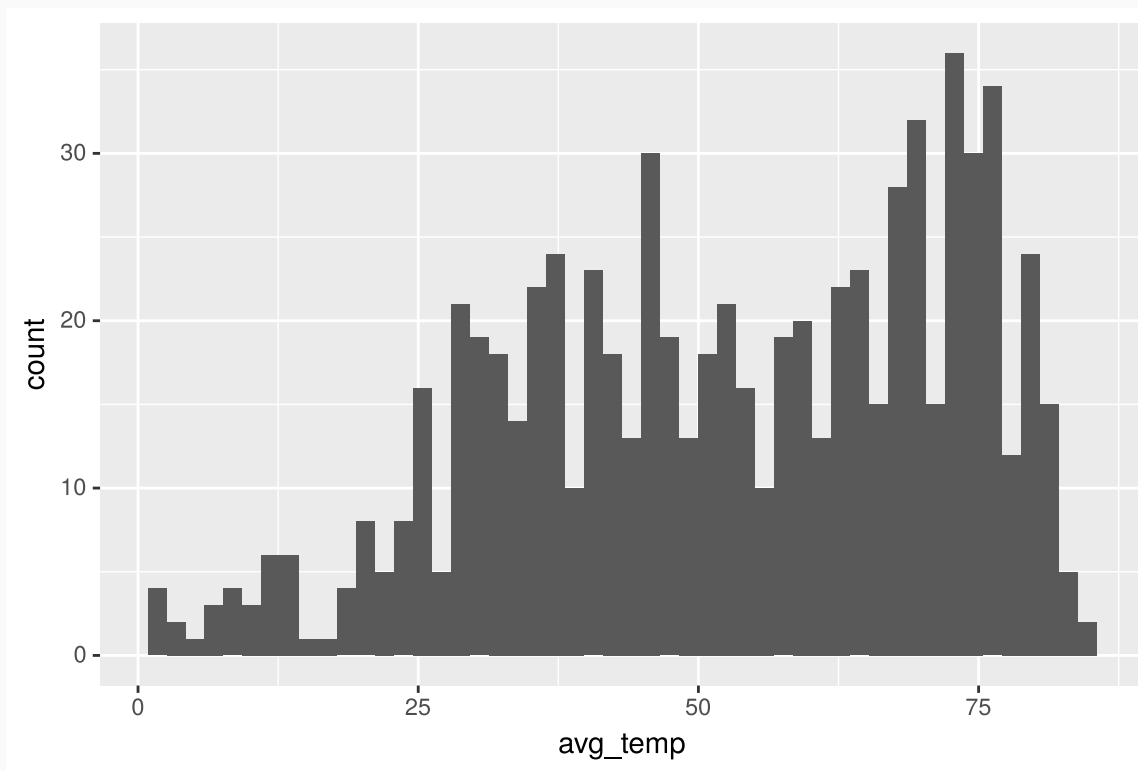
Weather patterns in Chicago

```
ggplot(Chicago, aes(x=avg_temp)) +  
  geom_histogram()
```



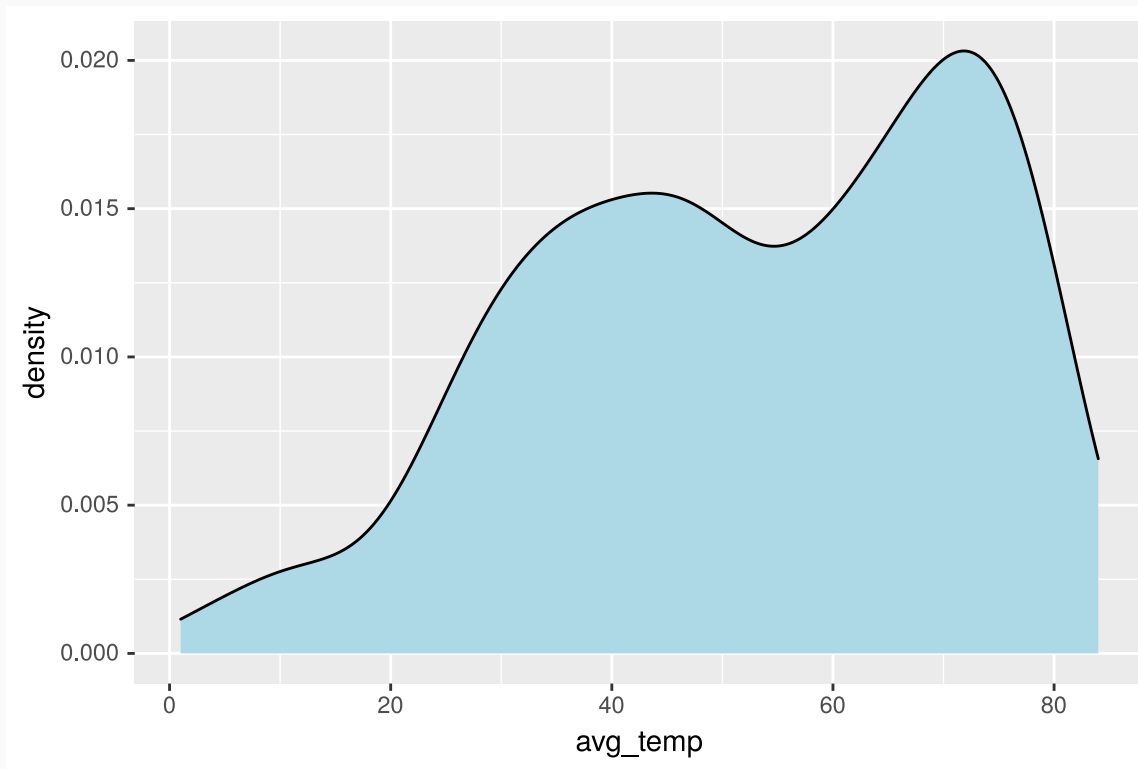
Weather patterns in Chicago

```
ggplot(Chicago, aes(x=avg_temp)) +  
  geom_histogram(bins = 50)
```



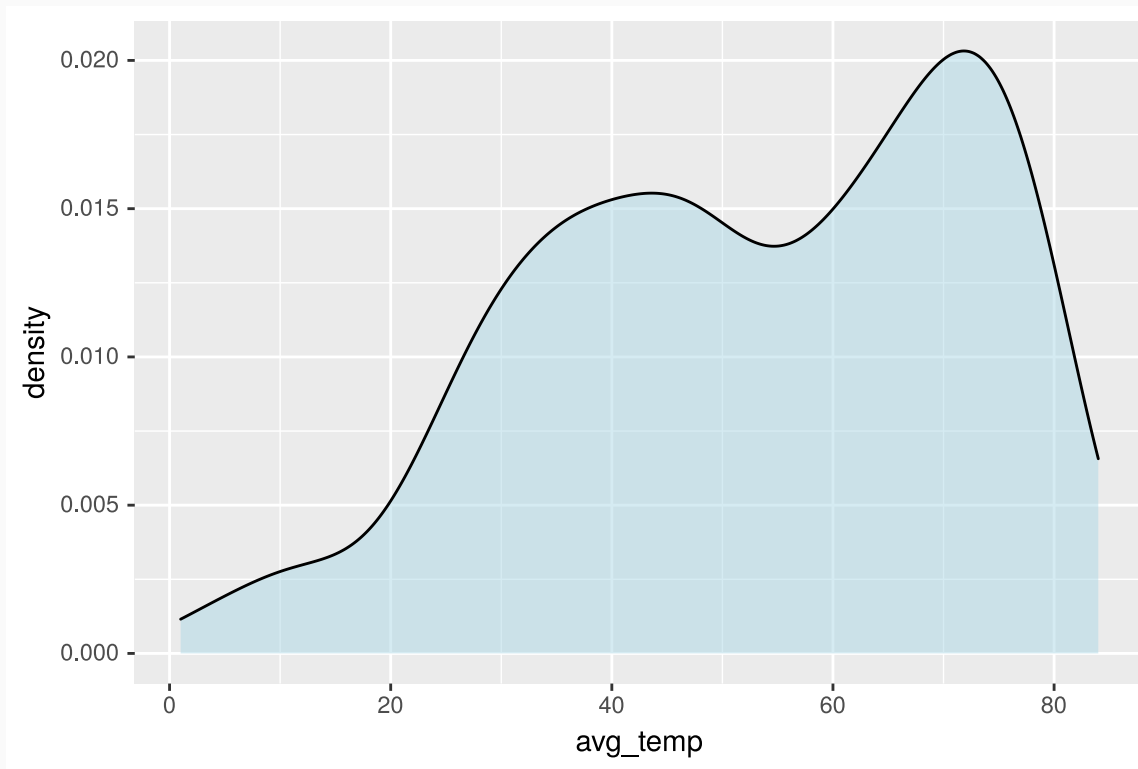
Weather patterns in Chicago

```
ggplot(Chicago, aes(x=avg_temp)) +  
  geom_density(fill='lightblue')
```



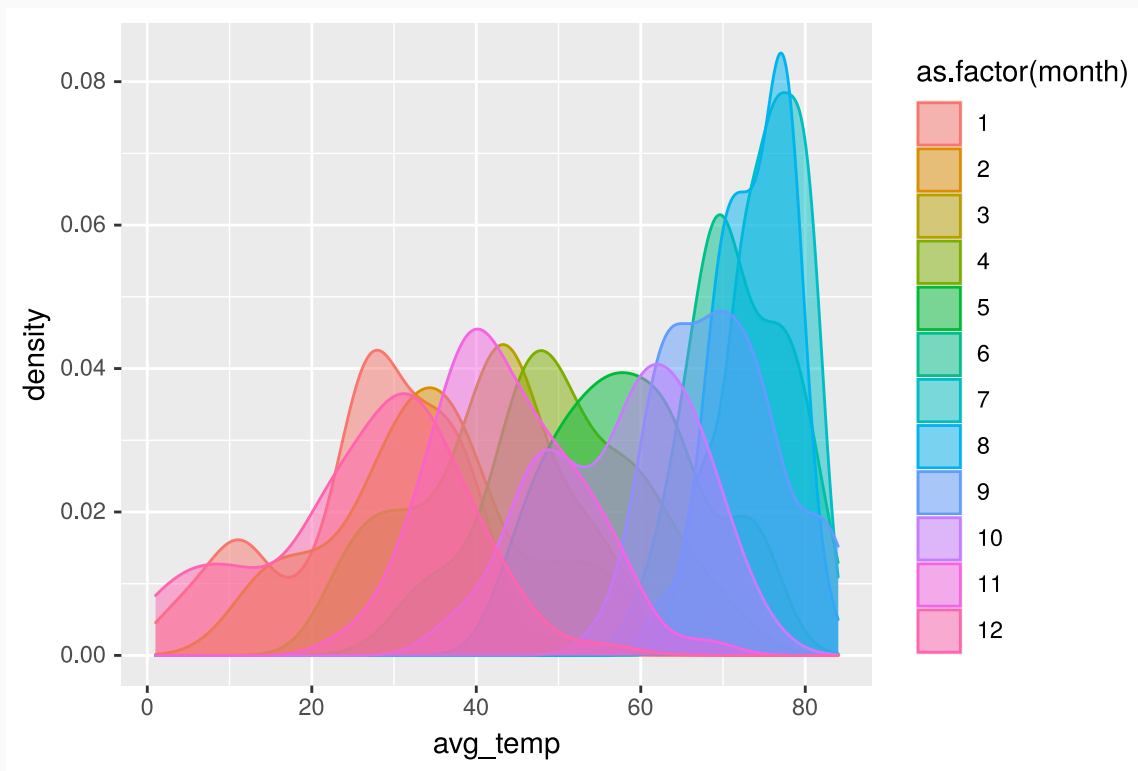
Weather patterns in Chicago

```
ggplot(Chicago, aes(x=avg_temp)) +  
  geom_density(fill='lightblue', alpha=0.5)
```



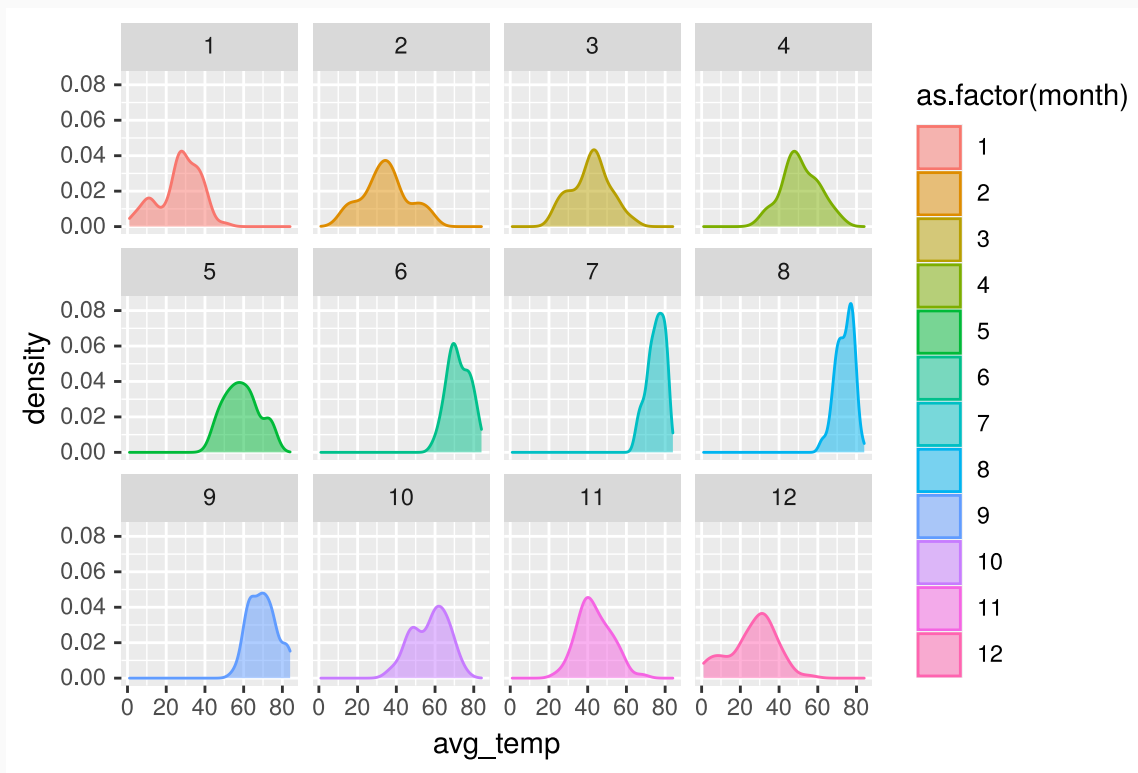
Weather patterns in Chicago

```
ggplot(Chicago, aes(x=avg_temp, group=month)) +  
  geom_density(aes(color=as.factor(month), fill=as.factor(month)), alpha=0.5)
```



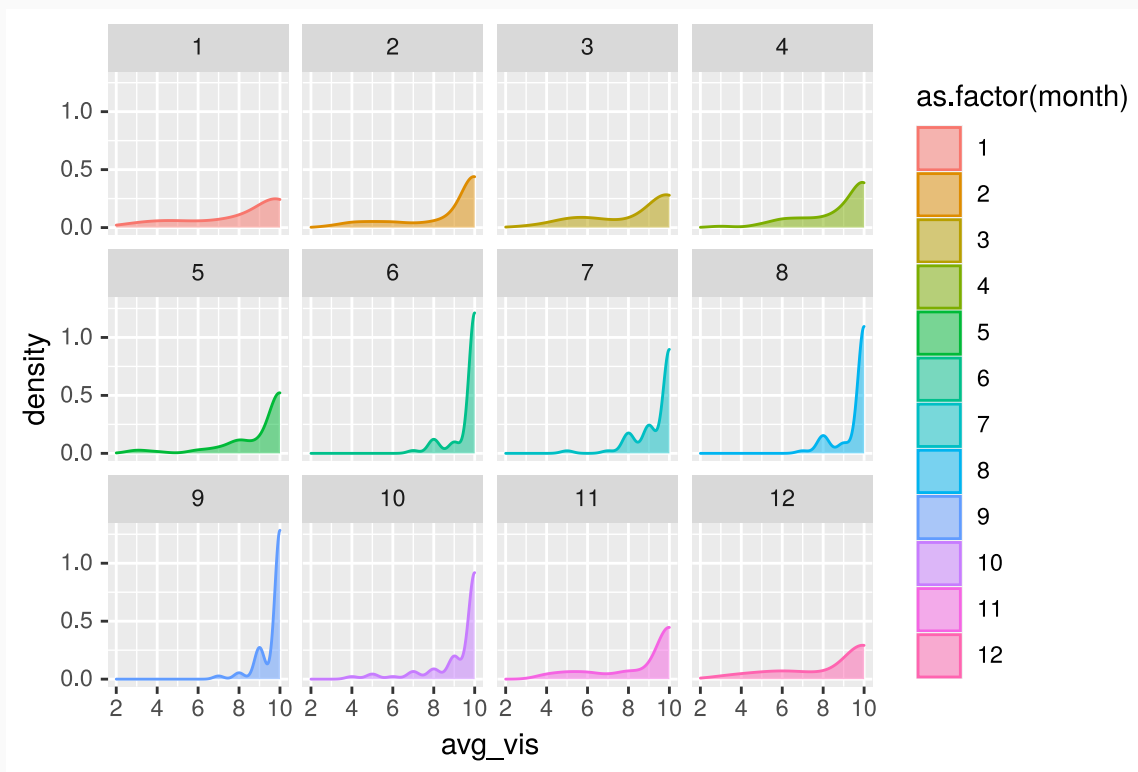
Weather patterns in Chicago

```
ggplot(Chicago, aes(x=avg_temp, group=month)) +  
  geom_density(aes(color=as.factor(month), fill=as.factor(month)), alpha=0.5) +  
  facet_wrap(~month, nrow=3)
```



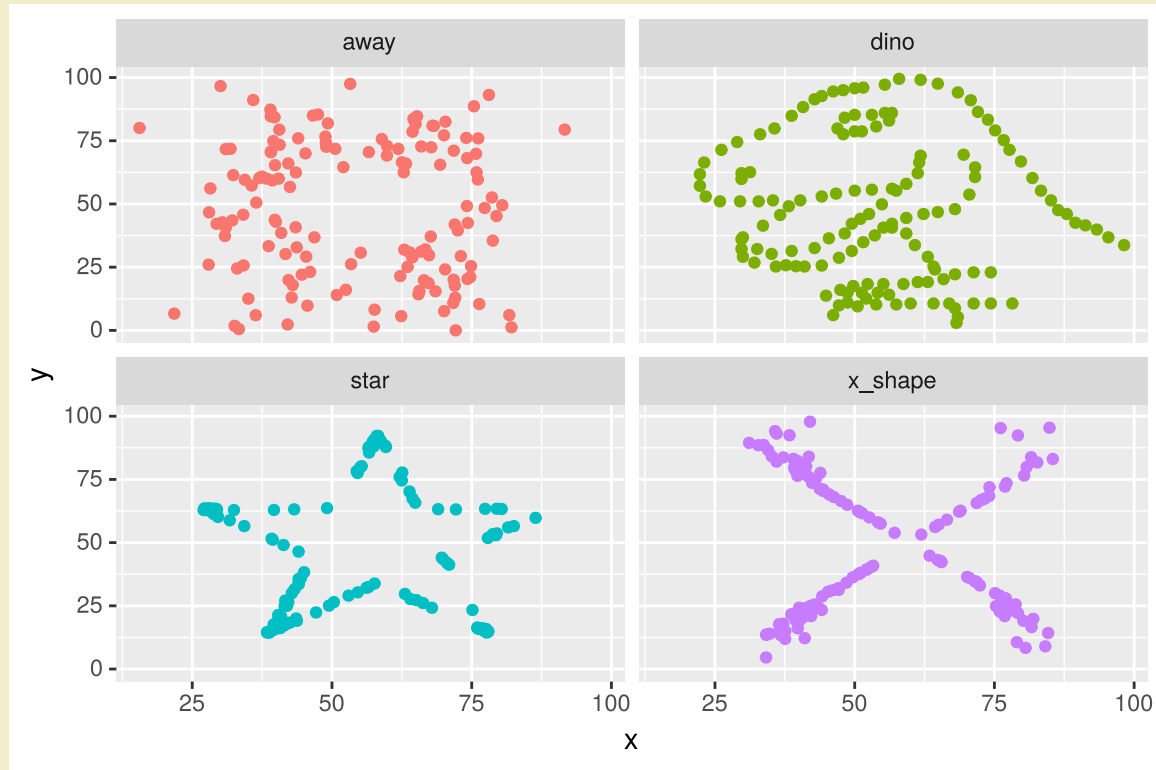
Weather patterns in Chicago

```
ggplot(Chicago, aes(x=avg_vis, group=month)) +  
  geom_density(aes(color=as.factor(month), fill=as.factor(month)), alpha=0.5) +  
  facet_wrap(~month, nrow=3)
```



Your Turn 2

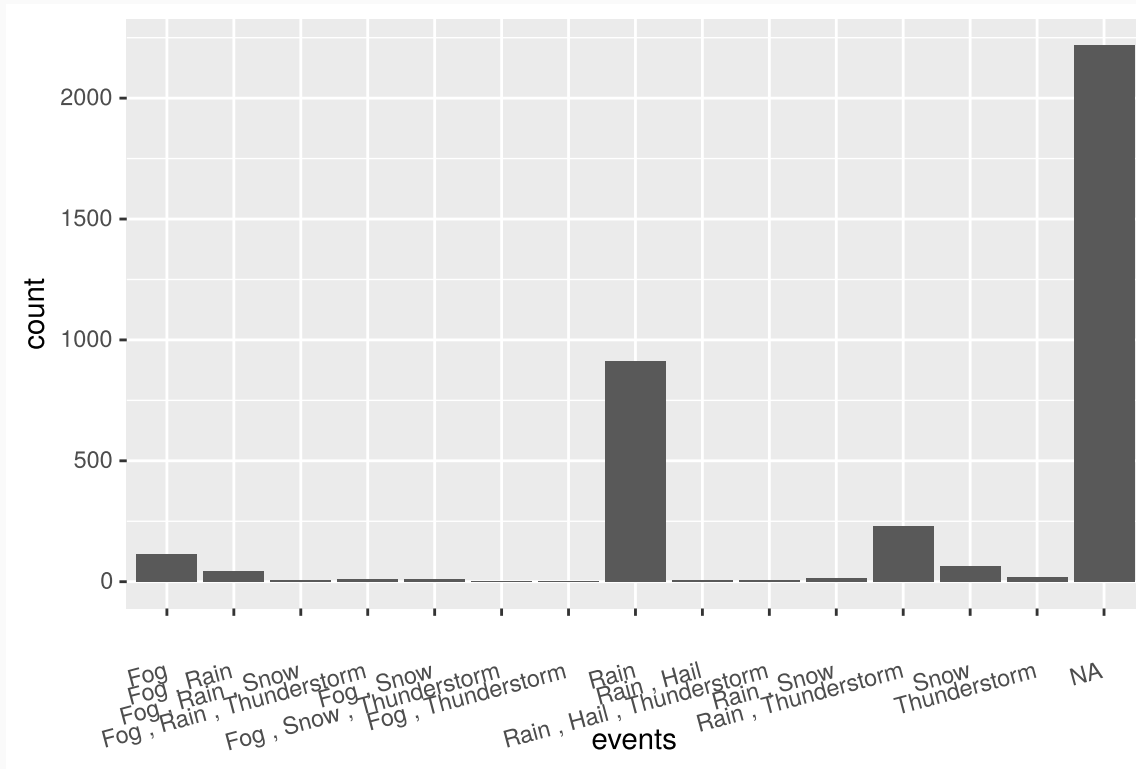
- `datasaurus_dozen` in the `datasauRus` package contains 13 datasets. We will use 4 of these datasets for this exercise.
- Create this ribbon of plots using `facet_wrap`.



05:00

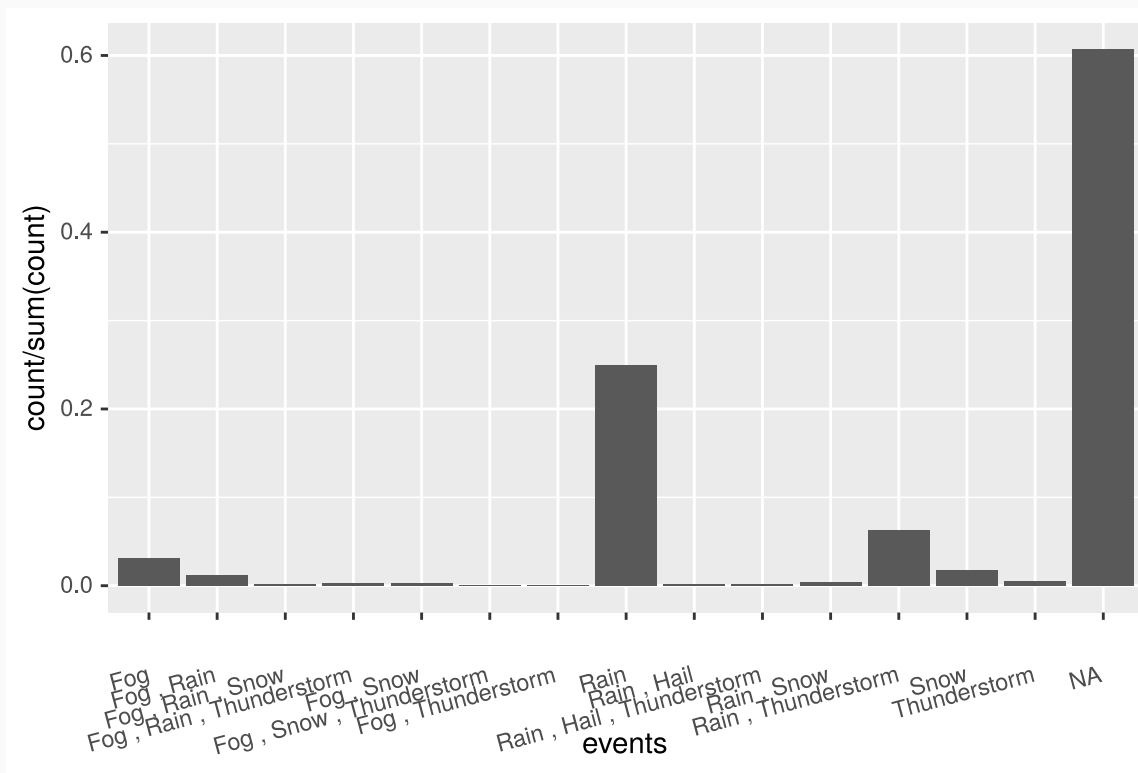
Weather events

```
ggplot(Weather, aes(x=events)) +  
  geom_bar()+  
  theme(axis.text.x = element_text(angle = 15, vjust = 0.5, hjust=1))
```



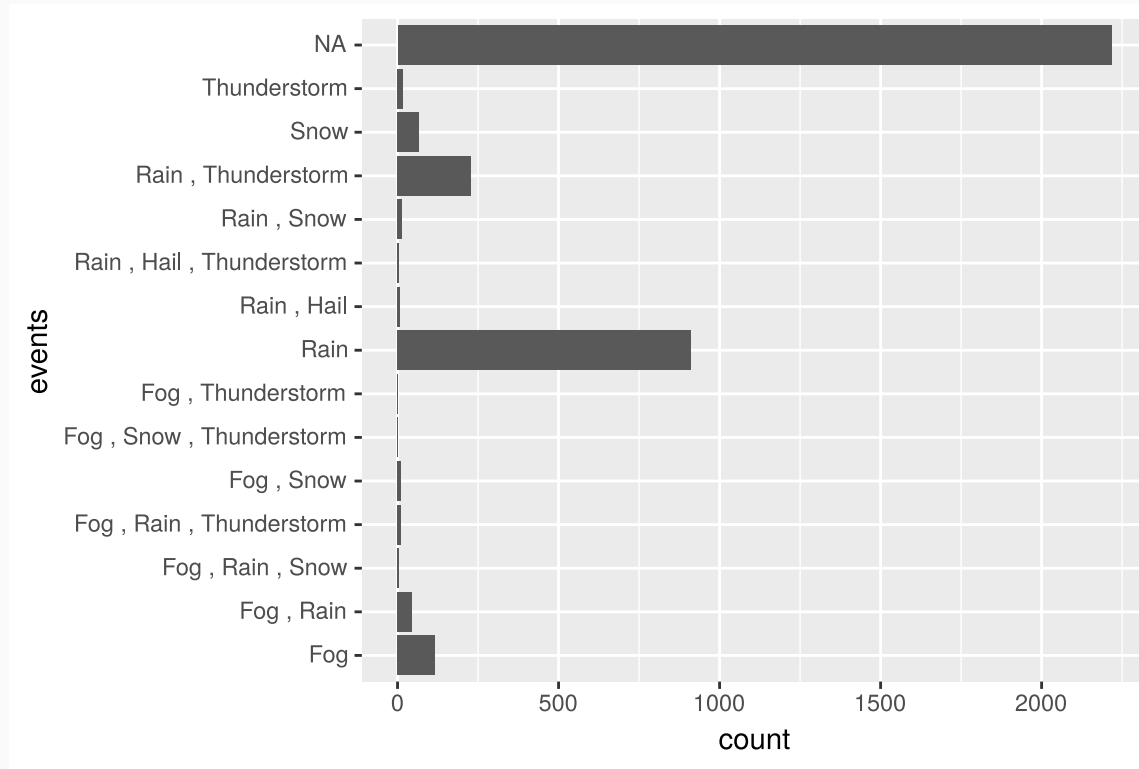
Weather events — proportions

```
ggplot(Weather, aes(x=events, y = ..count../sum(..count..))) + # change y-axis to proportion
  geom_bar()+
  theme(axis.text.x = element_text(angle = 15, vjust = 0.5, hjust=1))
```



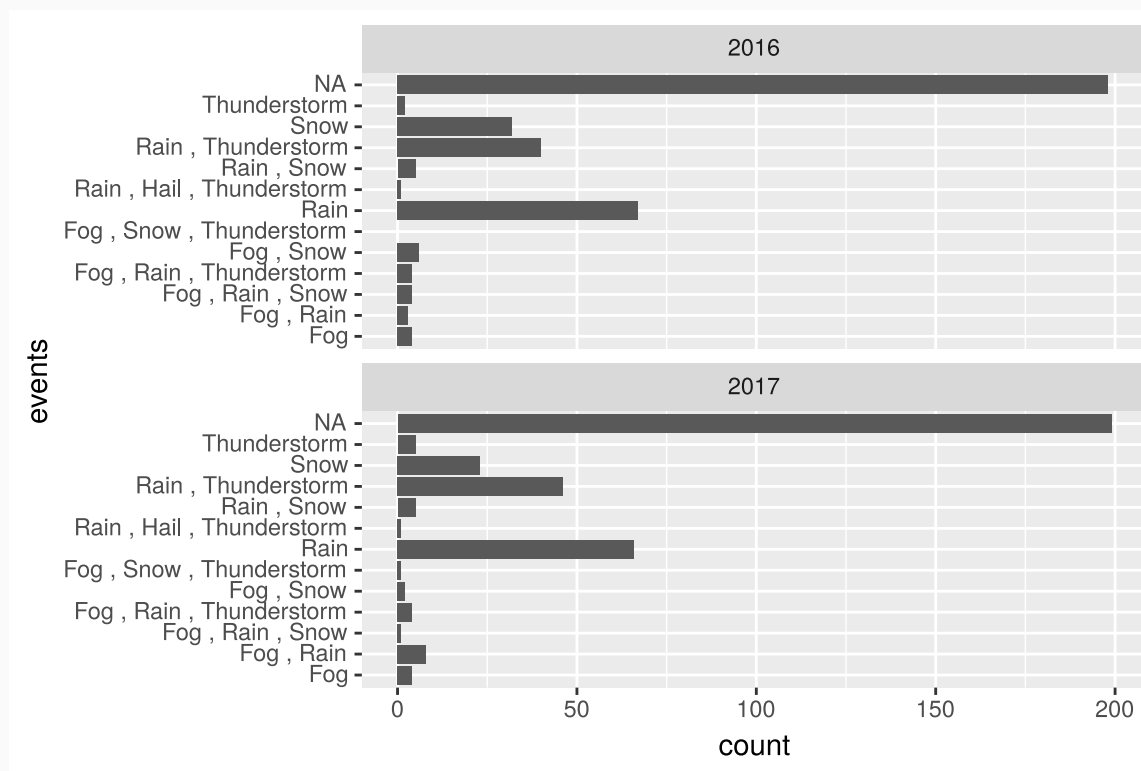
Weather events

```
ggplot(Weather, aes(x=events)) +  
  geom_bar() +  
  coord_flip()
```



Weather events in Chicago

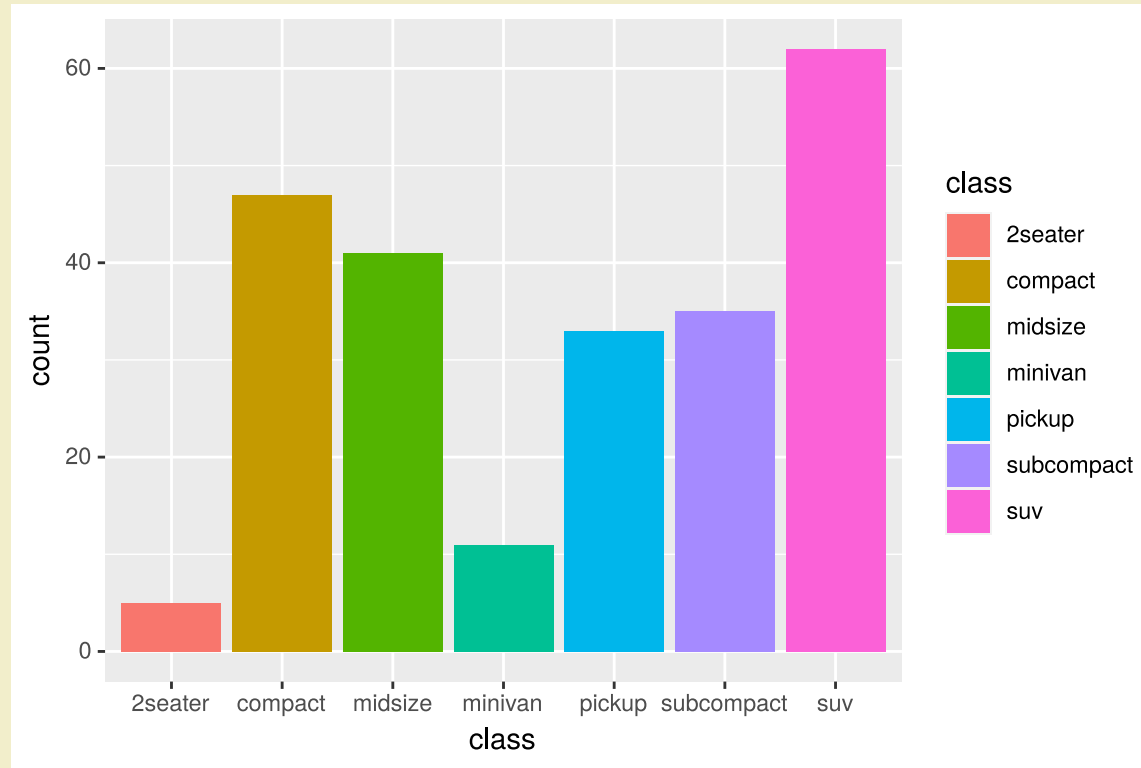
```
ggplot(Chicago, aes(x=events)) +  
  geom_bar() +  
  coord_flip() +  
  facet_wrap(~year, nrow=2)
```



Your Turn 3

03:00

- The `mpg` data set is loaded with the `tidyverse`. Run `?mpg` for info.
- Create this bar chart of vehicle `class`



Gapminder dataset

```
gap_dat <- read.csv("https://raw.githubusercontent.com/deepbas/statdatasets/main/gapminder2018.csv")
gapminder <- gap_dat %>% filter(year == 2018)
glimpse(gapminder)
Rows: 193
Columns: 8
$ country      <chr> "Afghanistan", "Albania", "Algeria", "Andorra", "Angol...
$ year         <int> 2018, 2018, 2018, 2018, 2018, 2018, 2018, 2018, 2018, ...
$ income       <int> 1870, 12400, 13700, 51500, 5850, 21000, 18900, 8660, 4...
$ life_expectancy <dbl> 58.7, 78.0, 77.9, NA, 65.2, 77.6, 77.0, 76.0, 82.9, 81...
$ population   <int> 36400000, 2930000, 42000000, 77000, 30800000, 103000, ...
$ four_regions <chr> "asia", "europe", "africa", "europe", "africa", "ameri...
$ eight_regions <chr> "asia_west", "europe_east", "africa_north", "europe_we...
$ six_regions  <chr> "south_asia", "europe_central_asia", "middle_east_nort...
```

Your Turn 4

The `gapminder` dataset provides values for life expectancy, GDP per capita, and population, every five years, from 1960 to 2018. Use `gapminder` dataset to answer the given set of problems.



10:00