

Lecture 5: 3D and Interactive graphics

October 17, 2016

Packages:

- **plotly** - package translates ‘ggplot2’ graphs to an interactive web-based version and/or create custom interactive web-based visualizations directly from R.
- **heatmaply** - package for generating interactive heatmaps
- **ggmap** - package that let you retrieve maps from popular online mapping services like Google Maps, OpenStreetMap, Stamen Maps, and plot them using the ggplot2 framework

Install the packages

```
# Install packages:  
.packages <- c("ggplot2", "plotly", "heatmaply", "ggmap", "dplyr")  
lapply(.packages, install.packages, dependencies = TRUE)
```

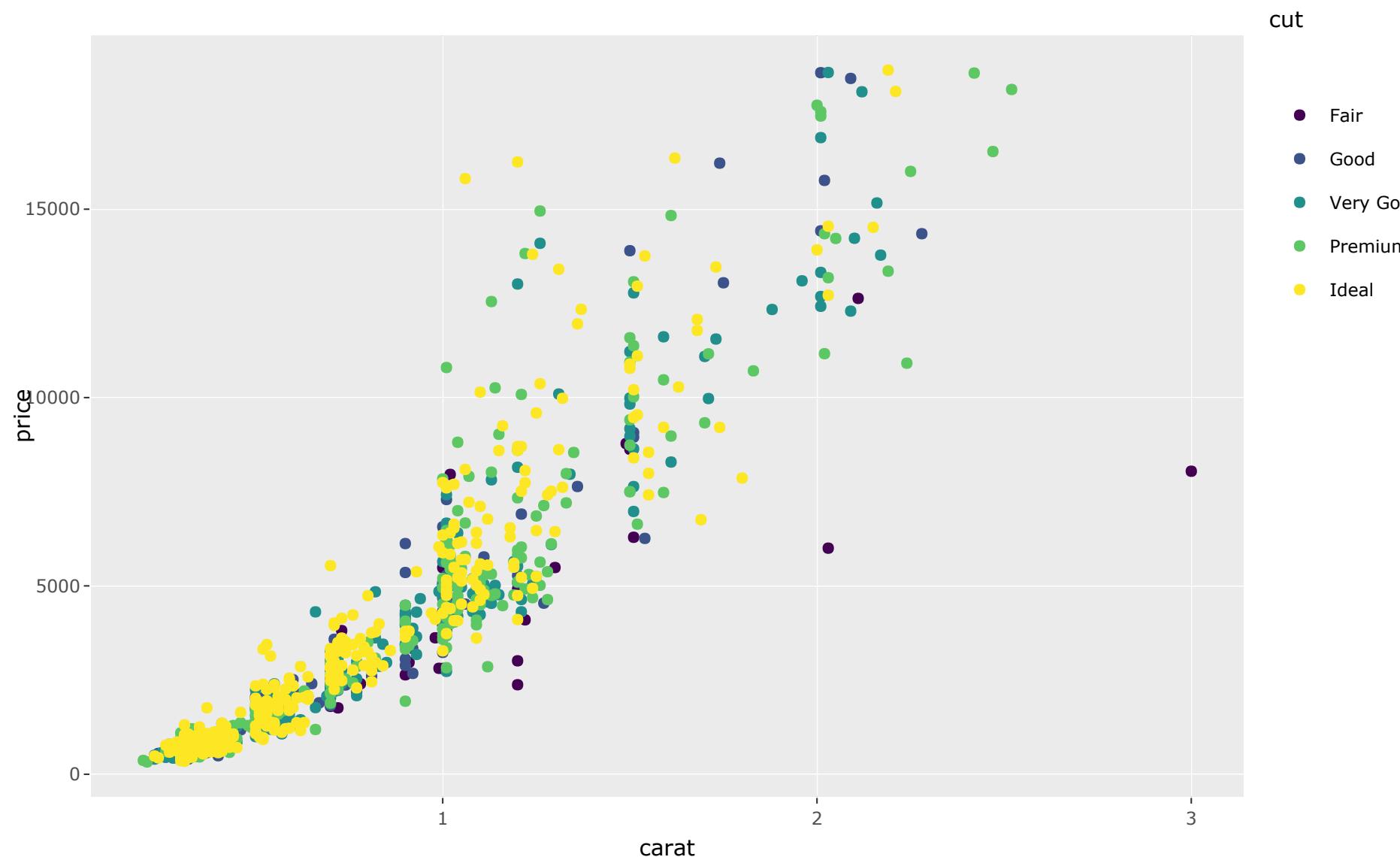
plotly

plotly package

- **plotly** is a **package for visualization** and a collaboration platform for data science
- Available in **R, python, MATLAB, scala.**
- You can produce **interactive graphics including 3D plots** (with zooming and rotating).
- You can open a '**plotly**' account to upload '**plotly**' graphs and view or modify them in a web browser.
- Resources: [cheatsheet](#), [book](#)

plotly integration with ggplot2

```
library(plotly); library(ggplot2); library(dplyr)
plt <- ggplot(diamonds %>% sample_n(1000), aes(x = carat, y = price)) +
  geom_point(aes(color = cut))
ggplotly(plt)
```



```
plt <- ggplot(diamonds %>% sample_n(1000), aes(x = carat, y = price)) +  
  geom_text(aes(label = clarity), size = 4) +  
  geom_smooth(aes(color = cut, fill = cut)) +  
  facet_wrap(~cut)  
ggplotly(plt)
```

```
## `geom_smooth()` using method = 'loess' and formula 'y ~ x'
```

3D Scatter plots

```
z <- seq(0, 10, 0.2)
df <- data.frame(U = seq(0, 10, 0.2), V = cos(z), W = sin(z)*z)
plot_ly(df, x = ~V, y = ~W, z = ~U) %>%
  add_markers(marker = list(size = 5, color = "orange"))
```

```
z <- seq(0, 10, 0.2);
cols <- rep_len(c("orange", "blue", "green"), length.out = length(z))
df <- data.frame(U = seq(0, 10, 0.2), V = cos(z), W = sin(z)*z, color = cols)
plot_ly(df, x = ~V, y = ~W, z = ~U) %>%
  add_markers(marker = list(size = 5), color = ~cols)
```

Volcano dataset

- **volcano** - a built-in dataset storing topographic information for Maunga Whau (Mt Eden), one of 50 volcanos in Auckland, New Zealand.
- It consist of a 87×61 matrix with entries corresponding to the mountain's atlitudes [m] on a 10m by 10m grid.
- rows run east to west, and columns south to north

```
dim(volcano)
```

```
## [1] 87 61
```

```
volcano[1:5, 1:5]
```

```
##      [,1] [,2] [,3] [,4] [,5]
## [1,] 100  100  101  101  101
## [2,] 101  101  102  102  102
## [3,] 102  102  103  103  103
## [4,] 103  103  104  104  104
## [5,] 104  104  105  105  105
```

2D contour plots

```
plot_ly(z = volcano) %>% add_contour()
```

3D surface plots

```
plot_ly(z = volcano) %>% add_surface()
```

Heatmaps

Heatmap

- A *heatmap* is a popular graphical method for **visualizing high-dimensional data (table or matrix format)**
- Heatmaps display data (variables, correlations, sparsity/missing data pattern) as a grid of colored cells.
- They are commonly used in genomics papers to show **gene expression levels**,

The rows and the columns of the heatmaps are usually ordered to highlight the patterns in the data, and are usually accompanied by dendograms.

Scale mtcars data

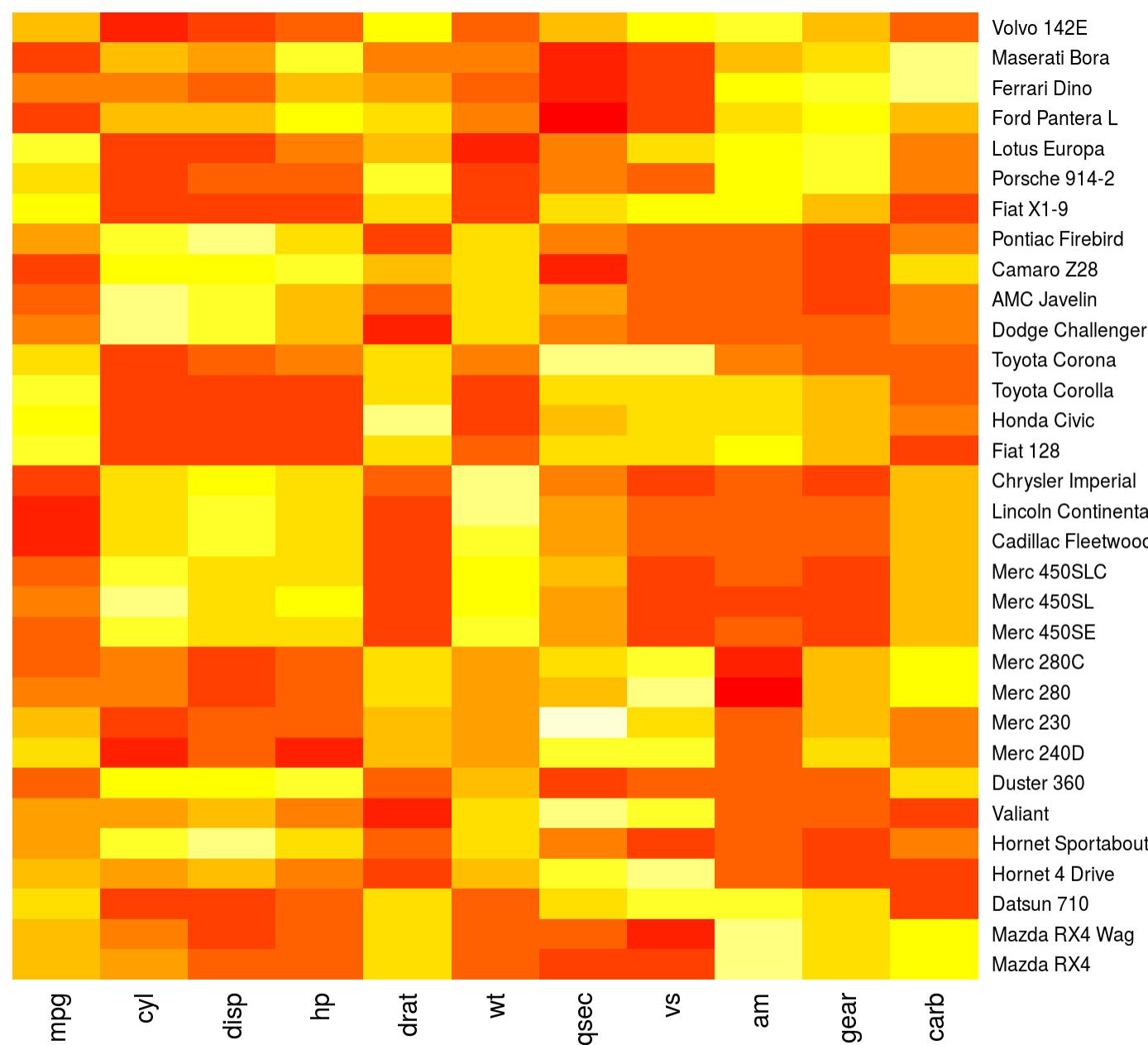
We convert each variable (column) in `mtcars` to Z-scores, i.e. we subtract the column mean and scale by its standard deviation.

```
mtscaled <- as.matrix(scale(mtcars))
head(mtscaled)
```

```
##          mpg        cyl      disp       hp     drat
## Mazda RX4  0.1508848 -0.1049878 -0.57061982 -0.5350928  0.5675137
## Mazda RX4 Wag 0.1508848 -0.1049878 -0.57061982 -0.5350928  0.5675137
## Datsun 710  0.4495434 -1.2248578 -0.99018209 -0.7830405  0.4739996
## Hornet 4 Drive 0.2172534 -0.1049878  0.22009369 -0.5350928 -0.9661175
## Hornet Sportabout -0.2307345  1.0148821  1.04308123  0.4129422 -0.8351978
## Valiant    -0.3302874 -0.1049878 -0.04616698 -0.6080186 -1.5646078
##          wt        qsec       vs       am     gear
## Mazda RX4 -0.610399567 -0.7771651 -0.8680278  1.1899014  0.4235542
## Mazda RX4 Wag -0.349785269 -0.4637808 -0.8680278  1.1899014  0.4235542
## Datsun 710 -0.917004624  0.4260068  1.1160357  1.1899014  0.4235542
## Hornet 4 Drive -0.002299538  0.8904872  1.1160357 -0.8141431 -0.9318192
## Hornet Sportabout 0.227654255 -0.4637808 -0.8680278 -0.8141431 -0.9318192
## Valiant    0.248094592  1.3269868  1.1160357 -0.8141431 -0.9318192
##          carb
## Mazda RX4   0.7352031
## Mazda RX4 Wag 0.7352031
## Datsun 710  -1.1221521
## Hornet 4 Drive -1.1221521
## Hornet Sportabout -0.5030337
## Valiant    -1.1221521
```

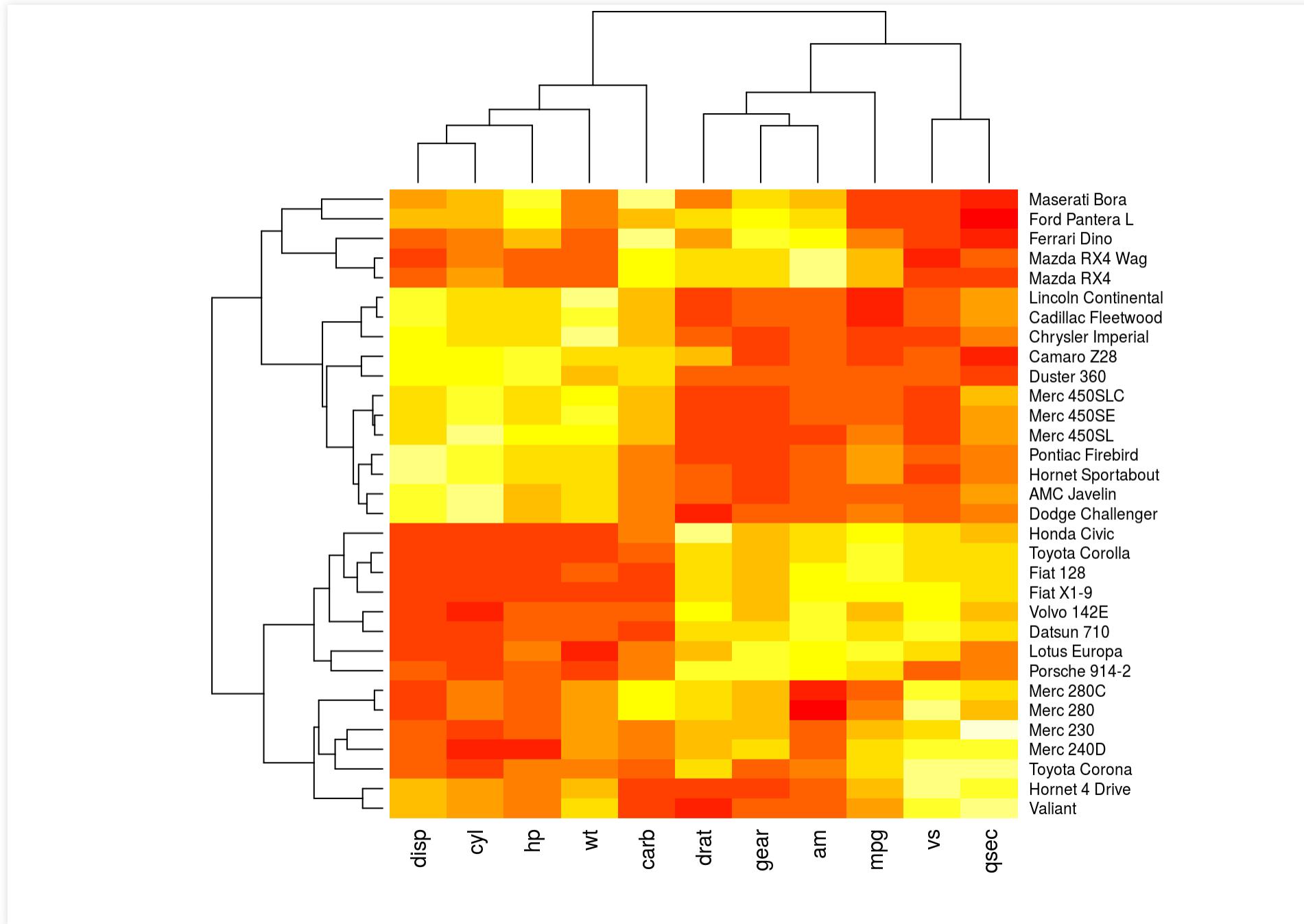
Generate a basic heatmap

```
# Using basic `stats:::heatmap` function:  
heatmap(mtscaled, Rowv = NA, Colv = NA)
```



Heatmap with dendograms

```
# heatmaps with dendograms and rows and columns reordered by means  
heatmap(mtscaled)
```

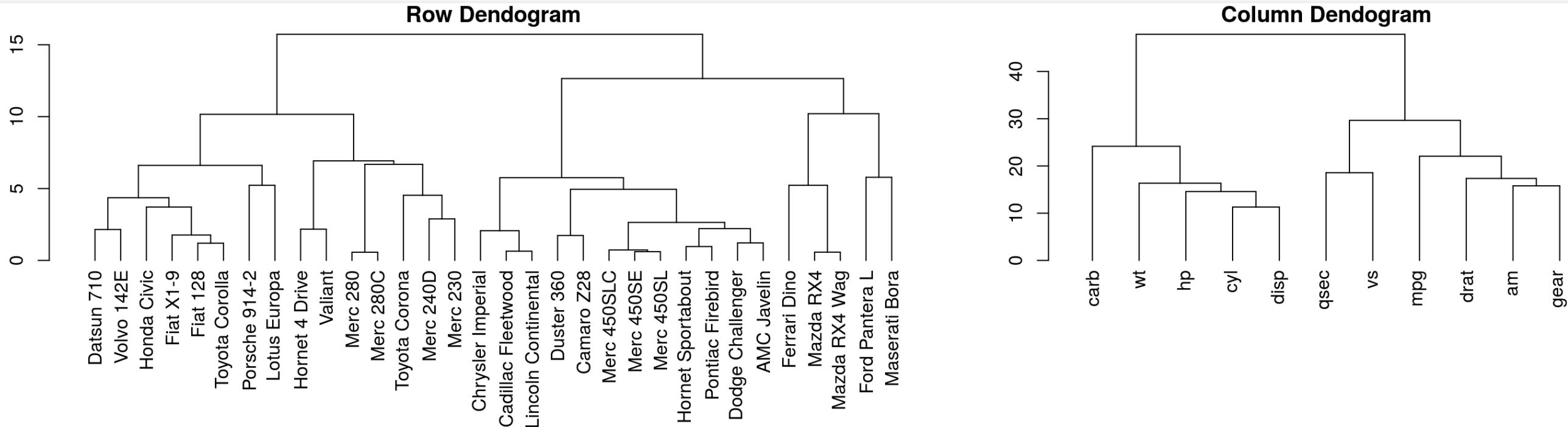


Compute dendograms

```
# Cluster rows (by default healmaps uses method = "euclidean")
row.clus <- hclust(dist(mtscaled), method = "manhattan"), "aver")
row.dend <- as.dendrogram(row.clus)

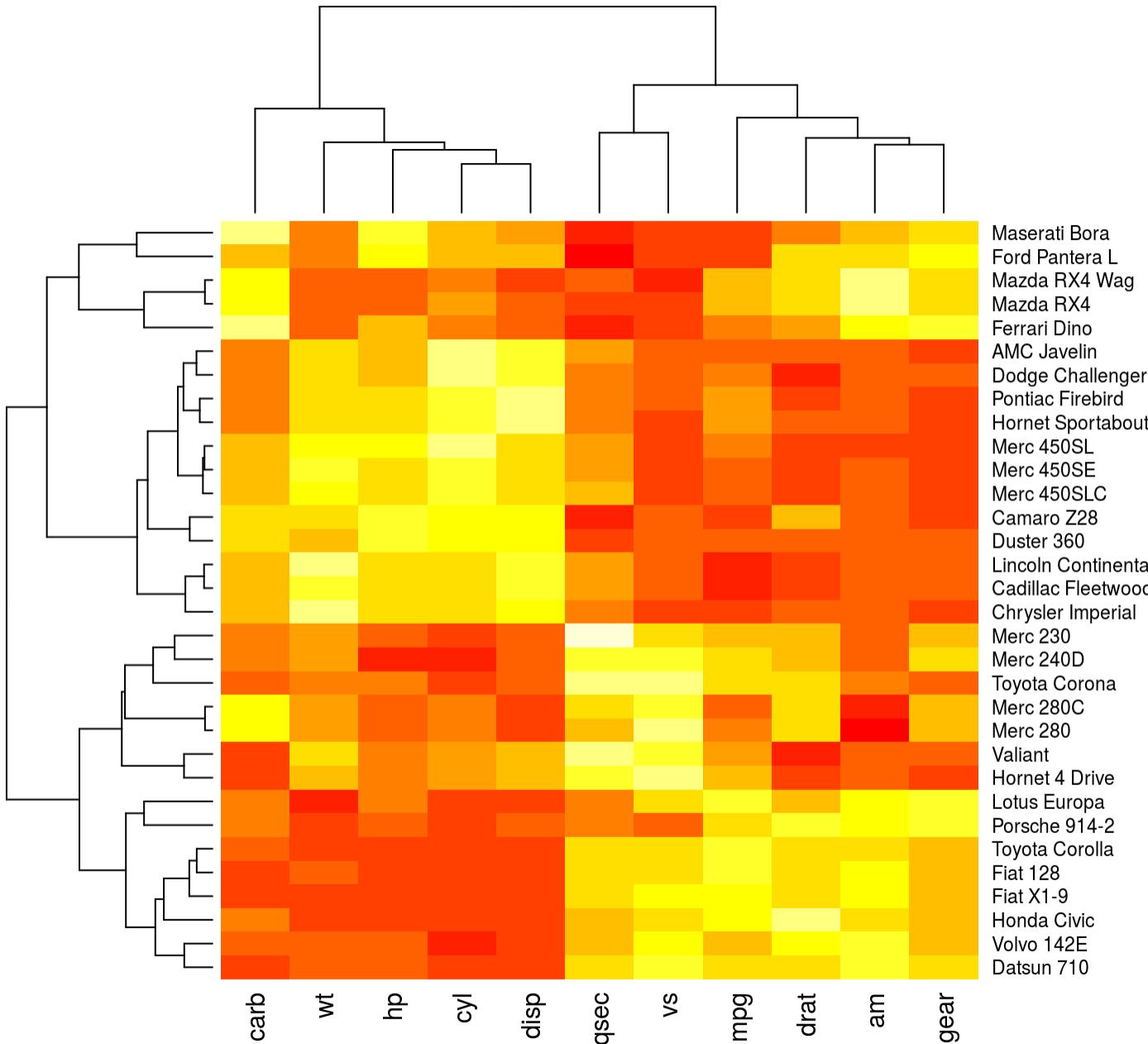
# Cluster columns
col.clus <- hclust(dist(t(mtscaled)), method = "manhattan"), "aver")
col.dend <- as.dendrogram(col.clus)

# Plot dendograms:
layout(matrix(c(1,2), nrow=1), widths=c(3,2)); par(mar = c(8,4,1,1))
plot(row.dend, xlab = "", main = "Row Dendogram", sub = "")
plot(col.dend, xlab = "", main = "Column Dendogram", sub = "")
```

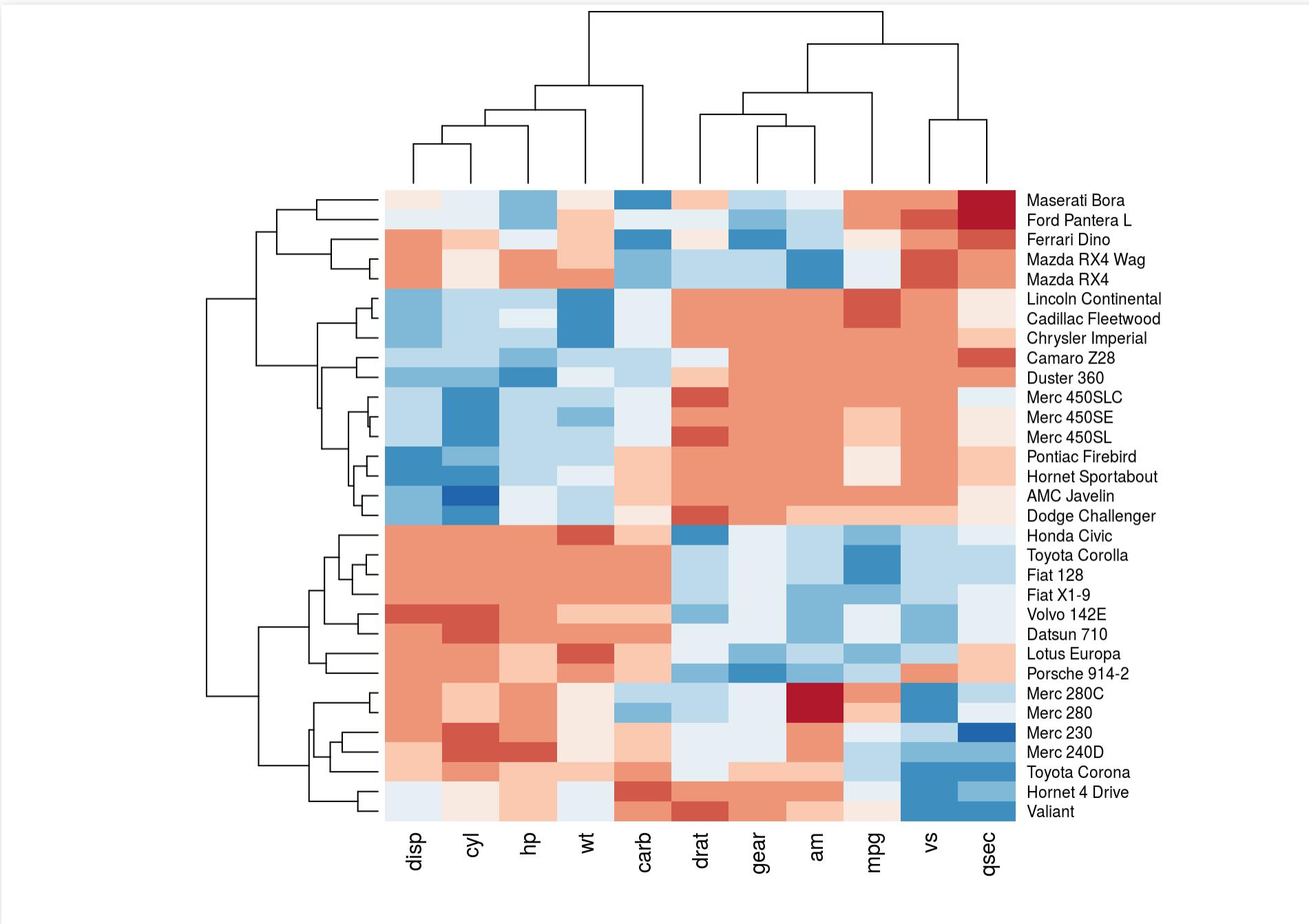


Heatmaps with customized dendograms

```
# heatmaps with dendograms and reordered (by means) rows and columns  
heatmap(mtscaled, Rowv = row.dend, Colv = col.dend)
```



```
library(RColorBrewer)
# Define color scheme
scaleredblues <- colorRampPalette(colors = brewer.pal(9, "RdBu"))(10)
# "#B2182B" "#D25849" "#ED9576" "#FAC9B0" "#F9EAE1" "#E6EFF3" "#BCDAEA" "#80B9D8
heatmap(as.matrix(mtscaled), col = scaleredblues)
```



heatmaply package for interactive heatmap

```
library(heatmaply)
heatmaply(mtscaled) %>% layout(margin = list(l = 150, b = 50))
```

Color the dendogram branches:

```
heatmaply(mtscaled, k_col = 2, k_row = 3) %>%  
  layout(margin = list(l = 150, b = 50), autosize = F, width = 600, height = 700)
```

ggmap for visualizing spatial data

Part 1: Downloading the map raster

1. Define location: 3 ways

- location/address
- lat/long
- bounding box

2. Define map source, type, and color with

`get_map():`

- Google Maps (“google”),
- OpenStreetMap (“osm”),
- Stamen Maps (“stamen”),
- CloudMade maps (“cloudmade”)

3. You can set the scale of the map using

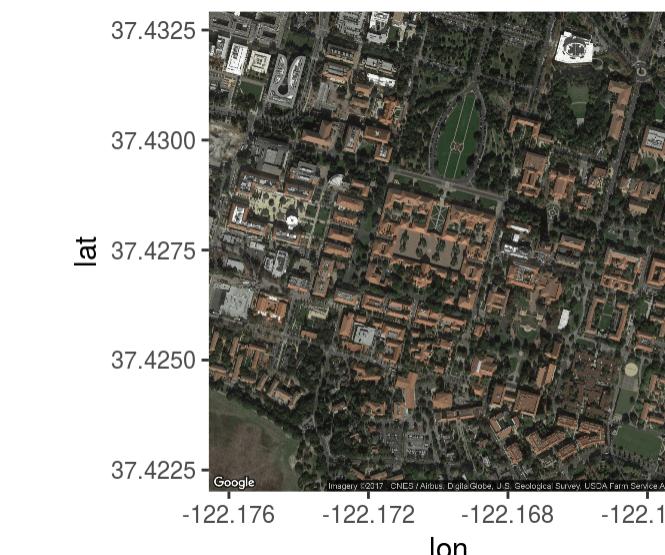
`zoom`. `get_map()` takes a guess at the zoom level, but you can alter it.

- 3 (continent) to 21 (building), default value 10 (city).
- “openstreetmap” and “stamen” limit is 0-18.

```
library(ggmap)  
geocode("Stanford University")
```

```
##          lon      lat  
## 1 -122.1697 37.42747
```

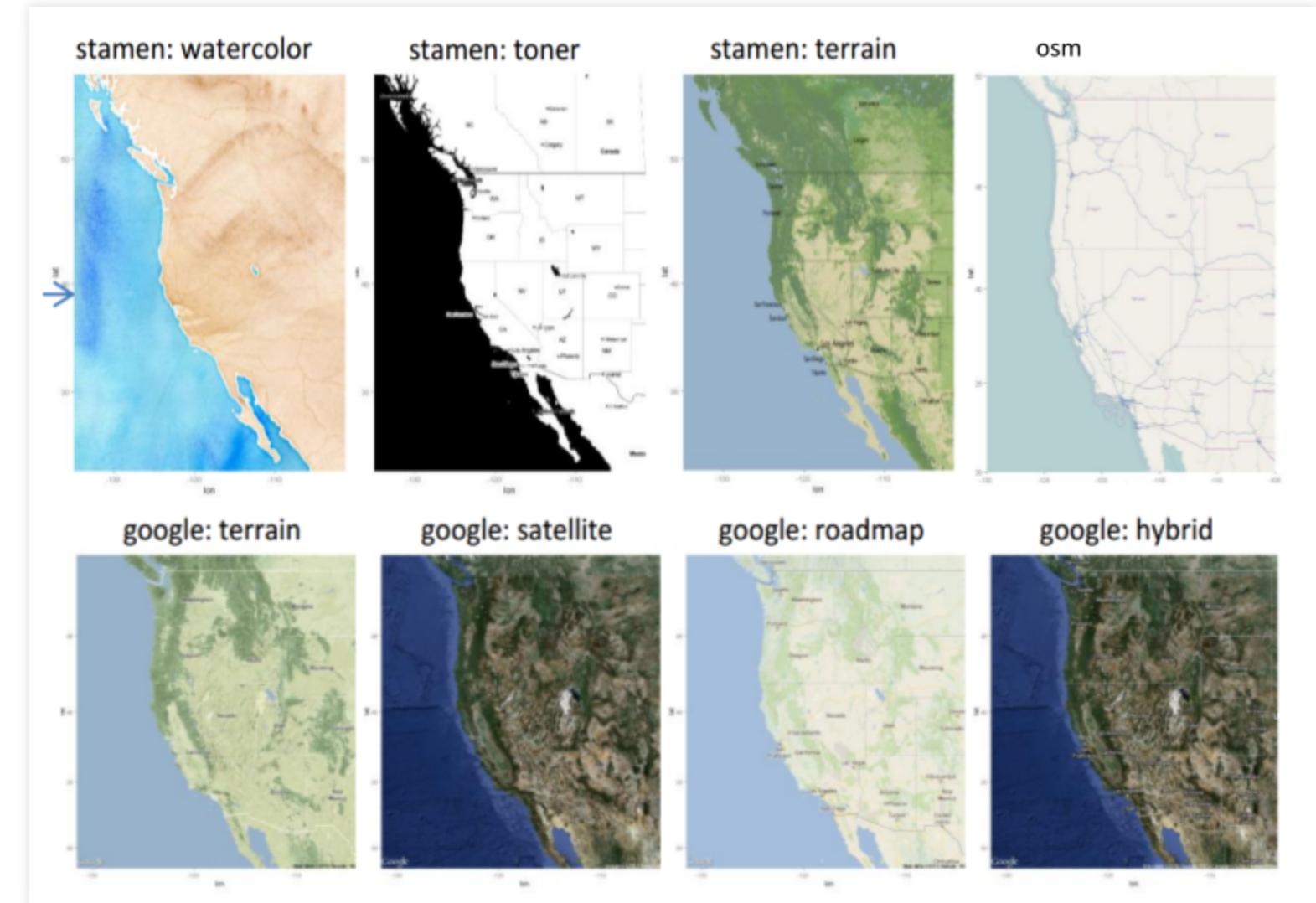
```
myLoc <- "Stanford University" # or myLoc <- c(-125,  
# or myLoc <- c(lon = -122.1697, lat = 37.42747)  
myMap <- get_map(location = myLoc, crop = TRUE, zoom  
                  source = "google", maptype="satellite")  
ggmap(myMap)
```



Map sources

There are different map sources to obtain a map raster, and each of these sources has multiple “map types”. Here we show examples of maps from:

- **stamen**: `maptype = c("terrain", "toner", "watercolor")`
- **google**: `maptype = c("roadmap", "terrain", "satellite", "hybrid")`
- **osm**: [open street map](#)



```
US <- c(left = -125, right = -67, bottom = 25.75, top = 49)
us_map <- get_map(location = US, source = "stamen", maptype = "toner-lite", zoom = 4)
class(us_map)
```

```
## [1] "ggmap"    "raster"
```

```
ggmap(us_map)
```



```
europe <- c(left = -12, right = 30, bottom = 35.0, top = 63)
europe_map <- get_stamenmap(bbox = europe, zoom = 5, maptype = "watercolor")
ggmap(europe_map)
```

