Visualisation with ggplot2

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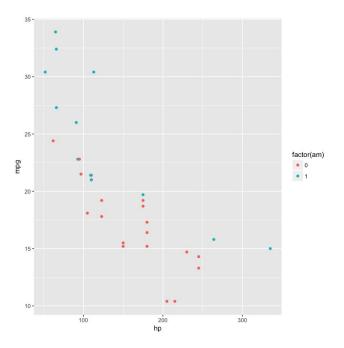
ggplot2 definition

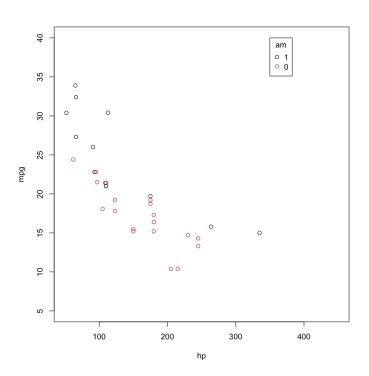
ggplot2 (grammar of graphics) is a data visualization package for R which splits graphs into semantic components such as scales and layers.

Created by Hadley Wickham in 2005.

Why using ggplot2?

- Consistent underlying grammar of graphics (Wilkinson, 2005)
- Plot specification at a high level of abstraction
- Theme system to refine plot appearance
- Very active development





ggplot2 (left) vs basic-graphics (right)

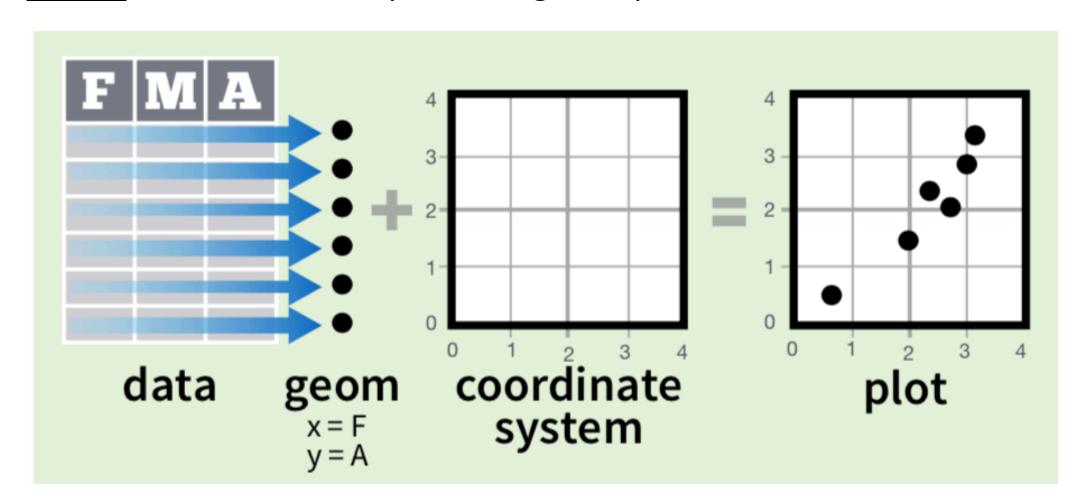
ggplot2 vs base-graphics

```
library(ggplot2)
ggplot(mtcars, aes(x=hp, y=mpg,
color=factor(am))) + geom_point()
```

```
par(mar = c(4,4,.1,.1))
plot(mpg ~ hp,
data=subset(mtcars, am==1), xlim=
c(50, 450), vlim=c(5, 40)
points(mpg ~ hp, col="red",
data=subset(mtcars, am==0))
legend(350, 40,
c("1", "0"), title= "am", col=
c("black", "red"), pch= c(1, 1))
```

Build every graph from:

- A data set
- A <u>coordinate</u> system
- geoms visual marks representing data points



Concept of ggplot2

Specifying a plot via building blocks:

- Data
- Aesthetic mapping
- Geometric object

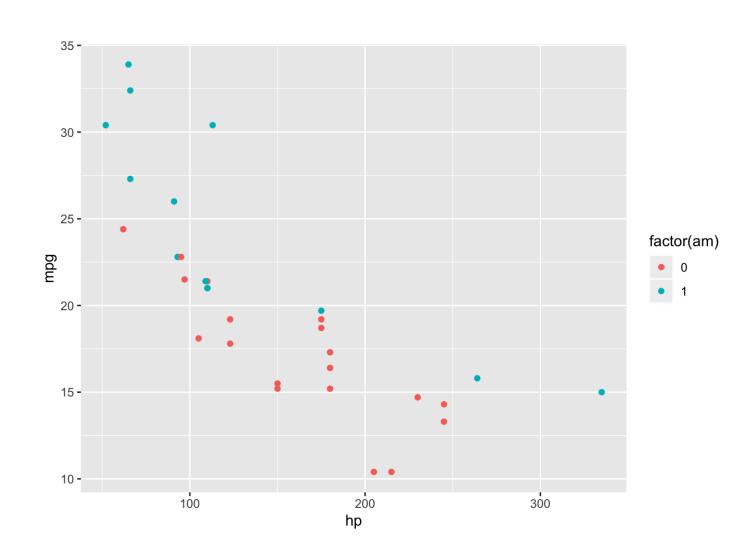
- Statistical transformations
- Scales
- Coordinate system
- Position adjustments
- Faceting

A plot includes multi layers

- Data
- Mappings between variables and aesthetics
- geometric object
- statistical transformation
- Scales to control the details of the mapping

```
library(ggplot2)
ggplot(mtcars, aes(x=hp,y=mpg,color=factor(am)))+geom_point()
```

- initializes a ggplot object
- declare the input data frame for a graphic
- specify the set of plot aesthetics intended to be common throughout all subsequent layers unless specifically overridden.

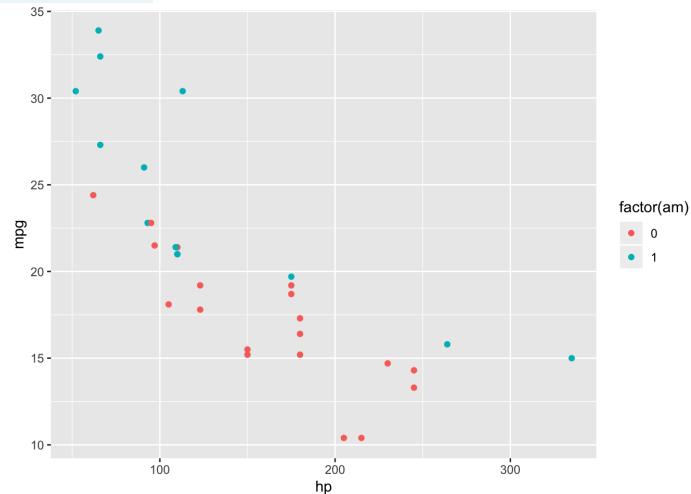


```
library(ggplot2)
ggplot(mtcars, aes(x=hp,y=mpg,color=factor(am)))+geom_point()
```

```
geom_point(mapping = NULL, data = NULL, stat = "identity",
   position = "identity", ..., na.rm = FALSE, show.legend = NA,
   inherit.aes = TRUE)
```

- Create scatterplots
- Useful for displaying the relationship between two continuous variables

- mapping Set of aesthetic mappings created by <u>aes()</u>
- data The data to be displayed in this layer
- **stat** The statistical transformation to use on the data for this layer, as a string.
- ... Other arguments passed on to <u>layer()</u>.
- na.rm If FALSE, the default, missing values are removed with a warning
- **show.legend** Should this layer be included in the legends?
- **inherit.aes** If FALSE, overrides the default aesthetics, rather than combining with them.



ggplot2 structure

- Aesthetics and geometric Objects
- Statistical Transformations
- Scales
- Faceting
- Themes

Aesthetics and geometric Objects

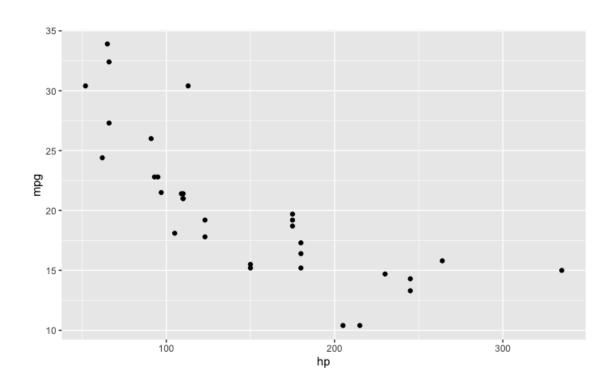
Aesthetics

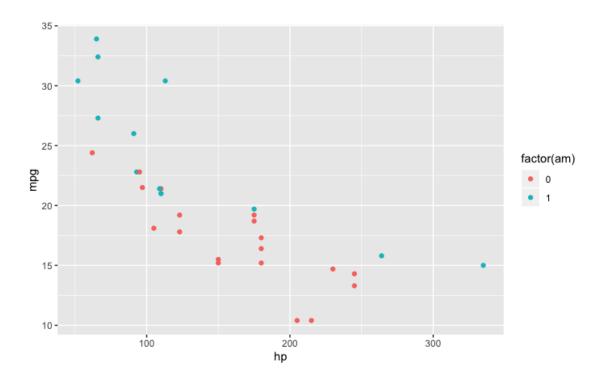
Aesthetic mappings describe how variables in the data are mapped to visual properties (aesthetics) of geoms

- Set with aes() function
- Position (x and y)
- Color (outline color)
- Fill
- Shape (point shapes)
- Line type
- Size

aes(x, y, ...)

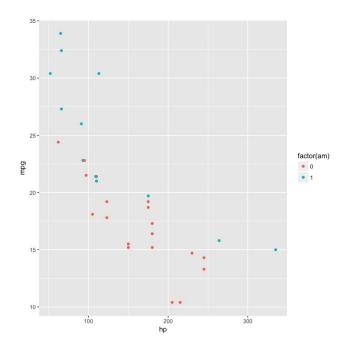
ggplot(mtcars, aes(x=hp,y=mpg))
 +geom_point()





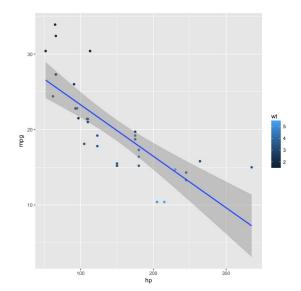
Geometric

- Points (geom_point: dot plots, scatter plots)
- Lines (geom_line: trend lines, time-series)
- Boxplot (geom_boxplot)
- At least one geom is required (no upper limit)
- Added to the plot using the + operator

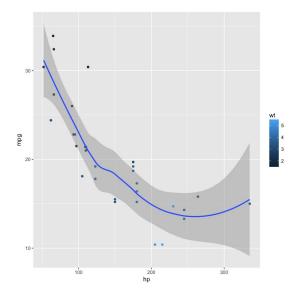


geom_smooth / geom_line

```
ggplot(mtcars, aes(x = hp, y = mpg)) +
geom_point(aes(color = wt)) +
geom_smooth(method = lm)
```

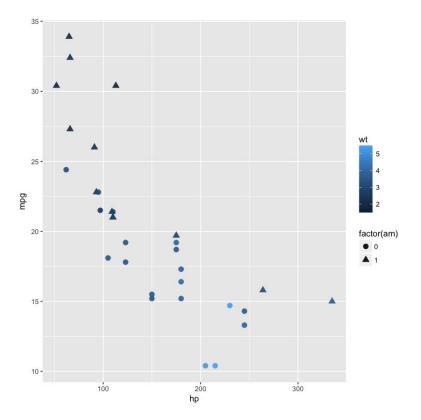


```
ggplot(mtcars, aes(x = hp, y = mpg)) +
geom_point(aes(color = wt)) +
geom_smooth(method = loess)
```



Map variables to aesthetics

```
ggplot(mtcars, aes(x= hp, y= mpg)) +
geom_point(aes(color= wt, shape= factor(am)),
size= 3, alpha= .9)
```



Statistical Transformations

Default Statistics for geom

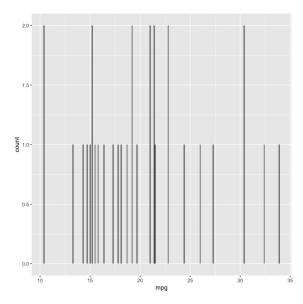
```
args(geom_bar)
function (mapping = NULL, data = NULL, stat = "bin", position = "stack", ...)

args(geom_boxplot)
function (mapping = NULL, data = NULL, stat = "boxplot", position = "dodge",
outlier.color = "black", outlier.shape = 16, outlier.size = 2,
notch = FALSE, notchwidth = 0.5, ...)

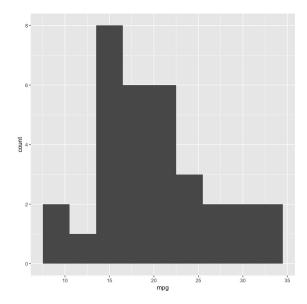
args(geom_histogram)
function (mapping = NULL, data = NULL, stat = "bin", position = "stack", ...)
```

Arguments that can be customised

```
ggplot(mtcars, aes(x= mpg)) +
geom_bar()
```



```
ggplot(mtcars, aes(x= mpg)) +
geom_histogram(stat= "bin", binwidth= 3)
```



Scales

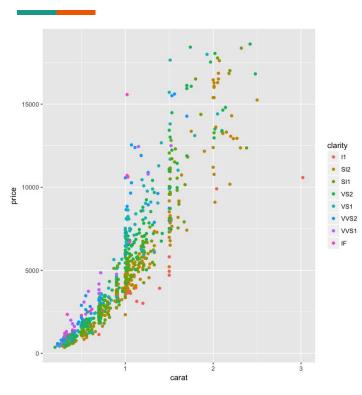
The scale contains:

position, color, shape, size and line type

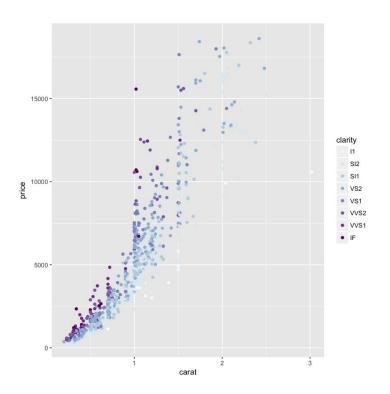
Range - Colour

```
dSample<-
diamonds[sample(nrow(diamonds),
1000),]
diamond <- ggplot(dSample,
aes(x= carat, y= price, color=
clarity)) + geom_point()
diamond</pre>
```

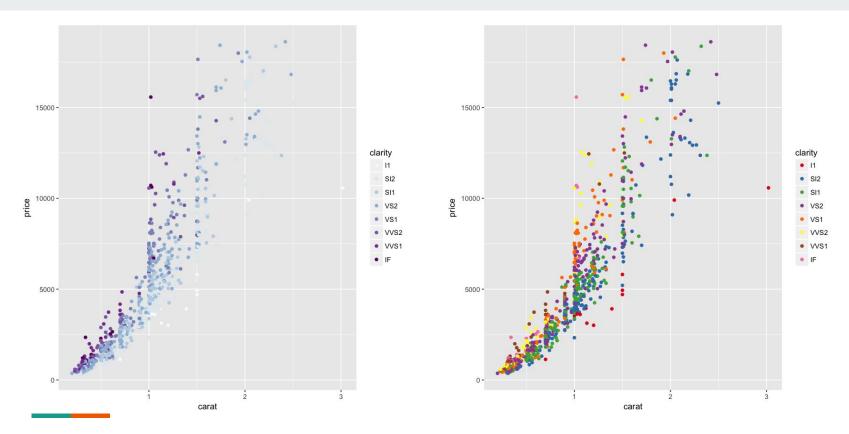
```
diamond +
scale_color_brewer(type= "seq",
palette= 3)
```



dSample<- diamonds[sample(nrow(diamonds), 1000),]
diamond <- ggplot(dSample, aes(x= carat, y= price,
color= clarity)) + geom_point()
diamond</pre>



diamond + scale_color_brewer(type= "seq",
palette= 3)

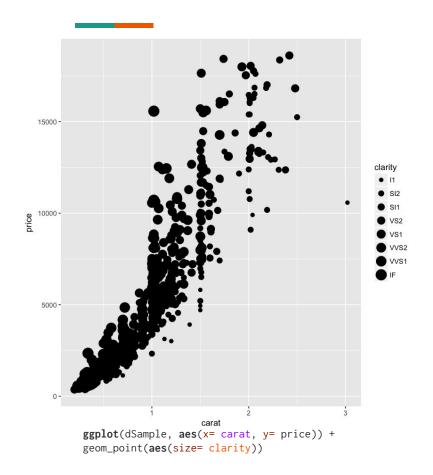


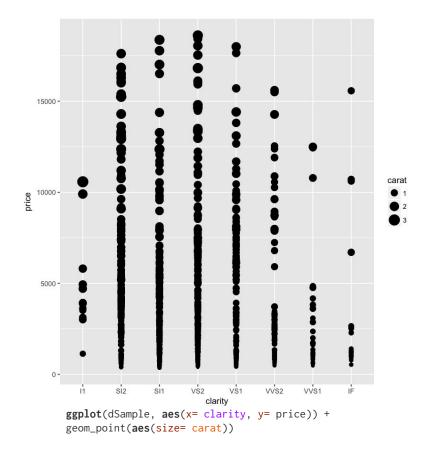
Using the sequential compare to the categorical colour

Range - Shapes

```
ggplot(dSample, aes(x= carat,
y= price)) +
geom_point(aes(size= clarity))
```

```
ggplot(dSample, aes(x=
  clarity, y= price)) +
geom_point(aes(size= carat))
```



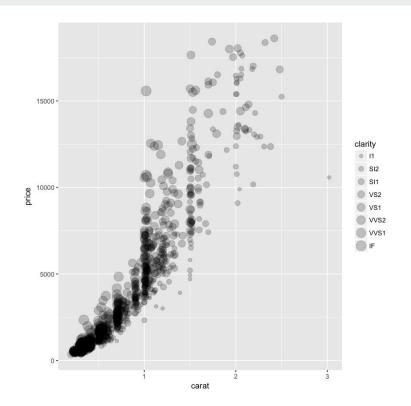


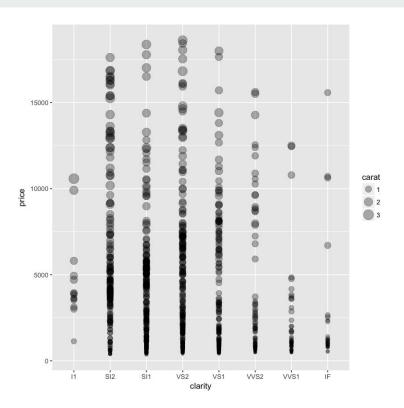
Question

Can we improve the plot?

Yes, by adding alpha=0.5 to geom_point.

```
geom_point(aes(size= clarity), alpha= .5)
```



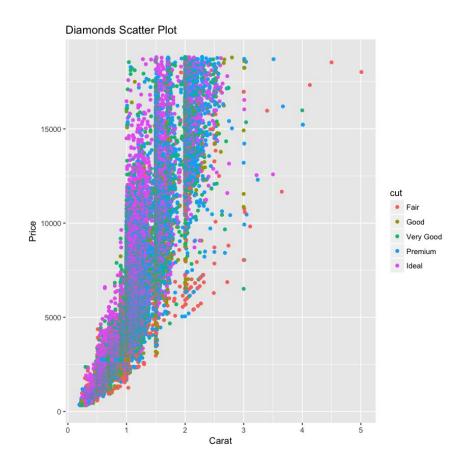


Using alpha to overcome the overlapping issue geom_point(aes(size= clarity), alpha= .5)

Facet

Without Facet

```
ggplot(diamonds, aes(x= carat,
y= price, color= cut)) +
geom_point()
```

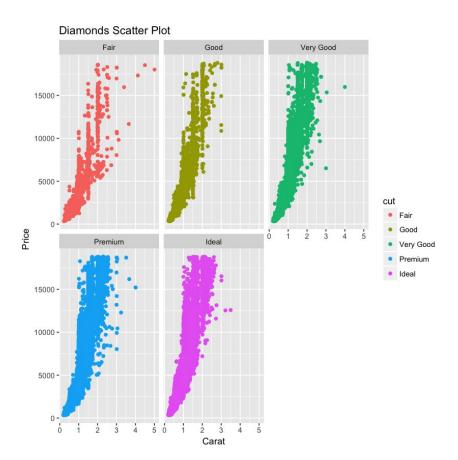


Try it

```
Add + facet_wrap(~ cut, ncol = 3) to the main ggplot2.
```

Adding Facet

```
ggplot(diamonds, aes(x= carat, y=
price, color= cut)) +
geom_point() + facet_wrap(~ cut,
ncol= 3)
```



Theme

The theme handles non-data elements within the plot:

- Axis labels
- Facet label background
- Legend
- Plot background

More details:

https://github.com/hadley/ggplot2/

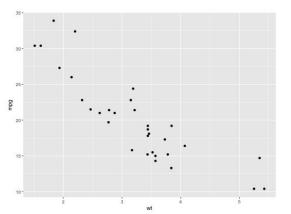
wiki/Themes

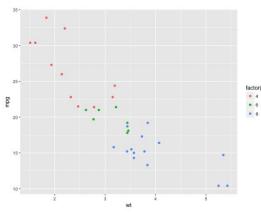
https://github.com/jrnold/ggthemes

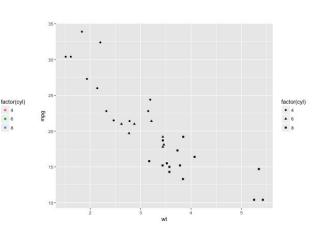
ggplot2 Practice

We are going to do coding...

Let's plot the mtcars dataset





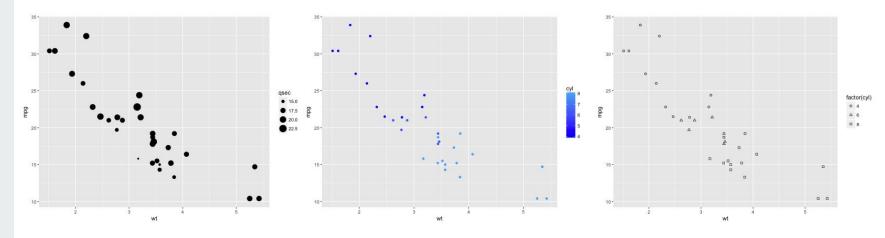


```
p <- ggplot(mtcars, aes(wt,
mpg))
p + geom_point()</pre>
```

```
p + geom_point(aes(colour =
factor(cyl)))
```

```
p + geom_point(aes(shape =
factor(cyl)))
```

Now use size, colour, shape

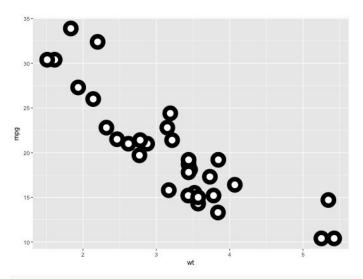


```
p + geom_point(aes(size =
qsec))
```

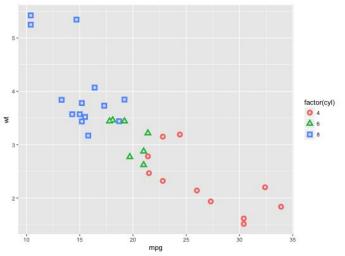
```
p + geom_point(aes(colour =
cyl))
```

```
p + geom_point(aes(shape =
factor(cyl))) +
scale_shape(solid = FALSE)
```

Now customise the colour/shape

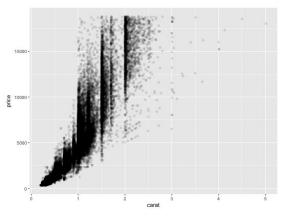


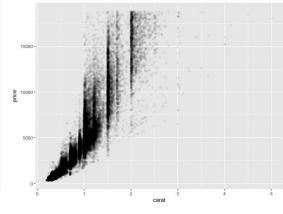
```
ggplot(mtcars, aes(wt, mpg)) +
  geom_point(shape= 21, colour= "black",
fill= "white", size= 5, stroke= 5)
```

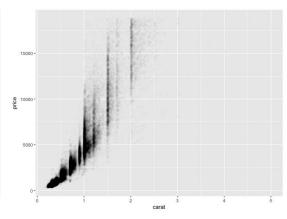


```
ggplot(mtcars, aes(mpg, wt, shape=
factor(cyl))) + geom_point(aes(colour=
factor(cyl)), size= 4) + geom_point(colour=
"grey90", size= 1.5)
```

Plot large-scale diamonds datasets







```
d + geom_point(alpha= 1/20)
```

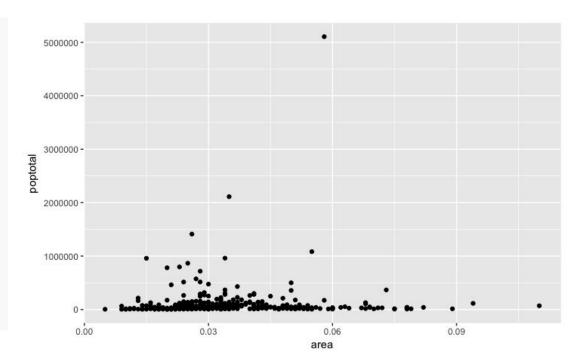
```
d <- ggplot(diamonds,
aes(carat, price))
d + geom_point()</pre>
```

```
d + geom_point(alpha= 1/100)
```

Scatterplot - Simple

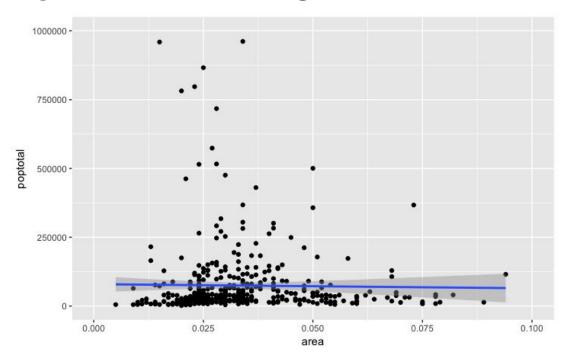
```
library(ggplot2)
theme_set(theme_bw())
data("midwest", package =
  "ggplot2")

g <- ggplot(midwest,
  aes(x=area, y=poptotal)) +
  geom_point()</pre>
```



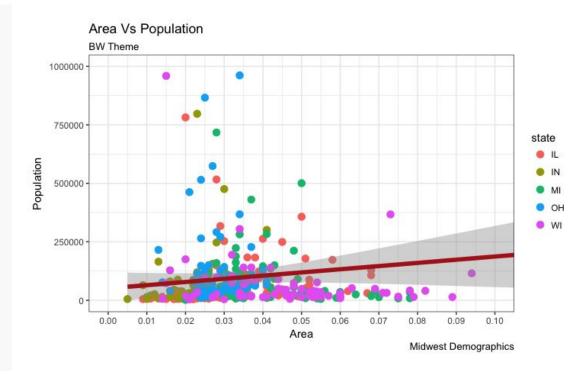
Scatterplot - Simple + line fitting

```
library(ggplot2)
theme_set(theme_bw())
data("midwest", package =
"ggplot2")
g <- ggplot(midwest,</pre>
aes(x=area, y=poptotal)) +
geom_point() +
geom_smooth(method="lm")
# Not outside the limits
g + x \lim(0, 0.1) + y \lim(0, 0.1)
1000000)
```



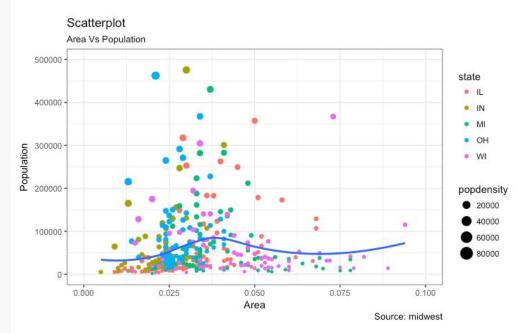
Scatterplot - Simple + theme

```
library(ggplot2)
theme_set(theme_bw())
data("midwest", package =
"ggplot2")
g <- ggplot(midwest,</pre>
aes(x=area, y=poptotal)) +
geom_point(aes(col=state),
size=3) +
geom_smooth(method="lm",
col="firebrick", size=2
g + xlim(0, 0.1) + ylim(0,
1000000)
g + theme_bw()
```



Scatterplot - Pro

```
library(ggplot2)
theme_set(theme_bw())
data("midwest", package = "ggplot2")
gg <- ggplot(midwest, aes(x=area,</pre>
y=poptotal)) +
  geom_point(aes(col=state,
size=popdensity)) +
  geom_smooth(method="loess", se=F) +
  xlim(c(0, 0.1)) +
  ylim(c(0, 500000)) +
  labs(subtitle="Area Vs Population",
       y="Population",
       x="Area",
       title="Scatterplot",
       caption = "Source: midwest")
plot(gg)
```



Thank you

Useful sources

http://r-statistics.co/ggplot2-Tutorial-With-R.html

http://r-statistics.co/Complete-Ggplot2-Tutorial-Part1-With-R-Code.html

http://r-statistics.co/Complete-Ggplot2-Tutorial-Part2-Customizing-Theme-With-R-Code.html

http://r-statistics.co/Top50-Ggplot2-Visualizations-MasterList-R-Code.html

http://r-statistics.co/ggplot2-cheatsheet.html

http://tutorials.iq.harvard.edu/R/Rgraphics/Rgraphics.html

References

Kevin Davenport, Data visualization with R, San Diego R Users Group, June 2013.

Jessie Kennedy, Principles of Information Visualization Tutorial, Institute for Informatics & Digital Innovation, Edinburgh Napier University, 2012.

Martin Krzywinski, visual design principles, Vizbi 2013.

Jeffrey Heer, Interactive Data Analysis, University of Washington.

R. Jordan Crouser, Introduction to Visual Analytics, Computer Science Department at Tufts University, 2015.

Colin Ware. 2004. Information Visualization: Perception for Design. Morgan Kaufmann Publishers Inc., San Francisco, CA, USA.

Boris Naujoks, Jörg Stork, Martin Zaefferer, Thomas Bartz-Beielstein, Meta-Model Assisted (Evolutionary) Optimization Tutorial at PPSN 2016, 18.09.2016 Hadley's ggplot2 book http://amzn.com/0387981403

ggplot2 google group http://groups.google.com/group/ggplot2

stackoverflow http://stackoverflow.com/tags/ggplot2

Lattice to applot2 conversion

http://learnr.wordpress.com/?s=lattice

Winston Chang's Cookbook for common graphics http://wiki.stdout.org/rcookbook/Graphs/