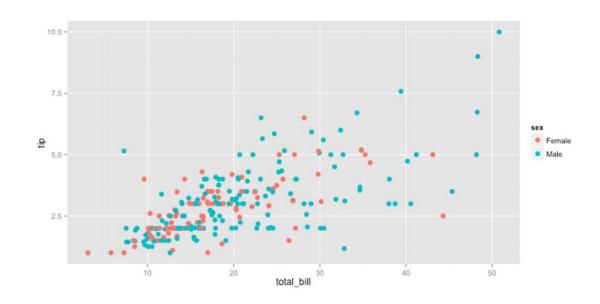


#### Three Main Components of Graph

- Primary data component design
  - Bars, dots, lines, etc.
- Secondary data component design
  - Text, grid lines, sticks, legends
- Non-data component design
  - Background

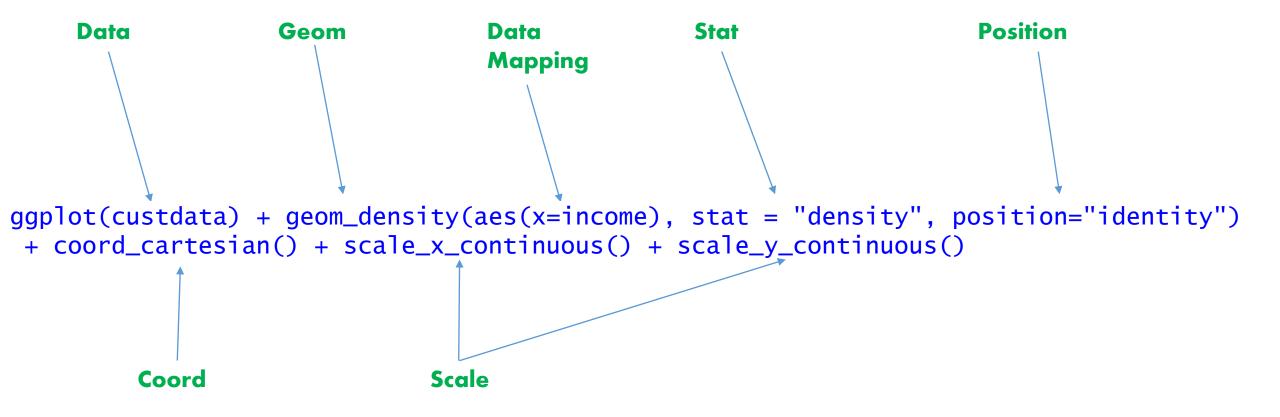


#### Components of ggplot2

- A default dataset with aesthetic mappings,
- One or more layers, each with a geometric object ("geom"), a statistical transformation ("stat"), and a dataset with aesthetic mappings (possibly defaulted),
- A scale for each aesthetic mapping (which can be automatically generated),
- A coordinate system, and
- A facet specification.

#### Grammar of Graphics

- Data source
- Data mapping
- Geom
- Coord
- Scale
- Facet ? Layer
- Stat
- Position



#### Common Geoms

- geom\_Line
- geom\_point
- geom\_bar
- geom\_density
- geom\_histogram
- geom\_smooth
- geom\_abline
- geom\_boxplot
- geom\_hex

#### Common stat

- identity
- count
- sum
- unique
- spoke
- function
- summary

#### Geom specific stat

- bin
- contour
- abline
- Density
- Boxplot
- qq
- smooth
- quantile

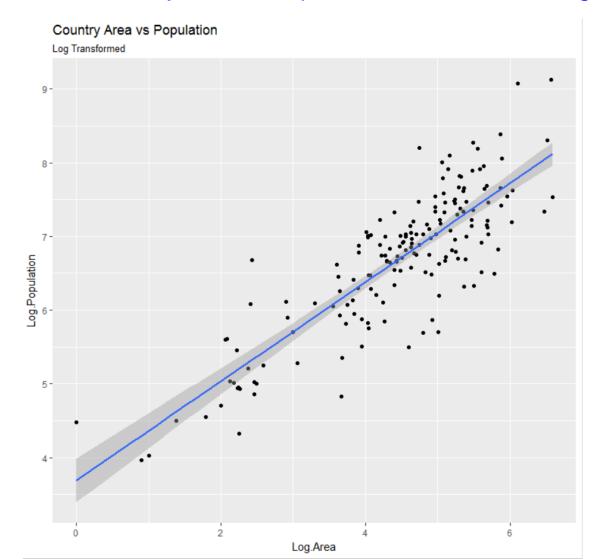
## position

- identity
- jitter
- dodge
- stack
- fill

# Chart Components

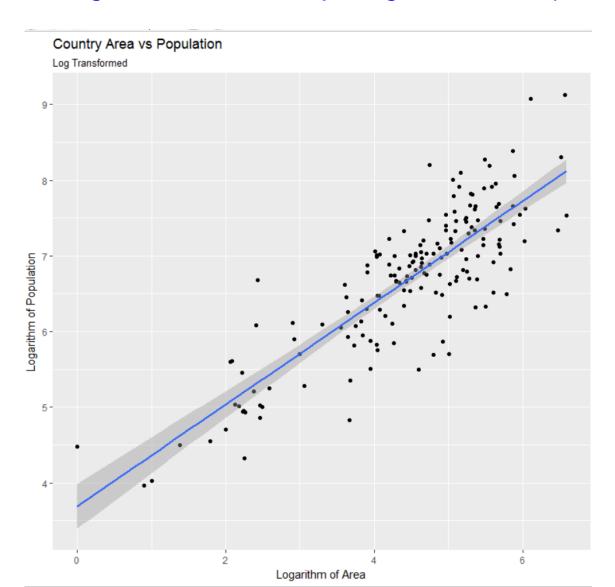
#### Title

```
g1 <- ggplot(df,aes(x=Log.Area,y=Log.Population)) + geom_point() + geom_smooth(method="lm")
g2 <- g1 + labs(title = "Country Area vs Population", subtitle = "Log Transformed")</pre>
```



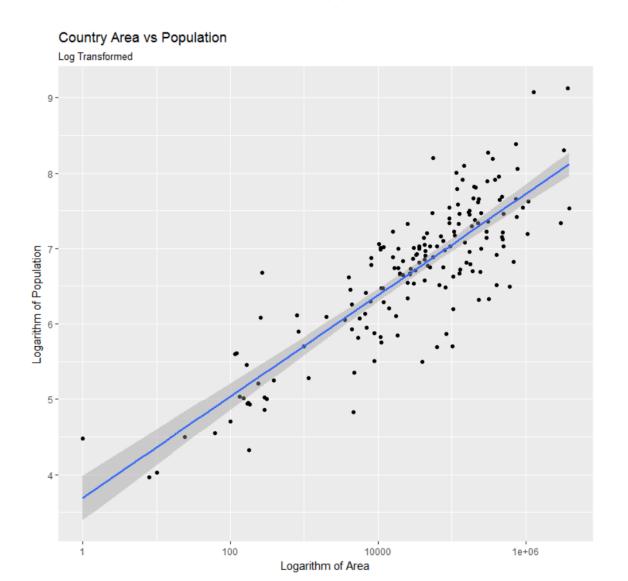
#### Axis

g3 <- g2 + labs(x="Logarithm of Area", y="Logarithm of Population")



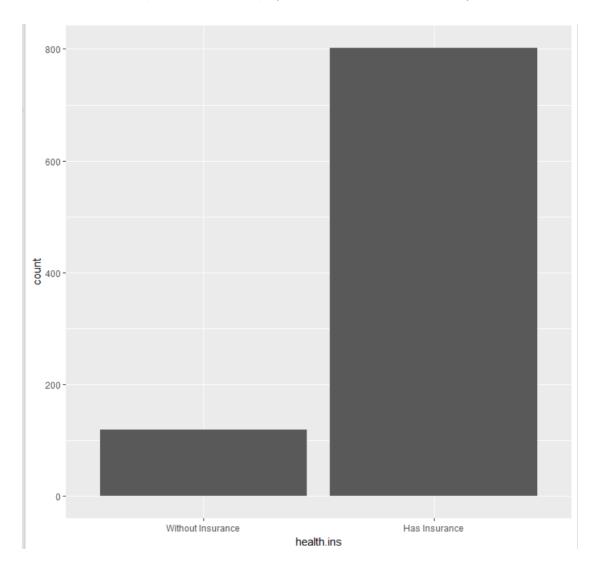
#### Change Labels with Function

 $g4 \leftarrow g3 + scale_x_continuous(labels = function(x) 10^x)$ 



#### Another Example

```
ggplot(custdata) + geom_bar(aes(x=health.ins))
+scale_x_discrete(labels = function(x) ifelse(x, "Has Insurance", "Without Insurance"))
```



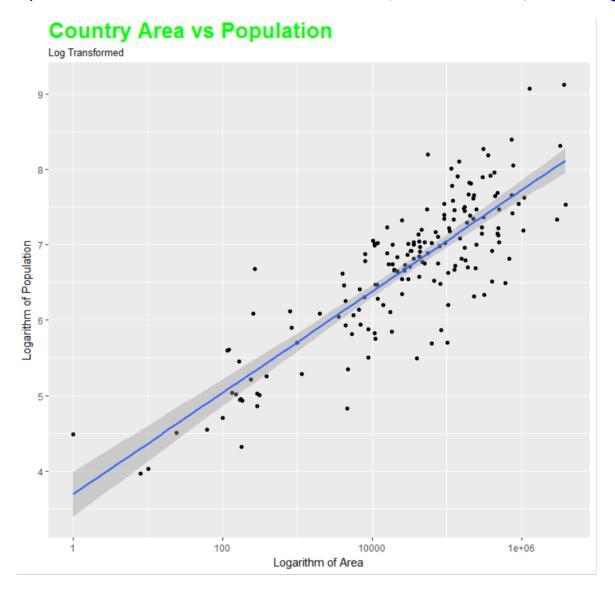
#### Title Theme

```
g5 <- g4 + theme(title=element_text(size=20,face="bold",color="green"))</pre>
```



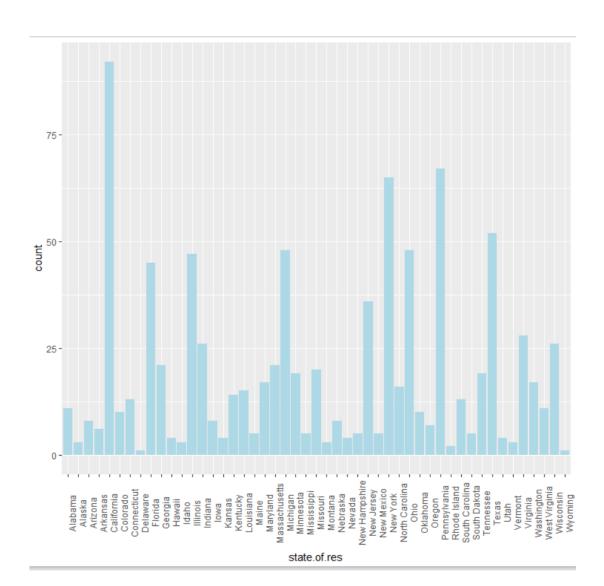
# Change only main title

g5 <- g4 + theme(plot.title =element\_text(size=20,face="bold",color="green"))</pre>



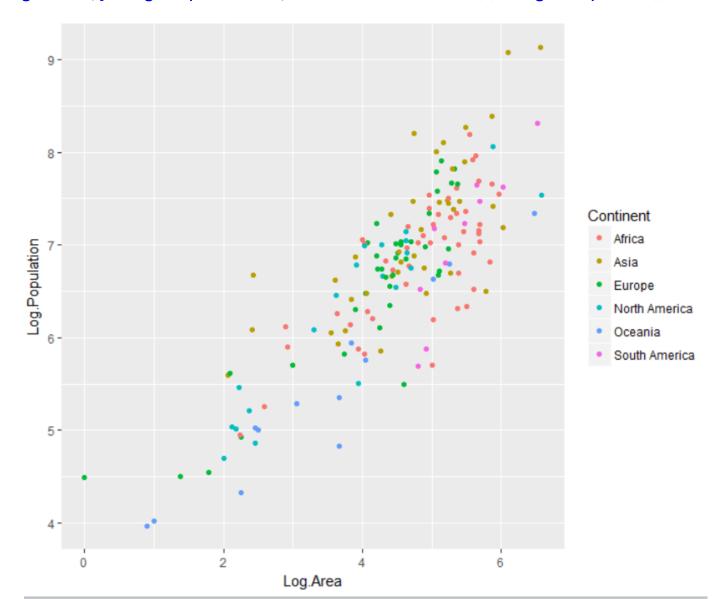
#### Tick text theme

ggplot(custdata) + geom\_bar(aes(x=state.of.res), fill="lightblue") + theme(axis.text.x=element\_text(angle=90))



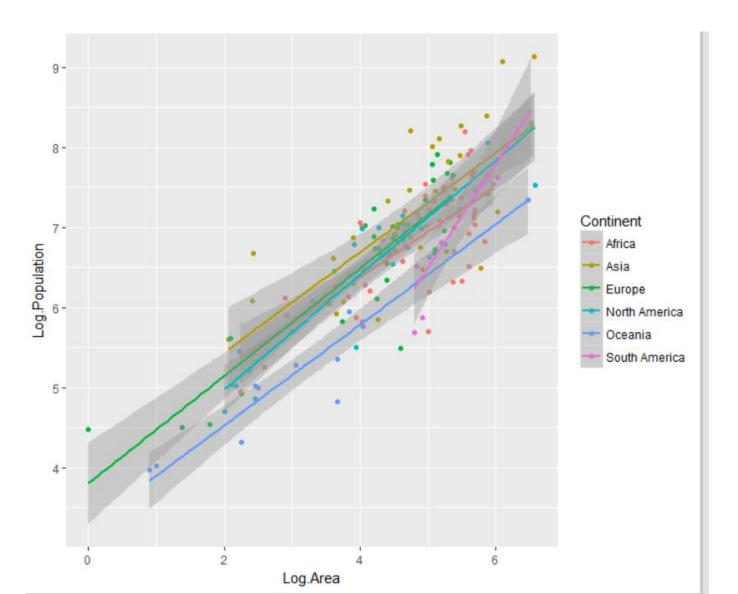
## Colouring by factor

ggplot(df,aes(x=Log.Area,y=Log.Population, colour=Continent)) + geom\_point()



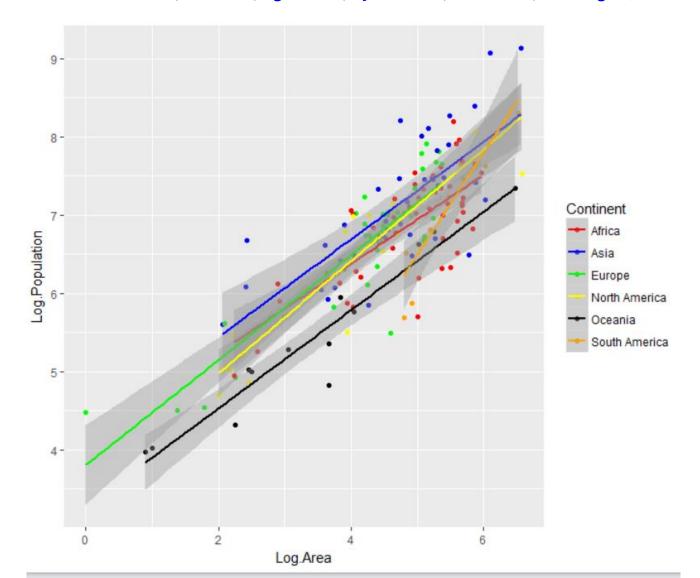
## What can you tell?

ggplot(df,aes(x=Log.Area,y=Log.Population, colour=Continent)) + geom\_point() + geom\_smooth(method="lm")



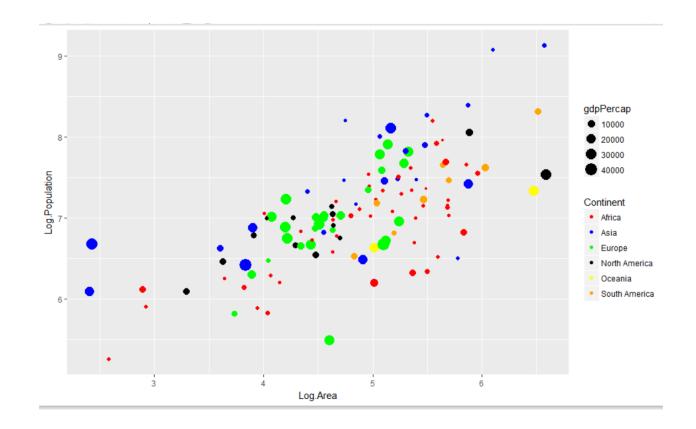
#### Set your own colours

ggplot(df,aes(x=Log.Area,y=Log.Population, colour=Continent)) + geom\_point() + geom\_smooth(method="lm")
+ scale\_color\_manual(values = c("red","blue","green","yellow","black","orange"))



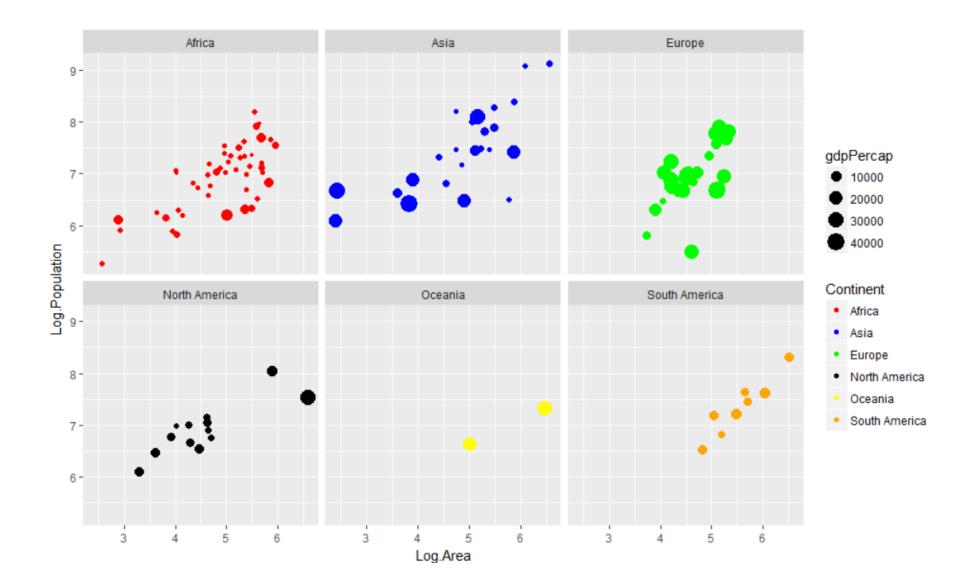
#### Sizing by factor

```
data<-read.csv("gapminder.csv") to country
data <- data[data$year=1997,]
df1 <- merge(df,data,by.x="Country",by.y="country")
ggplot(df1,aes(x=Log.Area,y=Log.Population, colour=Continent, cex=gdpPercap)) + geom_point()
+ scale_color_manual(values = c("red","blue","green","black","yellow","orange"))</pre>
```



#### Breaking graph by factor

ggplot(df1,aes(x=Log.Area,y=Log.Population, colour=Continent, cex=gdpPercap)) + geom\_point()
+ scale\_color\_manual(values = c("red","blue","green","black","yellow","orange")) + facet\_wrap(~Continent)



#### Reshape2 package

```
1991
                                                                                       Male 2714.489
                                                                              1992
                                                                                       Male 2997.675
                                                                              1993
                                                                                       Male 2267.493
                                                                              1994
                                                                                       Male 2889.264
                                                                             1995
                                                                                       Male 3355.927
                    Female
            Male
   Year
  1990 3236.524 3256.140
                                                                              1996
                                                                                       Male 2855.110
                              melt(df, id.vars='Year',variable.name = 'series')
   1991 2714.489 3012.146
                                                                             1997
                                                                                       Male 3218.254
   1992 2997.675 3232.426
                                                                          9
                                                                             1998
                                                                                       Male 2515.704
                                                                          10 1999
   1993 2267.493 3066.795
                                                                                       Male 2937.085
                                                                          11 1990
                                                                                     Female 3256.140
  1994 2889.264 2733.549
  1995 3355.927 3205.007
                                                                          12 1991
                                                                                     Female 3012.146
                                dcast(df, Year ~ variable, value.var='value')
   1996 2855.110 2927.405
                                                                          13 1992
                                                                                     Female 3232.426
                                                                          14 1993
                                                                                     Female 3066.795
   1997 3218.254 2358.021
                                                                          15 1994
                                                                                     Female 2733.549
   1998 2515.704 2699.584
10 1999 2937.085 3049.776
                                                                          16 1995
                                                                                     Female 3205.007
                                                                          17 1996
                                                                                     Female 2927.405
                                                                          18 1997
                                                                                     Female 2358.021
                                                                          19 1998
                                                                                     Female 2699.584
                                                                          20 1999
                                                                                     Female 3049.776
```

Year variable

1990

value

Male 3236.524

# Visualizing Spatial Data

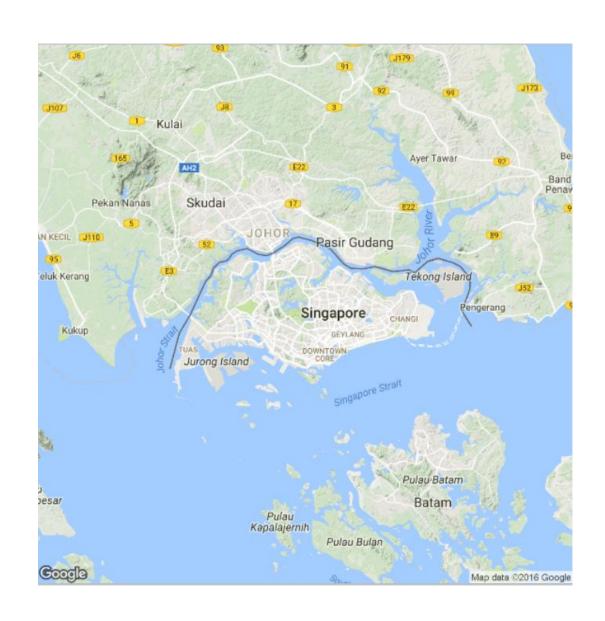
with ggmap

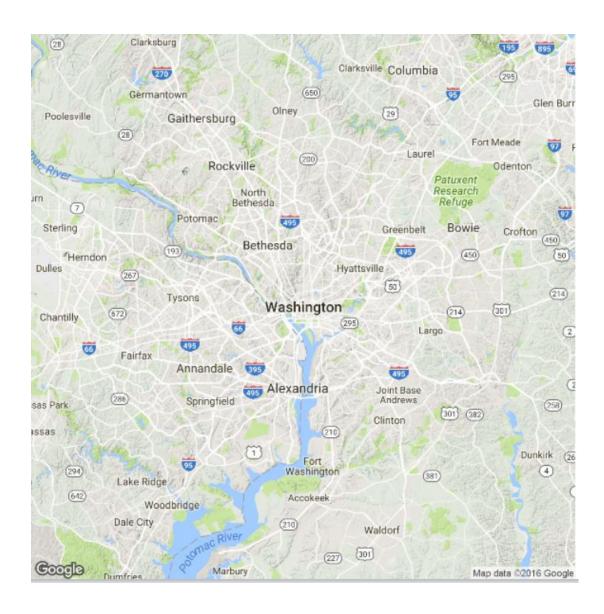
#### Introduction to ggmap

- "ggmap" is a package developed on top of *ggplot2* for visualizing spatial data.
- ggmap plots have all elements of ggplot2, but certain elements are fixed to map components.
  - The x aesthetic is fixed to longitude, the y aesthetic is fixed to latitude, and the coordinate system is fixed to the Mercator projection
- Other ggplot2 plots can overlay with ggmap plot.

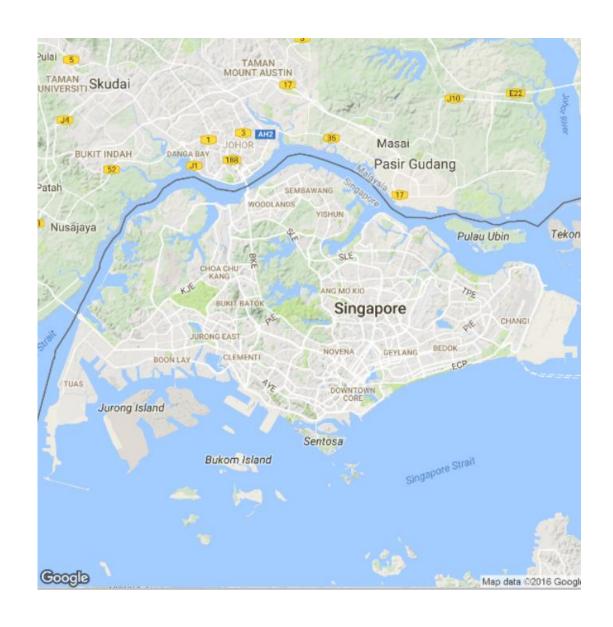
#### qmap(location="Singapore")

#### qmap(location="Washington")





#### qmap(location="Singapore", zoom=11)



#### maptype

- satellite
- hybrid
- toner
- watercolor
- terrain-background
- toner-lite

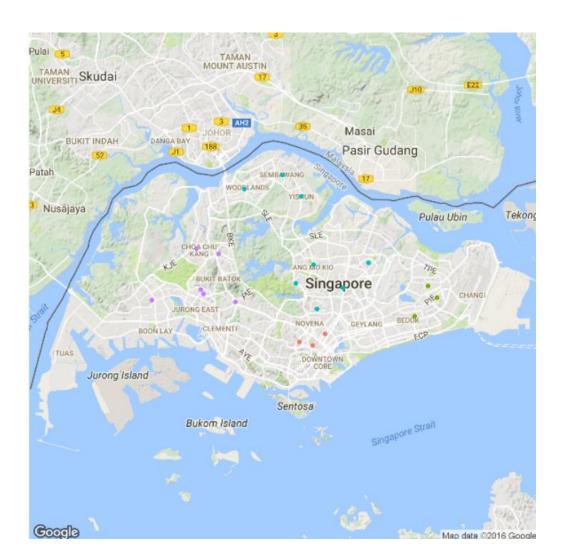


#### Now let's make it interesting ...

```
pizzahut.location <- read.csv("PizzaHut.csv",header = TRUE, colClasses = c("character","character","factor","character","numeric","numeric",))

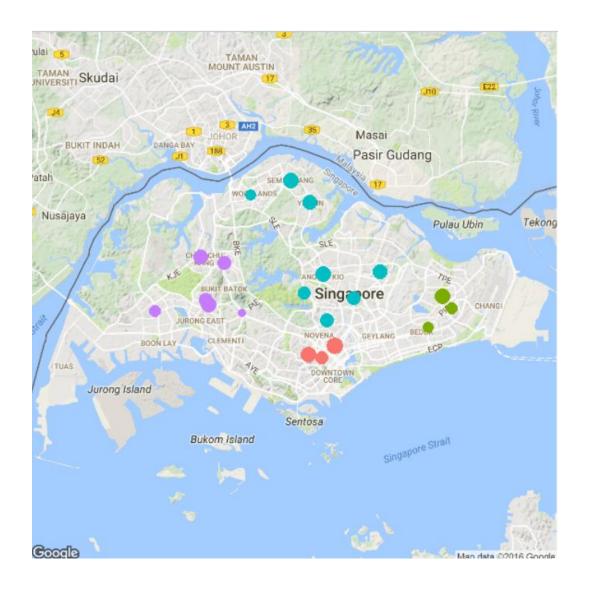
m1 <- qmap("Singapore", base_layer=ggplot(aes(x=lon, y = lat), data=pizzahut.location), zoom=11, scale=2)

m2 <-m1 + geom_point(aes(color=Region))</pre>
```



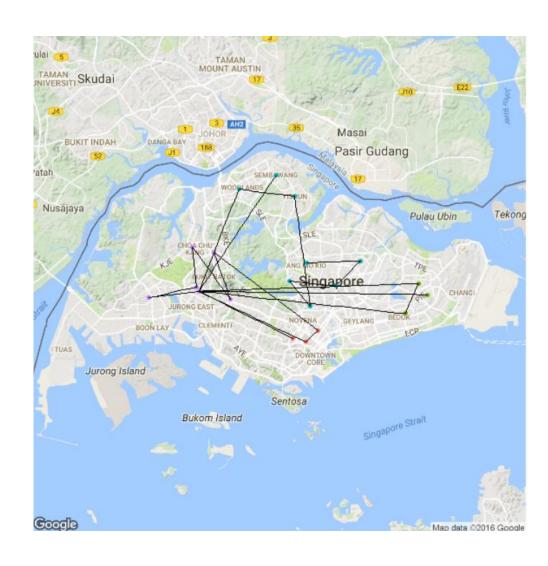
#### It is just ggplot2 ...

```
pizzahut.location$Visits = round(rnorm(nrow(pizzahut.location),15000,5000))
m1 <- qmap("Singapore", base_layer=ggplot(aes(x=lon, y = lat), data=pizzahut.location), zoom=11, scale=2)
m3 <- m1 + geom_point(aes(color=Region, size=Visits))</pre>
```



## More layer

m4 <- m1 + geom\_point(aes(color=Region)) + geom\_path()</pre>



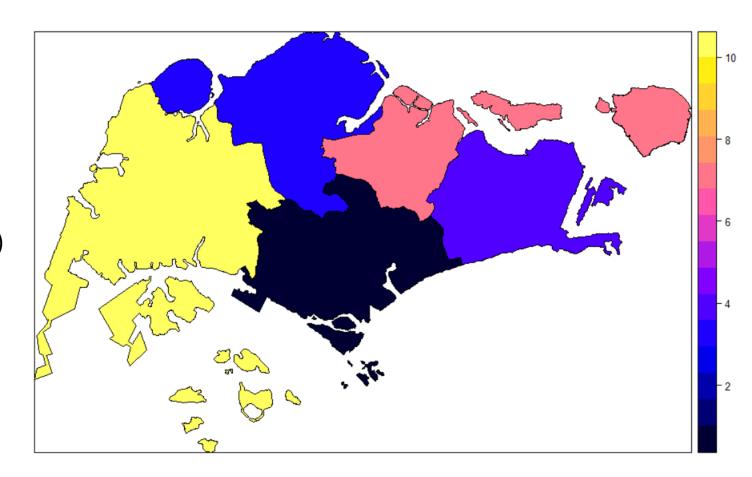
## Map view by borders

```
install.packages("raster")
```

install.packages("rgdal")

library(raster)
library(rgdal)
library(XML)

SG<-getData('GADM', country='SG', level=1) SG\$value<-c(1,4,7,3,10) spplot(SG,"value")



#### Interactive Plot with plotly

```
install.packages("plotly")
library(plotly)
set.seed(100)
d <- diamonds[sample(nrow(diamonds), 1000), ]</pre>
```

```
p <- ggplot(data = d, aes(x = carat, y = price)) +
  geom_point(aes(text = paste("Clarity:", clarity)), size = 1) +
  geom_smooth(aes(colour = cut, fill = cut)) + facet_wrap(~ cut)
(gg <- ggplotly(p))</pre>
```

