

UCLA Extension Track 1  
UCLA Department of Statistics  
Statistical Consulting Center

---

## R Graphics II: Graphics for Exploratory Data Analysis

Irina Kukuyeva  
ikukuyeva@stat.ucla.edu

---

January 12, 2011



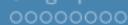
# Outline

- 1 Summary Plots
- 2 Time Series Plots
- 3 Geographical Plots
- 4 3D Plots
- 5 Simulation Plots
- 6 Useful Links for R

## Disclaimer

The author assumes you are familiar with R and are comfortable subsetting.





## 1 Summary Plots

- Basic Plots
- Looking at Distributions
- Identifying Observations
- Exercise I

## 2 Time Series Plots

## 3 Geographical Plots

## 4 3D Plots

## 5 Simulation Plots

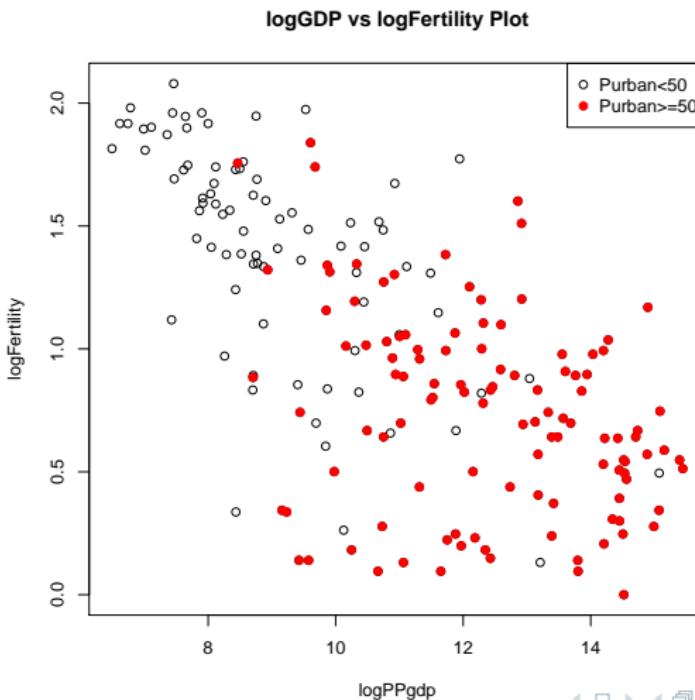
## 6 Useful Links for R

## Basic Plot I

```
1 # Step 1: Load the Data
2 library(alr3)
3 data(UN2)
4 attach(UN2)
5 # Step 2: Subset appropriately
6 ind<-which(Purban>50)
7 # Step 3: Plot
8 plot(logFertility~logPPgdp, xlab="logPPgdp",
       ylab="logFertility", main="logGDP vs
           logFertility Plot")
9 points(logFertility[ind]~logPPgdp[ind], col=
       "red", pch=19)
10 legend("topright", pch=c(1,19), col=1:2, c(
        "Purban<50", "Purban>=50"))
```



# Basic Plot II



## Segmented bar charts I

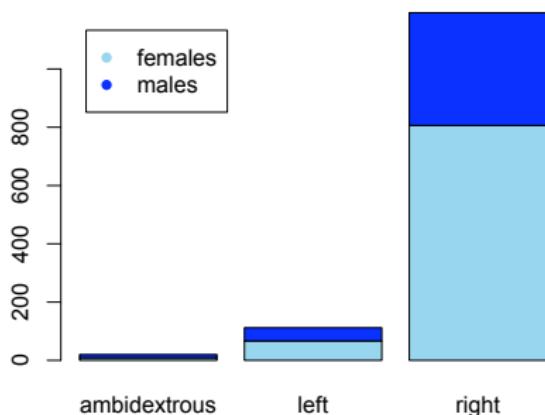
Displays two categorical variables at a time: <sup>1</sup>

```
1 survey = read.table("http://www.stat.ucla.edu/~mine/students_survey_2008.txt", header = TRUE, sep = "\t")
2 attach(survey)
3 barplot(table(gender, hand), col = c("skyblue", "blue"), main = "Segmented Bar Plot \n of Gender")
4 legend("topleft", c("females", "males"), col = c("skyblue", "blue"), pch = 16, inset = 0.05)
```



## Segmented bar charts II

## **Segmented Bar Plot of Gender**



	ambidextrous	left	right
female	9	67	806
male	11	45	387

<sup>1</sup>These two slides are modified from the SCC Mini-Course "Introductory Consulting Statistics with R" by Mine Çetinkaya

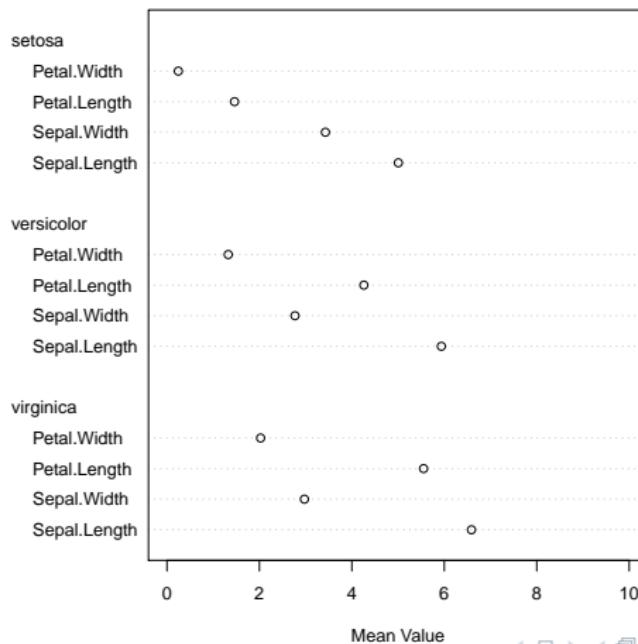
# Dot charts I

To compare values for variables in each category:

```
1 # Step 1: Load the data
2 data(iris)
3 attach(iris)
4 # Step 2: Calculate means for each species:
5 aggregate(iris[, -5], list(Species = Species),
   mean)->a
6 # Step 3: Assign row names
7 row.names(a)<-a[, 1]
8 # Step 4: Plot
9 dotchart(t(a[, -1]), xlim = c(0,10), main =
  "Plots of Means for Iris Data Set", xlab =
  "Mean Value")
```

## Dot charts II

## Plots of Means for Iris Data Set

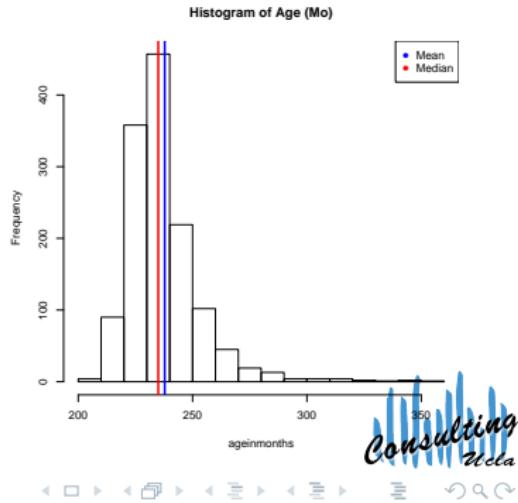


## Histograms

## Adding Summary Statistics to Plots

Add the mean and median to a histogram:

```
1 hist(ageinmonths, main =  
      "Histogram of Age (Mo)")  
2 abline(v = mean(ageinmonths),  
          col = "blue", lwd = 3)  
3 abline(v = median(ageinmonths),  
          col = "red", lwd = 3)  
4 legend("topright", c("Mean",  
          "Median"), pch = 16,  
          col = c("blue", "red"))
```



## Histograms I

## Checking Normality

One of the methods to test for normality of a variable is to look at the histogram (the sample density is in red, the theoretical normal density in blue):

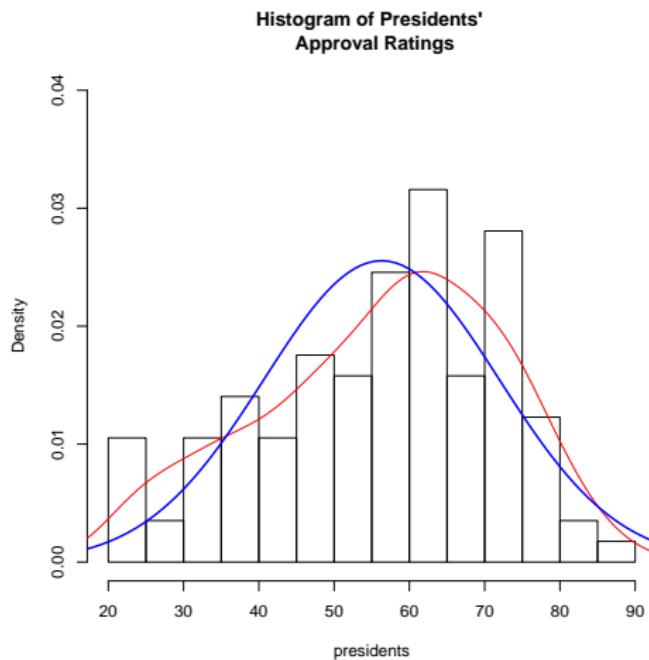
```
1 data(presidents)
2 hist(presidents, prob=TRUE, ylim=c(0, 0.04),
      breaks=20)
3 lines(density(presidents, na.rm=TRUE), col="red")
4 mu<-mean(presidents, na.rm=TRUE)
5 sigma<-sd(presidents, na.rm=TRUE)
6 x<-seq(10,100, length=100)
7 y<-dnorm(x,mu,sigma)
8 lines(x,y,lwd=2, col="blue")
```



Looking at Distributions

# Histograms II

## Checking Normality



## Box and Whisker Plot II

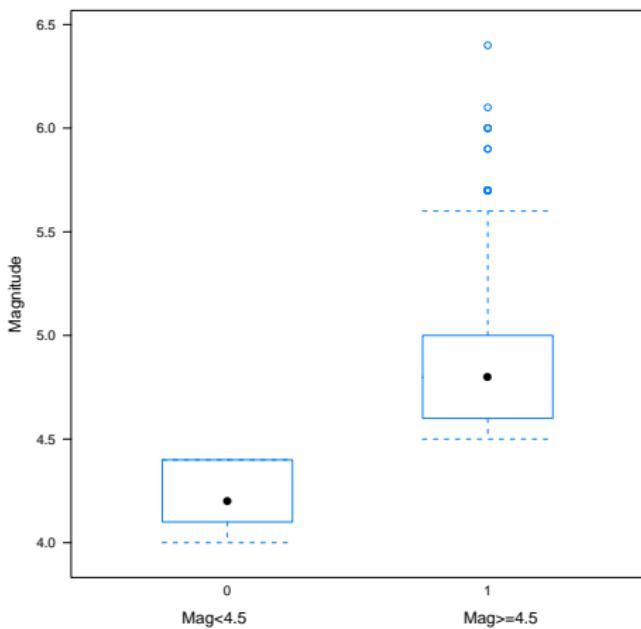
Another method of looking at the distribution of the data is via boxplot:

```
1 data(quakes)
2 # Subset the magnitude:
3 ind<-ifelse(quakes$mag<4.5, 0, 1)
4 ind<-as.factor(ind)
5 library(lattice)
6 bwplot(quakes$mag~ind, xlab=c("Mag<4.5", "Mag
    >=4.5"), ylab="Magnitude")
```



# Box and Whisker Plot II

Fiji EQ since 1964



## Beanplot I

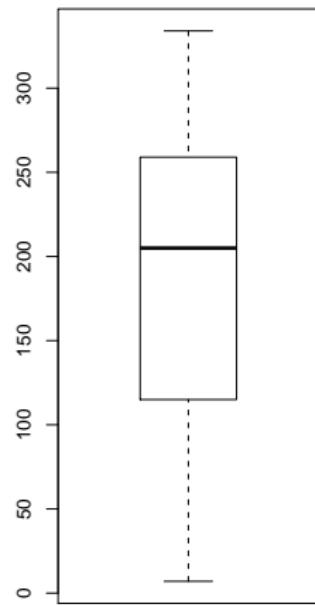
An alternative to the boxplot is the beanplot():

```
1 library(beanplot)
2 par(mfrow=c(1,2))
3 data(airquality)
4 boxplot(airquality[, 2], main="Boxplot", xlab=
+         "Solar")
5 beanplot(airquality[, 2], main="Beanplot",
+          xlab="Solar")
```

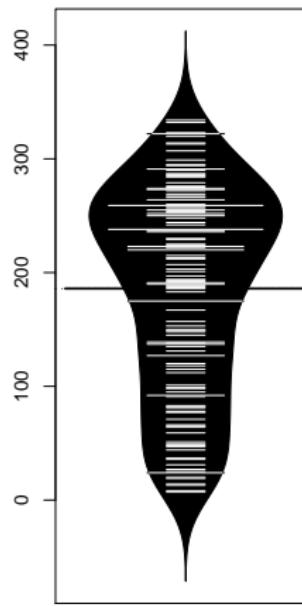


Beanplot II

## Boxplot



## Beanplot



## Scatterplots I

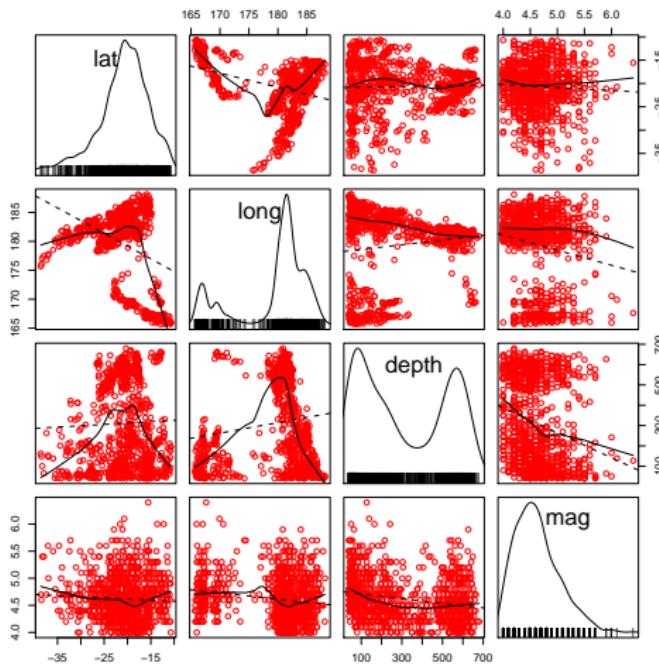
A method of looking at the distribution and correlation of the data is via `scatterplot.matrix()`:

```
1 data(quakes)
2 library(car)
3 scatterplot.matrix(quakes[, 1:4])
```



Looking at Distributions

# Scatterplots II



# Checking Equality of Distributions I

## Approach 1

A method of checking equality of distributions is via `qq()`:

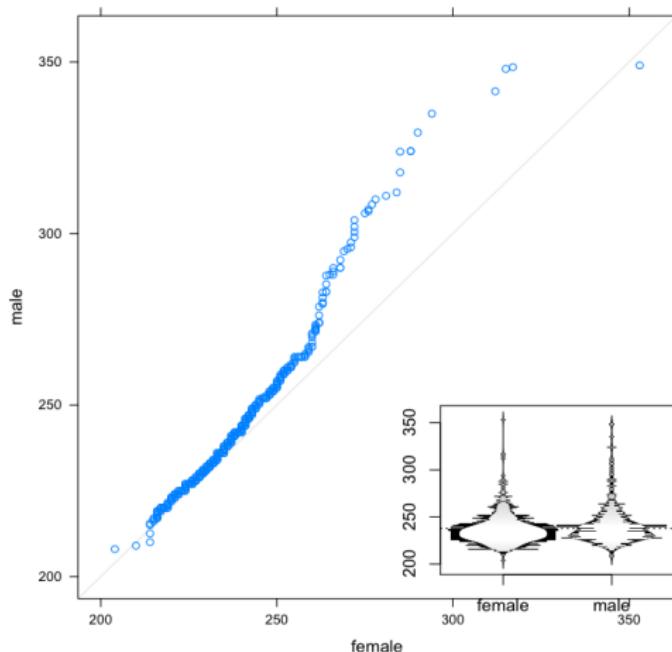
```
1 library(lattice)
2 survey = read.table("http://www.stat.ucla.edu
3   /~mine/students_survey_2008.txt", header =
4     TRUE, sep = "\t")
5 attach(survey)
6 qq(gender~ageinmonths)
```



Looking at Distributions

# Checking Equality of Distributions II

Approach 1



# Checking Equality of Distributions I

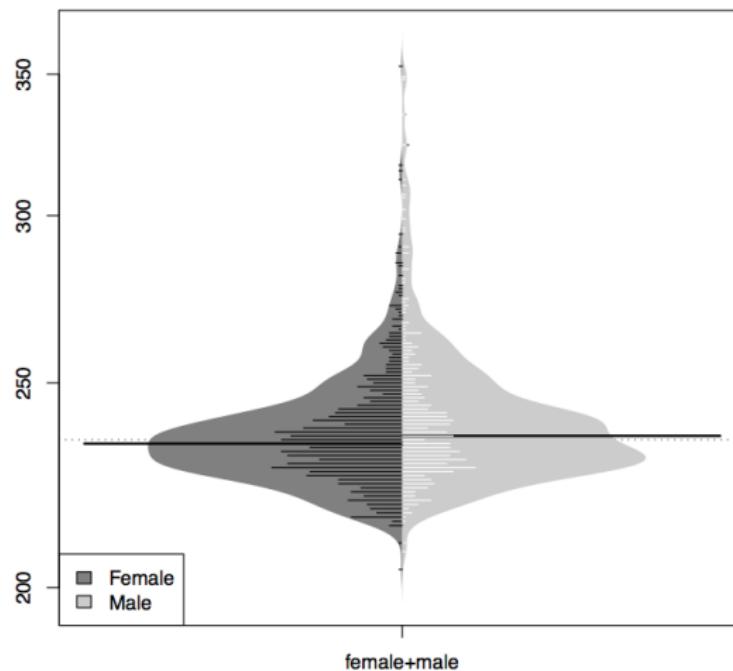
## Approach 2

Another method for checking equality of distributions is via `beanplot()`:

```
1 survey = read.table("http://www.stat.ucla.edu  
/~mine/students_survey_2008.txt", header =  
TRUE, sep = "\t")  
2 attach(survey)  
3 beanplot(ageinmonths ~ gender, data=survey,  
col=list(grey(0.5),c(grey(0.8),"white")),  
border = NA, overallline = "median", ll  
=0.01, side="both")  
4 legend("bottomleft",fill=c(grey(0.5),grey(0.8))  
, legend=c("Female","Male"))
```

# Checking Equality of Distributions II

## Approach 2



## Identifying Observations I

## Preliminaries

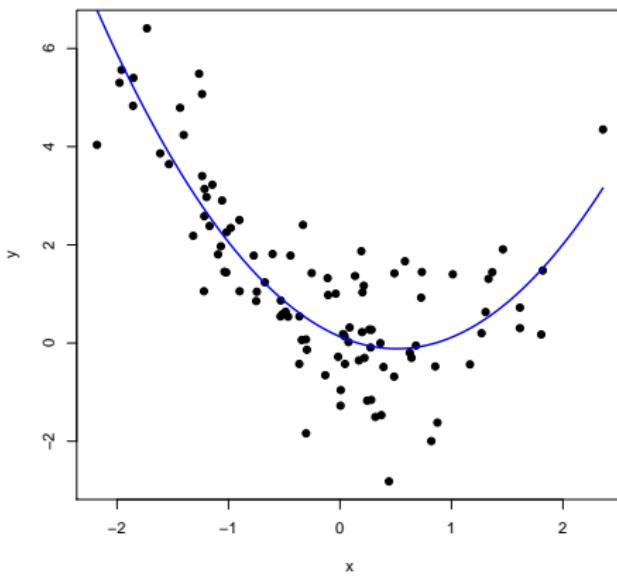
```
1 # Generate data and fit a regression curve:  
2 set.seed(3012008)  
3 x=rnorm(100); y=-x+I(x^2) +rnorm(100)  
4 fit<-lm(y~x+I(x^2)); fit
```

Intercept	x	$x^2$
0.1307	-0.9701	0.9549

```
1 # Plot the resulting regression curve:  
2 plot(y~x, pch=19)  
3 curve(expr=fit[[1]][1]+fit[[1]][2]*x+fit  
       [[1]][3]*I(x^2), from=range(x)[1], to=  
       range(x)[2], add=TRUE, col="blue", lw
```

# Identify-ing Observations II

## Preliminaries



## Identify-ing Observations

# Identify-ing Observations I

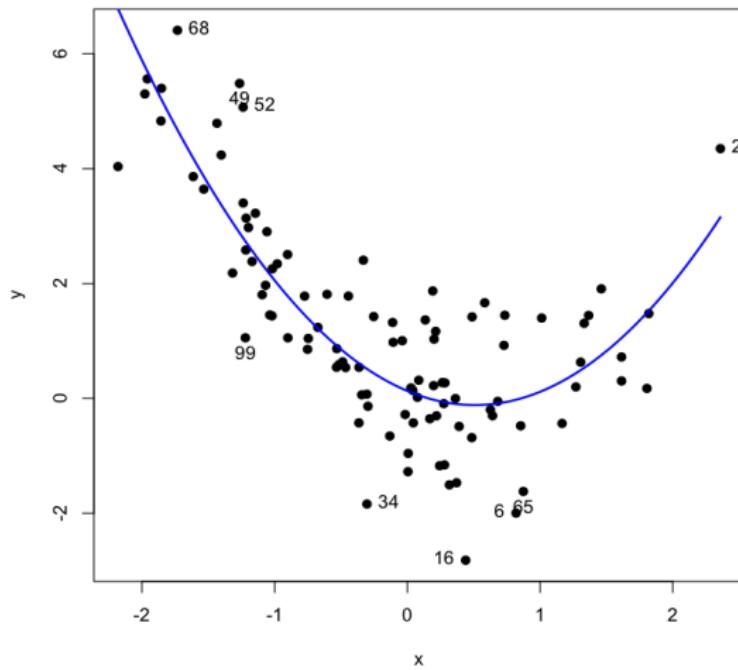
**Left-click** on the observations in the graphics window to see the row number.

**Right-click** on the observation to exit the function.

```
1 # Plot the data and fit the regression curve:  
2 plot(y~x, pch=19)  
3 curve(expr=fit[[1]][1]+fit[[1]][2]*x+fit  
        [[1]][3]*I(x^2), from=range(x)[1], to=  
        range(x)[2], add=TRUE, col="blue", lwd=2)  
4 # Identify the "outlying" observations:  
5 index<-identify(y~x); index
```

Identify-ing Observations

# Identify-ing Observations II



## Exercise I

Compare the distributions of the sepal widths for the three species of irises (using the `iris` data set).



## 1 Summary Plots

## 2 Time Series Plots

- Multivariate Plots
- Exercise II

## 3 Geographical Plots

## 4 3D Plots

## 5 Simulation Plots

## 6 Useful Links for R

# Multivariate Time Series Plots I

## Approach 1

To plot more than variables one at a time, use `plot()`:

```
1 # Convert data to a time series via ts() or
  # zoo():
2 data(airquality)
3 a<-airquality[, 1:3]
4 time<-ts(1:nrow(a), start=c(1973, 5),
  frequency=365)
5 # If your data is stored as a data frame,
6 # coerce it to be a matrix via as.matrix()
7 class(a)
8 a.mat<-as.matrix(a)
9
10
```

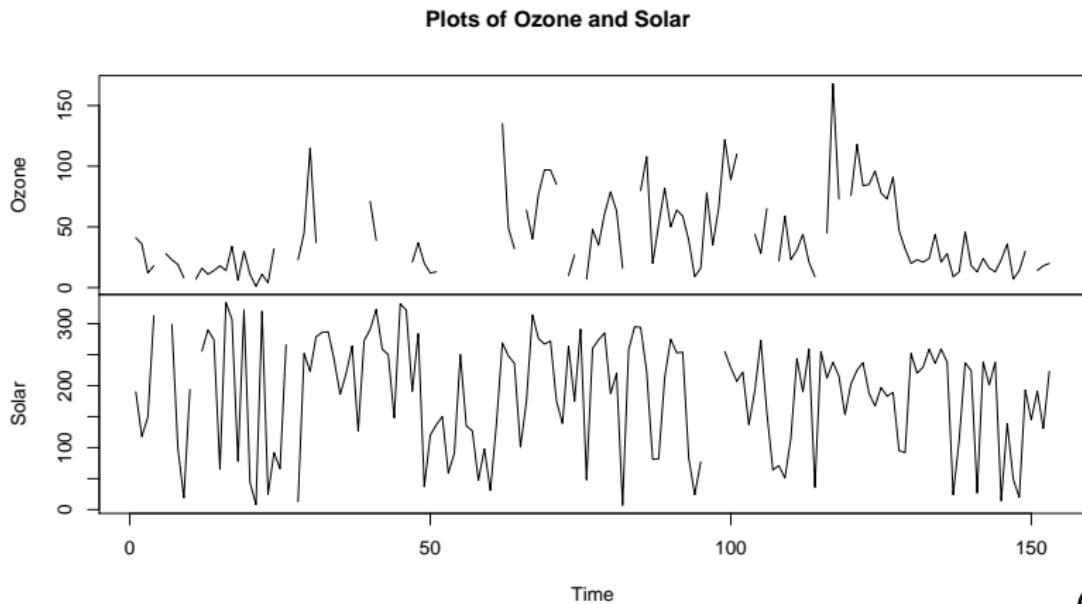
# Multivariate Time Series Plots II

## Approach 1

```
11 # Make a time series of the two (or more  
#      variables)  
12 library(zoo)  
13 name.zoo<-zoo(cbind(a.mat[, 1], a.mat[, 2]))  
14 colnames(name.zoo)<-c("Ozone", "Solar")  
15 ##### Plot the variables  
16 plot(name.zoo)
```

# Multivariate Time Series Plots III

## Approach 1



# Multivariate Time Series Plots

## Approach 2

To plot more than variables one at a time, use `mvtspplot()`:<sup>2</sup>

```
1 # Load the R Code for the function
2 source("http://www.biostat.jhsph.edu/~rpeng/RR
      /mvtspplot/mvtspplot.R")
3 # After processing data as in Approach 1
4 # Plot the variables
5 mvtspplot(name.zoo)
6 # Purple=low, grey=medium, green=high, white=
  missing values
```

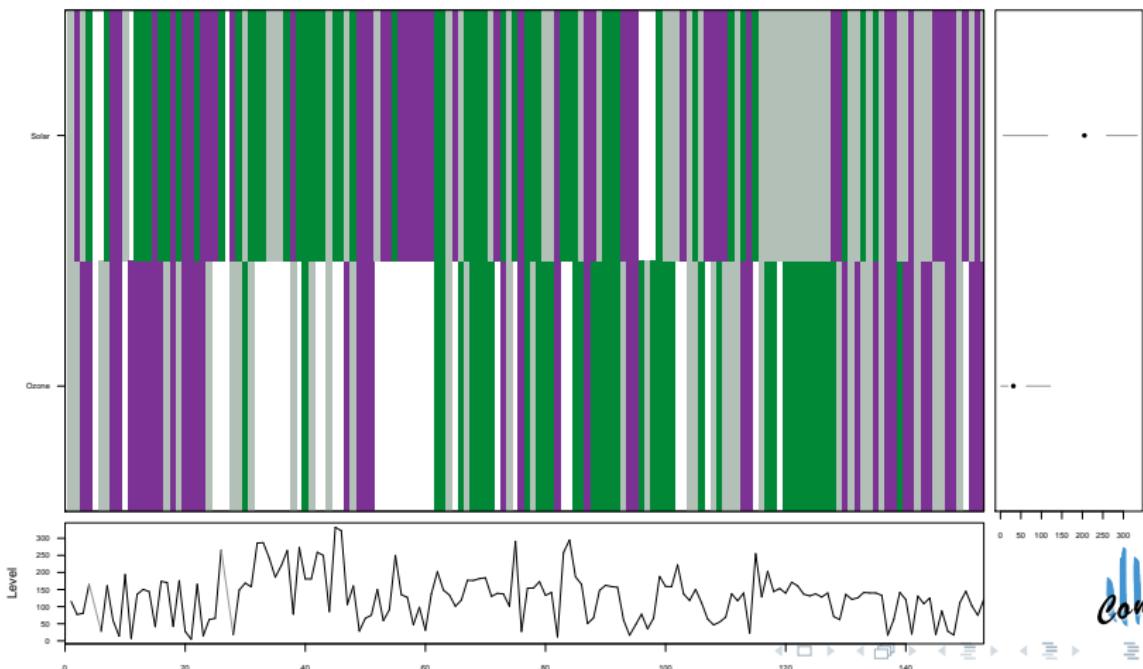


---

<sup>2</sup>For documentation, go to: [www.jstatsoft.org/v25/c01/paper](http://www.jstatsoft.org/v25/c01/paper)

# Multivariate Time Series Plots

## Approach 2



# Multivariate Time Series Plots II

## Approach 3

To plot more than variables one at a time, use `xyplot()`:

```

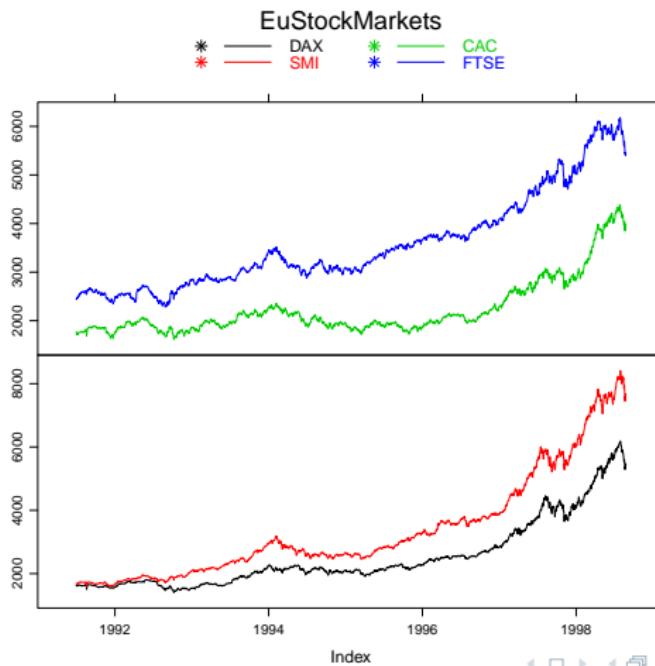
1 # After processing data as in Approach 1
2 # load both libraries:
3 library(lattice)
4 library(zoo)
5 data(EuStockMarkets)
6 z<-EuStockMarkets
7 xyplot(z, screen = c(1,1,2,2), col = 1:4,
       strip = FALSE, key = list(title =
         "EuStockMarkets", columns=2, points=FALSE,
         lines=TRUE, col=1:4, text=list(colnames(z))
       )))

```



# Multivariate Time Series Plots II

## Approach 3



## Exercise II

Analyze the `EuStockMarkets` data using the function `mtvspplot()`. What stands out and why?

## 1 Summary Plots

## 2 Time Series Plots

## 3 Geographical Plots

- Maps
- Projection Maps
- Exercise III

## 4 3D Plots

## 5 Simulation Plots

## 6 Useful Links for R

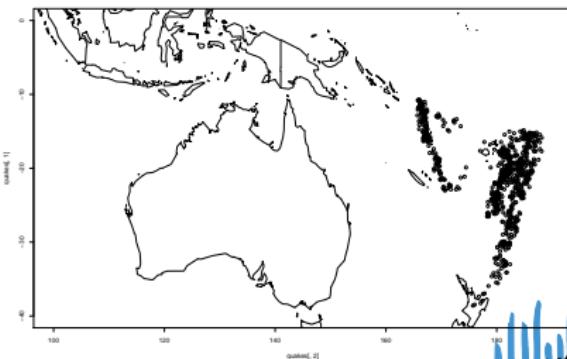


# Geographic Maps

## Map of Fiji Earthquakes Since 1964

To overlay a map to a plot containing latitude and longitude, load the package `maps`:

```
1 data(quakes)
2 library(maps)
3 plot(quakes$long,
      quakes$lat, xlim=
      c(100, 190), ylim=
      c(-40, 0))
4 map("world", add=TRUE
      )
```



# Geographic Maps I

## Map of Fiji Earthquakes Since 1964

To include information on magnitude of earthquake, use cex argument of the plot() function:

```
1 # Step 1: Recode magnitude to have 3
  categories only
2 ind<-which(quakes$mag<5)
3 ind2<-which(quakes$mag<6 & quakes$mag >=5)
4 ind3<-which(quakes$mag>=6)
5 color<-rep(NA, length(quakes$mag))
6 color[ind]<-1; color[ind2]<-2; color[ind3]<-3
7
8
9
10
```

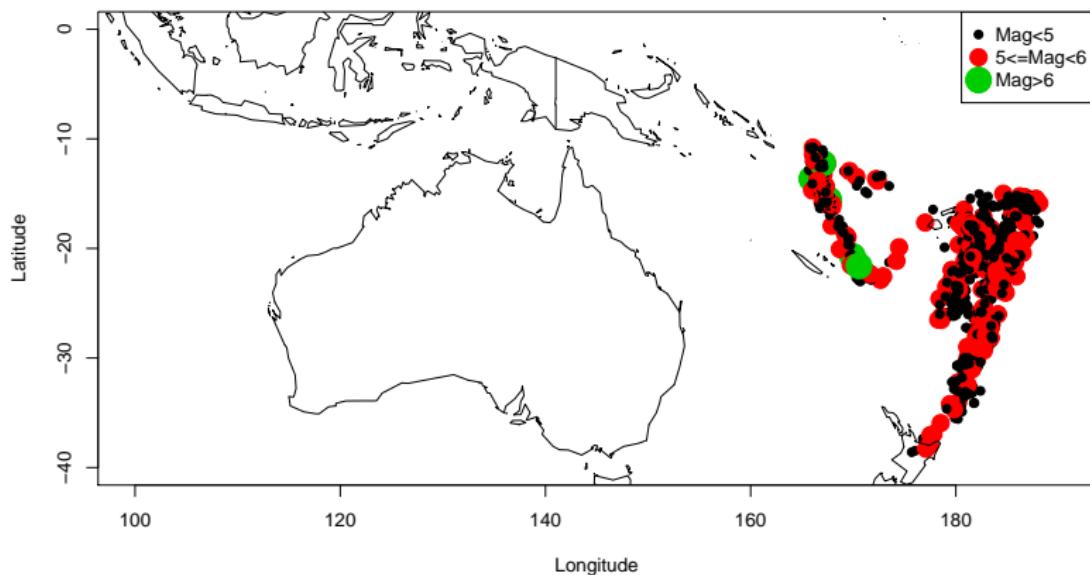
# Geographic Maps II

## Map of Fiji Earthquakes Since 1964

```
11  # Step 2: Plot
12  plot(quakes$long, quakes$lat, cex=color, pch
13    =19, col=color, xlim=c(100, 190), ylim=
14      c(-40,0), xlab="Longitude", ylab="Latitude
15    ")
16  legend("topright", pch=19, col=1:3, c("Mag<5",
17    "5<=Mag<6", "Mag>6"), pt.cex=1:3)
18  map("world", add=TRUE)
```

# Geographic Maps III

## Map of Fiji Earthquakes Since 1964



# Projection Maps I

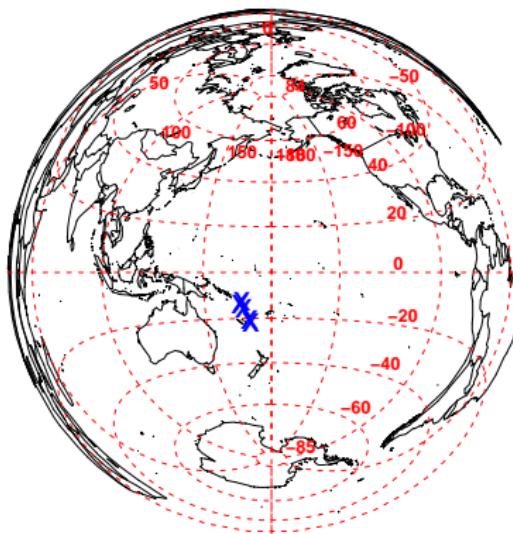
Map of Fiji Earthquakes Since 1964

For a different perspective of a map, use `mapproject()`:

```
1 library(mapproj)
2 library(maps)
3 m <- map('world',plot=FALSE)
4 # Projection is Azimuthal with equal-area
5 map('world',proj='azequalarea',orient=c(
       longitude=0,latitude=180,rotation=0))
6 map.grid(m,col=2)
7 points(mapproject(list(
      y=quakes$lat[which(quakes$mag>=6)],
      x=quakes$long[which(quakes$mag>=6)])),
      col="blue",pch="x",cex=2)
```

# Projection Maps II

Map of Fiji Earthquakes Since 1964



## Bonus Feature of the `maps` package:

To determine in which part of the world the observations are (based on latitude and longitude), use `map.where()`:

```
1 in.what.country <- map.where(database="world",
      quakes$long, quakes$lat)
```

To determine which observations are in the ocean:

```
1 # Number of points in ocean after filtering:
2 ind <- sum(is.na(in.what.country)); ind
3 # Number of observations: 1000
4 # Number in Ocean: 993
```

## Exercise III

# Exercise III

For the ozone data set <sup>3</sup>, determine the region of the world that the observations correspond to and overlay the appropriate map.

---

<sup>3</sup><http://www.ats.ucla.edu/stat/R/faq/ozone.csv>



## 1 Summary Plots

## 2 Time Series Plots

## 3 Geographical Plots

## 4 3D Plots

- lattice library
- ggplot2 library
- rgl library
- Saving Plots as a PDF
- Exercise IV

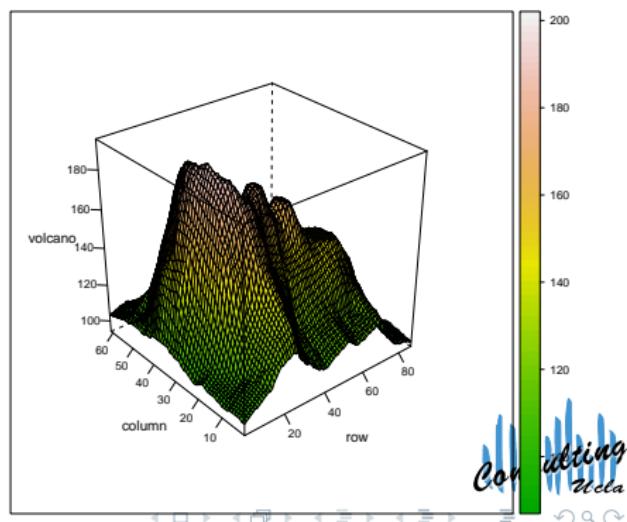
## 5 Simulation Plots

## 6 Useful Links for R

# 3D Images with Package lattice

**Method 1:** Using `wireframe()`:

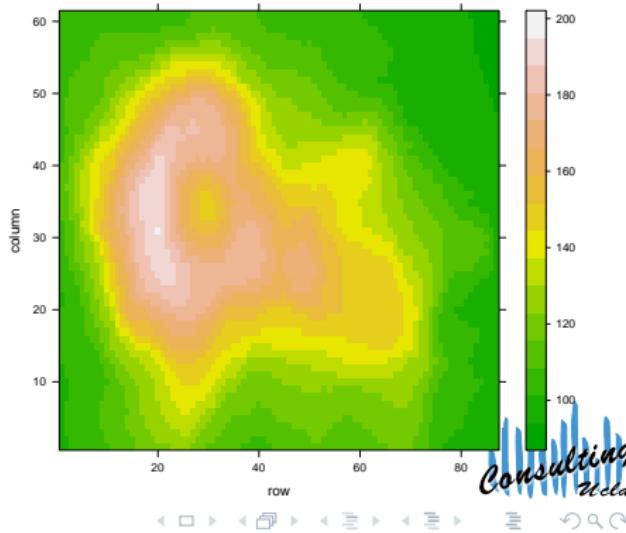
```
1 library(lattice)
2 wireframe(volcano ,
  col.regions =
terrain .
colors(100), asp =
  1, color.key=TRUE
, drape=TRUE,
scales = list(
arrows = FALSE))
```



# 3D Images with Package lattice

**Method 2:** Same image with the `levelplot()` function:

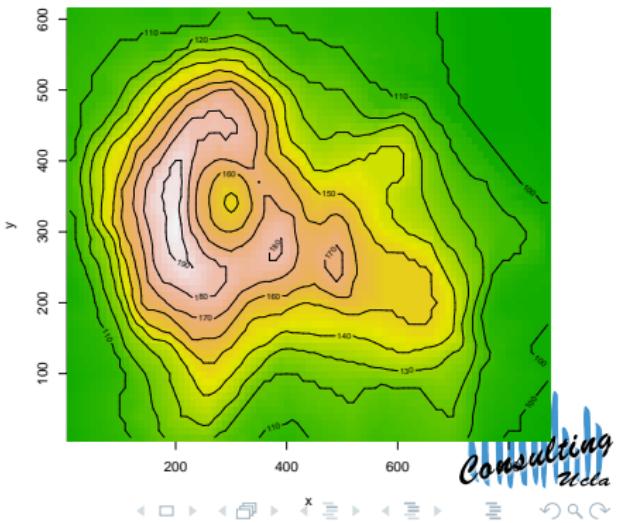
```
1 library(lattice)  
2 levelplot(volcano,  
           asp=1, col.regions  
           =terrain.colors)
```



# 3D Images with Package lattice

**Method 3:** Same image with the `image()` function:

```
1 x<-10*(1:nrow(volcano))
   )
2 y<-10*(1:ncol(volcano))
   )
3 image(x, y, volcano,
       col = terrain.
       colors(100), axes =
       TRUE)
4 contour(x, y, volcano,
       add = TRUE, col =
       1)
```



# 3D Images with Package ggplot2 |

Another way to create 3D images is with the package `ggplot2`:

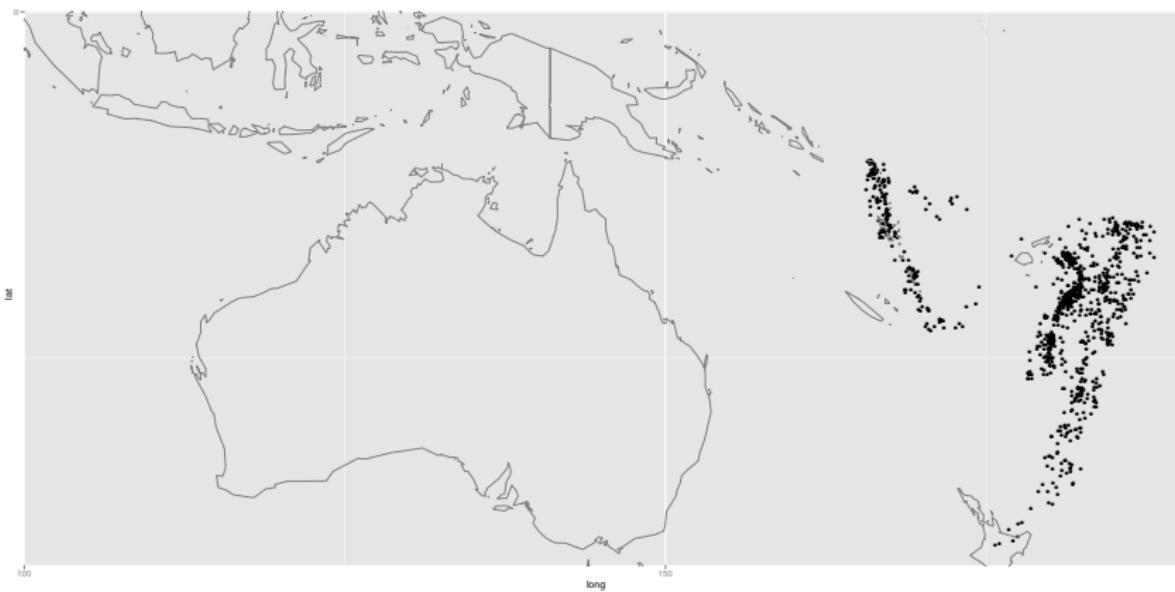
```
1 # Step 1: Load the data
2 data(quakes)
3 attach(quakes)
4 # Step 2: Create a categorical variable for
      earthquake magnitude
5 ind<-which(mag<5)
6 ind2<-which(mag<6 & mag >=5)
7 ind3<-which(mag >=6)
8 color<-rep(NA, length(mag))
9 color[ind]<-1; color[ind2]<-2; color[ind3]<-3
10 color<-as.factor(color)
11
12
```

# 3D Images with Package ggplot2 II

```
13 # Step 3: Plot
14 library(ggplot2)
15 ggplot(data=quakes, aes(long, lat))+
16 geom_point(aes(long, lat))+
17 borders("world")+
18 coord_cartesian(xlim=c(100,190), ylim=c(-40,0)
)
```

ggplot2 library

# 3D Images with Package ggplot2 III



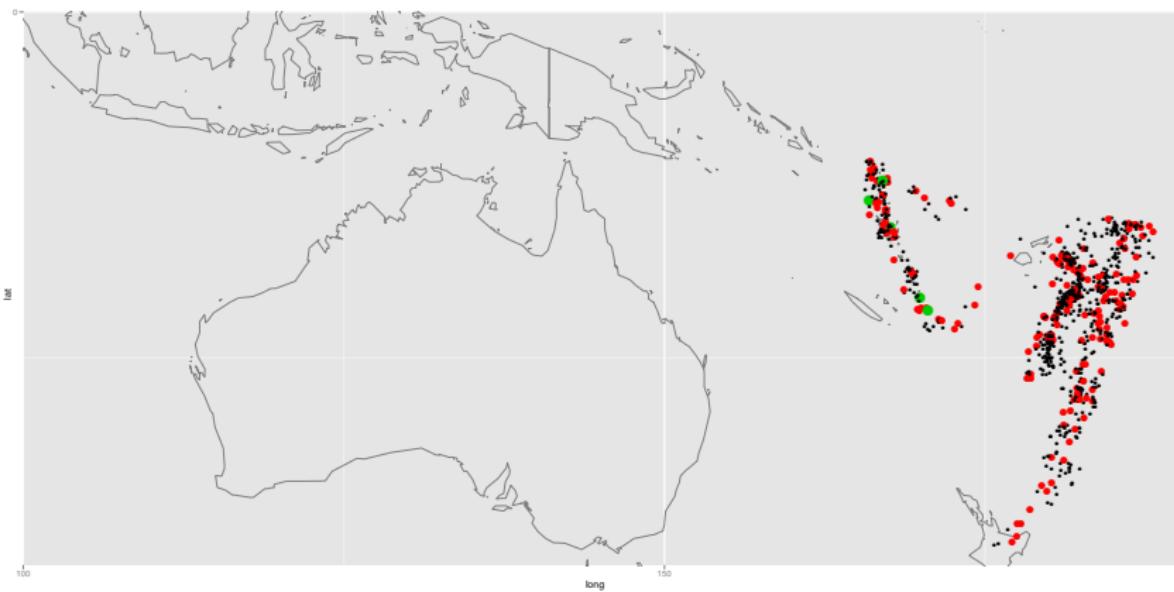
# 3D Images with Package ggplot2 |

To include information on magnitude of earthquake, use  
`geom_point()`:

```
1 ggplot(data=quakes, aes(long, lat))+
2 borders("world")+
3 coord_cartesian(xlim=c(100,190), ylim=c(-40,0))
4 +
4 geom_point(data=quakes, colour=as.numeric(color),
           size=2*uas.numeric(color))
```

ggplot2 library

# 3D Images with Package ggplot2 II

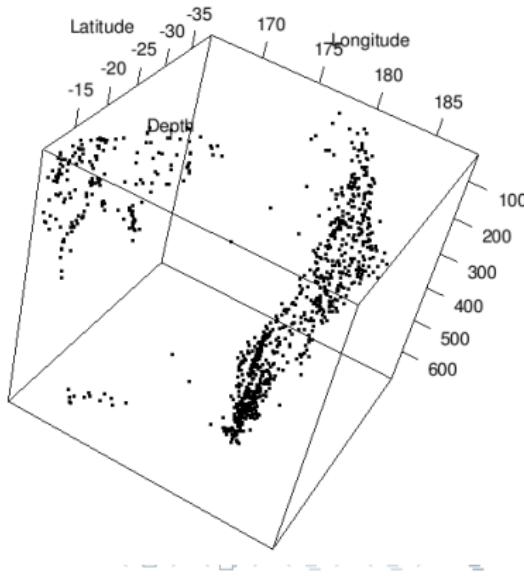
Consulting  
ucla

```
rgl library
```

# 3D Images with Package rgl

A way to create 3D interactive images is with the package `rgl`:

```
1 library(rgl)
2 data(quakes)
3 plot3d(x=quakes$long,
          y=quakes$lat, z=
          quakes$depth, xlab
          ="Longitude", ylab
          ="Latitude", zlab=
          "Depth")
```



# Saving Plots as a PDF

## For Static Plots

*Note:* The files will be saved in the folder specified with `setwd()`.  
To save a static plot in R as a PDF, use function `pdf()`:

```
1 # To save the image to the desktop:  
2 setwd("~/Desktop")  
3 wireframe(volcano, col.regions = terrain.  
           colors(100), asp = 1, color.key=TRUE,  
           drape=TRUE, scales = list(arrows = FALSE))  
4 dev.off()
```

# Saving Plots as a PDF

## For Dynamic Plots

*Note:* The files will be saved in the folder specified with `setwd()`.

To save a dynamic plot in R as a PDF, use function

`rgl.snapshot()`:

```
1 # To save the image to the desktop:  
2 setwd("~/Desktop")  
3 # Step 1: Produce the 3D image:  
4 plot3d(x=quakes$long, y=quakes$lat, z=quakes  
         $depth, xlab="Longitude", ylab="Latitude",  
         zlab="Depth")  
5 # Step 2: Can rotate before taking a snapshot:  
6 rgl.snapshot("quakes.png")
```

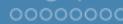
# Exercise IV

Use the `rgl` library to create a 3D plot of the `ozone` data set <sup>4</sup>.



---

<sup>4</sup><http://www.ats.ucla.edu/stat/R/faq/ozone.csv>



## 1 Summary Plots

## 2 Time Series Plots

## 3 Geographical Plots

## 4 3D Plots

## 5 Simulation Plots

- Preliminaries
- Performing the Simulation and Visualizing the Results
- Exercise V: Homework Problem

## 6 Useful Links for R



# Simulations

Preliminaries: The function `outer()`

```
1 x=5:6; y=1:3  
2 outer(x,y)
```

	[,1]	[,2]	[,3]
[1,]	5	10	15
[2,]	6	12	18

```
1 fcn<-function(x,y){z=x+y}  
2 outer(x,y,fcn)
```

	[,1]	[,2]	[,3]
[1,]	6	7	8
[2,]	7	8	9

# Simulations I

## Visualize with `persp()`

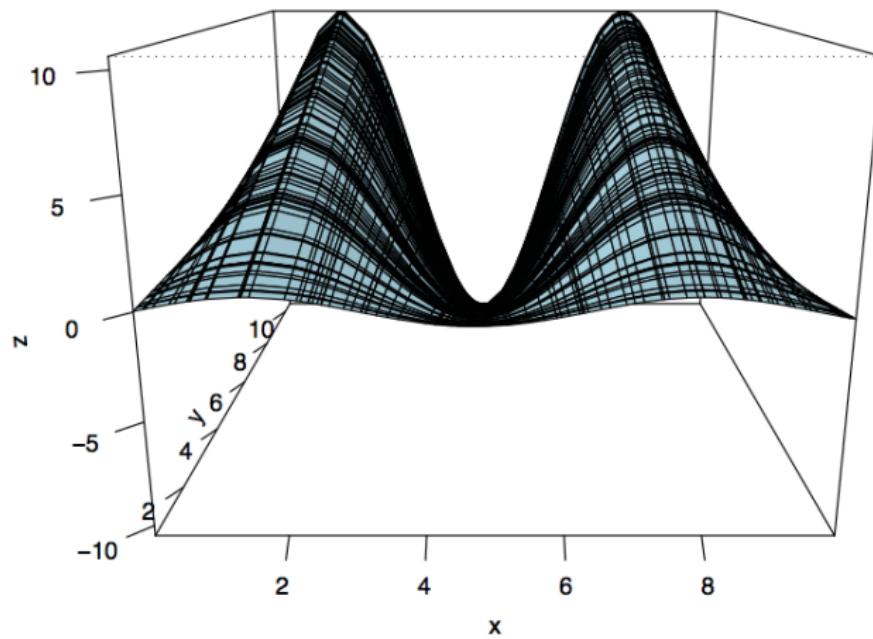
Suppose we want to know what the function  $y \times \sin(x)$  looks like:

```
1 # Sample from the random uniform:
2 x <- sort(runif(100, min=0, max=10))
3 y <- x+runif(1)
4 f <- function(x,y) { r <- y*sin(x)}
5 z <- outer(x,y,f)
6 persp(x, y, z, col = "lightblue", shade = 0.1,
       ticktype = "detailed", expand=0.7)
```



# Simulations II

Visualize with `persp()`



# Simulations I

## Visualize with `image()`

To visually see its maximum and minimum values, look at the contours of the function:

```
1 # Format the data appropriate for the image()
  function:
2 x.seq<-seq(from=range(x)[1], to=range(x)[2],
  length=100)
3 y.seq<-seq(from=range(y)[1], to=range(y)[2],
  length=100)
4 n1<-length(x.seq)
5 n2<-length(y.seq)
6 mat1<-matrix(z, nrow=n1, ncol=n2, byrow=FALSE)
7 image(x.seq, y.seq, mat1)
8 contour(x.seq, y.seq, mat1, add=TRUE)
```

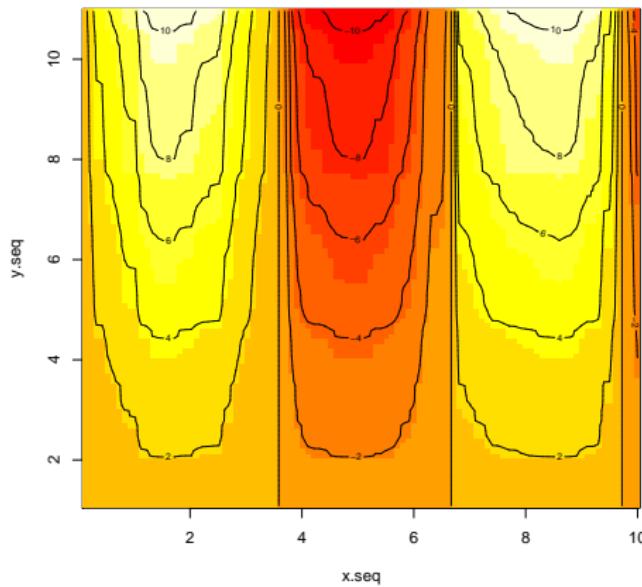




Performing the Simulation and Visualizing the Results

# Simulations II

Visualize with `image()`



# Simulations I

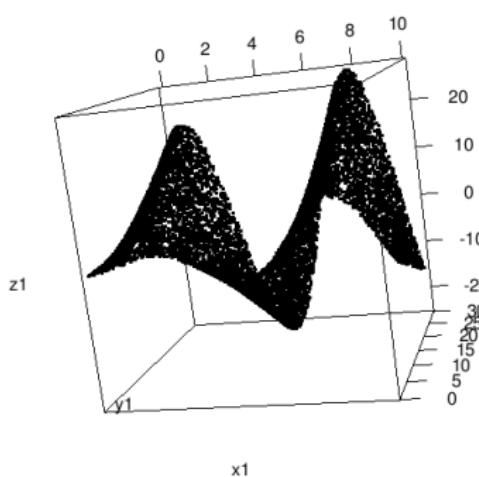
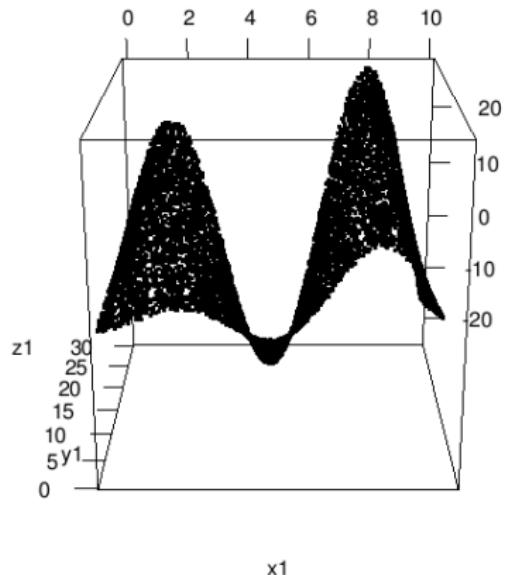
Visualize with `plot3d()`

Suppose we want to know what the function  $y \times \sin(x)$  looks like with a dynamic plot:

```
1 # Sample from the random uniform:  
2 x1 <- sort(runif(10000, min=0, max=10))  
3 y1 <- x1+runif(1000, min=0, max=20)  
4 z1<-y1*sin(x1)  
5 library(rgl)  
6 plot3d(x1, y1, z1)
```

# Simulations II

Visualize with `plot3d()`



# Exercise V: Homework Problem

Visualize the following function:

$$z = \frac{\sin(\frac{3}{2}x)}{y} \quad (1)$$



## 1 Summary Plots

## 2 Time Series Plots

## 3 Geographical Plots

## 4 3D Plots

## 5 Simulation Plots

## 6 Useful Links for R

# Online Resources for R

Download R: <http://cran.stat.ucla.edu/>

Search Engine for R: <http://rseek.org>

R Reference Card:

<http://cran.r-project.org/doc/contrib/Short-refcard.pdf>

R Graphics Gallery:

<http://research.stowers-institute.org/efg/R/>

R Graph Gallery: <http://addictedtor.free.fr/graphiques/>

UCLA Statistics Information Portal: <http://info.stat.ucla.edu/grad/>

UCLA Statistical Consulting Center: <http://scc.stat.ucla.edu>



Thank you.  
Any questions?

