Statistics with R - Beginner Level

Section 4

Building Charts

Lesson 21 – Histograms

```
geom histogram(data=demo, aes(x=income), fill="red",
color="black")
### represent the density on the y axis (relative
frequencies)
qqplot()+
  geom histogram(data=demo, aes(x=income, y=..density..),
fill="red", color="black")
#########
### create a facet grid (multiple histograms)
### we will create a histogram for each combination of the
variables
### gender and marital status
qqplot()+
  geom histogram(data=demo, aes(x=income, y=..density..),
fill="red", color="black")+
  facet grid(gender~marital)
### create multiple histograms on the same chart
### we will build a histogram for each gender category
qqplot()+
  geom histogram (data=demo, aes (x=income, y=..density..,
fill=gender), color="black")
### N. B. the "fill" variable must be a factor
```

Lesson 22 - Cumulative Frequency Line Charts

```
### we will create a cumulative frequencies line for the
variable income
### load the packages
require (ggplot2)
require(plyr) ### we need the count function
### create a data frame with the unique income values
mydata <- count(demo, 'income')</pre>
View(mydata)
### compute the cumulative counts and percentages
cumul <- cumsum(mydata$freq)</pre>
cumperc <- cumul/nrow(demo)</pre>
### add the cumulative frequencies column to the iniatial
mydata matrix
mydata <- cbind(mydata, cumperc)</pre>
View(mydata)
### plot the cumulative frequencies line (smooth)
ggplot()+geom line(data=mydata, aes(x=income, y=cumperc))
### OR plot the stepped line
ggplot()+geom step(data=mydata, aes(x=income, y=cumperc))
#################
### create grouped cumulative frequencies lines
### we will build a cumulative frequencies line chart for
the variable income
### for each gender group
```

```
## first we create two databases, by gender, using the
brackets
male <- demo[demo$gender=="Male",]</pre>
female <- demo[demo$gender=="Female",]</pre>
View(male)
View(female)
### for the male data frame, we get the unique income
values
### then compute the cumulative relative frequencies
mydata male <- count(male, "income")</pre>
cumulm <- cumsum(mydata male$freq)</pre>
cumpercm <- cumulm/nrow(male)</pre>
### add the cumulative relative frequencies column
mydata male <- cbind(mydata male, cumpercm)</pre>
View(mydata male)
### the same for the female data frame
mydata female <- count(female, "income")</pre>
cumulf <- cumsum(mydata female$freq)</pre>
cumpercf <- cumulf/nrow(female)</pre>
mydata female <- cbind(mydata female, cumpercf)</pre>
View(mydata female)
### now we can build the chart
ggplot()+geom line(data=mydata male, aes(x=income,
y=cumpercm), color="red")+
```

```
geom_line(data=mydata_female, aes(x=income, y=cumpercf),
color="blue")

### for a stepped line we must replace geom_line with
geom_step

### add a legent to the chart

lgd <- scale_color_manual("Legend", values=c(Male="red",
Female="blue"))

ggplot()+
   geom_line(data=mydata_male, aes(x=income, y=cumpercm,
color="Male"), size=1.3)+
   geom_line(data=mydata_female, aes(x=income, y=cumpercf,
color="Female"), size=1.3)+
   lgd</pre>
```

Lesson 23 - Column Charts

```
ggplot(demo, aes(x=educ, y=income))+
   stat_summary(fun.y=mean, geom="bar", fill="red")

### N.B. if the grouping variable is not a factor,
### we must convert it into a factor

### to create a clustered bar chart (by the variable gender)
### position_dodge will put the columns side by side

ggplot(demo,aes(x=educ, y=income, fill=gender)) +
   stat_summary(fun.y=mean, geom="bar",
position=position_dodge())

### to stack the columns we use position_stack

ggplot(demo,aes(x=educ, y=income, fill=gender)) +
   stat_summary(fun.y=mean, geom="bar",
position=position_stack())
```

Lesson 24 - Mean Plot Charts

```
demo <- read.csv("demographics.csv")

View(demo)

##########

### how to build mean plot charts with ggplot2
##########

### we will create a mean plot representing the average income
### for each gender category

### load the package

require(ggplot2)

### create the dataframe with the means of the gender groups</pre>
```

```
aggdata <- aggregate(demo$income, by=list(demo$gender),</pre>
FUN=mean)
View (aggdata)
### draw the plot
### the x axis is defined as discrete, with convenient
labels
ggplot()+geom line(data=aggdata, aes(x=(1:2),
y=aggdata$x))+
  scale x discrete(name="Gender", labels=c("Female",
"Male"))+
  scale y continuous(name="Income", limits=c(72, 85))
### change line color and thickness
ggplot()+geom line(data=aggdata, aes(x=(1:2), y=aggdata$x),
color="red", size=1.3)+
  scale x discrete(name="Gender", labels=c("Female",
"Male"))+
  scale y continuous(name="Income", limits=c(72, 85))
#########
### build the chart with a polytomous factor
### (we'll get a broken line)
### the factor will be education level (educ)
aggdata <- aggregate(demo$income, by=list(demo$educ),</pre>
FUN=mean)
View (aggdata)
ggplot()+geom line(data=aggdata, aes(x=(1:5),
y=aggdata$x))+
  scale x discrete(name="Education Level",
labels=c("College degree", "Did not complete high school",
"High school degree", "Post-undergraduate degree", "Some
college"))+
  scale y continuous(name="Income", limits=c(64, 116))
```

```
###############
### build a grouped mean plot
### the grouping variable will be the car category (carcat)
## create three data frames for the economy, standard and
luxury cars
demo ec <- demo[demo$carcat=="Economy",]</pre>
demo st <- demo[demo$carcat=="Standard",]</pre>
demo lu <- demo[demo$carcat=="Luxury",]</pre>
# compute the mean income for each education level and for
each data frame (car category)
agg ec <- aggregate (demo ec$income, by=list(demo ec$educ),
FUN=mean)
agg st <- aggregate (demo st$income, by=list(demo st$educ),
FUN=mean)
agg lu <- aggregate (demo lu$income, by=list(demo lu$educ),
FUN=mean)
View (agg ec)
View(agg st)
View(agg lu)
## plot the three lines on the same graph
gaplot()+
  geom line(data=agg ec, aes(x=(1:5), y=agg ec$x),
color="green") +
  geom line (data=agg st, aes (x=(1:5), y=agg st$x),
color="red") +
  geom line(data=agg lu, aes(x=(1:5), y=agg lu$x),
color="blue")+
  scale x discrete(name="Education Level",
labels=c("College degree", "Did not complete high school",
```

```
"High school degree", "Post-undergraduate degree", "Some
college"))+
  scale y continuous(name="Income", limits=c(15, 220))
## add legend
lgd <- scale color manual(name="Legend",</pre>
values=c(Economy="green", Standard="red", Luxury="blue"))
aaplot()+
  geom line(data=agg ec, aes(x=(1:5), y=agg ec$x,
color="Economy"))+
  geom line (data=agg st, aes (x=(1:5), y=agg st$x,
color="Standard"))+
  geom line(data=agg lu, aes(x=(1:5), y=agg lux,
color="Luxury"))+
  scale x discrete(name="Education Level",
labels=c("College degree", "Did not complete high school",
"High school degree", "Post-undergraduate degree", "Some
college"))+
  scale y continuous(name="Income", limits=c(15, 220))+lqd
### N.B. in the code above, the color argument is found in
the aesthetics section
```

Lesson 25 - Scatterplot Charts

```
hw <- read.csv("hw.csv")

View(hw)

##############
### how to build scatterplot charts with ggplot2
###########

### we will create a scatterplot with the variables
### height and weight

### load the package

require(ggplot2)</pre>
```

```
### create the plot
ggplot()+geom point(data=hw, aes(x=height, y=weight))+
  scale x continuous(limits=c(150,193))
########
### build a clustered scatterplot
### by gender
lqd <- hw$qender
### get points of different colors
ggplot()+geom point(data=hw, aes(x=height, y=weight,
color=lqd))+
  scale x continuous(limits=c(150,193))
### get points of different shapes
ggplot()+geom point(data=hw, aes(x=height, y=weight,
shape=lqd))+
  scale x continuous(limits=c(150,193))
### get points of both different shapes and colors
ggplot()+geom point(data=hw, aes(x=height, y=weight,
shape=lgd, color=lgd))+
  scale x continuous(limits=c(150,193))
##########
### add a trendline to the scatterplot
### create a linear model
### with weight as the dependent variable and height as the
explainer
model <- lm(weight~height, data=hw)</pre>
print(model)
### get the minimum and the maximum height
```

```
minh <- min(hw$height)</pre>
maxh <- max(hw$height)</pre>
### create a new vector height
height <- c(minh, maxh)
print(height)
### predict the weight based on the height, with the model
above
fit <- predict(model, data.frame(height))</pre>
print(fit)
### create a data frame with the line end points
endpoints <- data.frame(height, fit)</pre>
View(endpoints)
### build the scatter plot with trend line
ggplot()+
  geom point(data=hw, aes(x=height, y=weight))+
  geom line(data=endpoints, aes(x=height, y=fit),
color="red", size=1)
Lesson 26 - Boxplot Charts
```

```
### for each gender category
### load the package
require(ggplot2)
### create the plot
ggplot()+geom boxplot(data=demo, aes(x=gender, y=income))+
  scale x discrete(labels=c("Female", "Male"))
### N.B. if the grouping variable is not a factor
### make sure you convert it into a factor first
### set the color of the outliers
ggplot()+geom boxplot(data=demo, aes(x=gender, y=income),
outlier.colour="red")+
  scale x discrete(labels=c("Female", "Male"))
### set the shape of the outliers
ggplot()+geom boxplot(data=demo, aes(x=gender, y=income),
outlier.colour="red", outlier.shape=4)+
  scale x discrete(labels=c("Female", "Male"))
### add a legend
lgd <- demo$gender</pre>
ggplot()+geom boxplot(data=demo, aes(x=gender, y=income,
fill=lgd), outlier.colour="red")+
  scale x discrete(labels=c("Female", "Male"))
#########
### build a clustered boxplot
### we will group the boxplots by gender and marital status
```

```
### the legend will represent the two marital statuses

lgd <- demo$marital

ggplot()+geom_boxplot(data=demo, aes(x=gender, y=income, fill=lgd))+
   scale x discrete(labels=c("Female", "Male"))</pre>
```

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