# Lab 02R

36-290 - Statistical Research Methodology

Week 2 Thursday - Fall 2021

#### Data

We'll importing the same data that we used for Tuesday's lab, with rows with data with value 99 removed:

```
suppressMessages(library(tidyverse))
 rm(list=ls())
 file.path = "https://raw.githubusercontent.com/pefreeman/36-290/master/EXAMPLE DATASETS/BUZZARD/Buzzard DC1.Rdata"
 load(url(file.path))
rm(file.path)
set.seed(101)
s = sample(nrow(df), 4000)
df = df[s,]
\label{eq:continuity} $$ df  \space{-0.005in} $$ filter(.,u!=99\&g!=99\&r!=99\&i!=99\&z!=99\&y!=99) -> df.new $$ $$ filter(.,u!=99\&g!=99\&r!=99\&i!=99\&z!=99\&y!=99) -> df.new $$ $$ filter(.,u!=99\&g!=99\&r!=99\&i!=99\&z!=99\&y!=99) -> df.new $$ $$ filter(.,u!=99\&g!=99\&p!=99\&i!=99\&z!=99\&y!=99) -> df.new $$ $$ filter(.,u!=99\&g!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=99\&p!=9
predictors = df.new[,-c(7:14)]
 response = as.vector(df.new[,14])
type = rep("FAINT",nrow(predictors))
w = which(predictors$i<25)
type[w] = "BRIGHT"
type = factor(type)
predictors = cbind(type,predictors)
rm(df,df.new,s,type,w)
objects()
```

```
## [1] "predictors" "response"
```

If everything loaded correctly, you should see two variables in your global environment: predictors and response predictors is a data frame with 3624 rows and 6 columns, and response is a vector of length 3624, and it represents redshift.

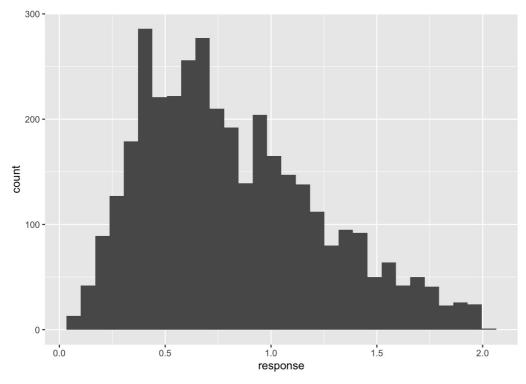
#### ggplot

In this course, we will use ggplot to create our plots. You will learn much about ggplot in 36-315; here, we will use it relatively simply, to make histograms and boxplots and scatter plots and the like. In order to become comfortable quickly with ggplot, you should read through Chapter 3 (online; Chapter 1 in print) of *R for Data Science* by Wickham and Grolemund, available for free here (http://r4ds.had.co.nz/).

The following is an example of a ggplot call:

```
ggplot(data=data.frame(response),mapping=aes(x=response)) + geom_histogram()
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```



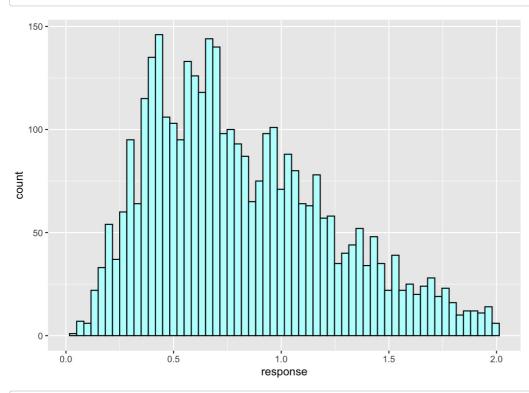
Note that ggplot expects data frames (or tibbles...), so I had to add the call to data.frame . Also note the structure here: the first function is ggplot(), and you pass a data frame (as data) and an identification of which feature goes along which axis (as mapping). Once you've identified the data and the mapping, you "add on" another function that describes what you will do to the data; in this case, that function is geom\_histogram().

# Questions

### Question 1

Re-run ggplot from above, doubling the number of bins. (Use the documentation for <code>geom\_histogram()</code> to figure out how to do this...e.g., type <code>?geom\_histogram</code> in the Console pane.) Also, change the color of the histogram via the <code>fill</code> argument. Google "R colors" to find listings of the names of R 's colors. My personal favorite is "papayawhip". Add a comment within the code chunk describing the empirical distribution of redshifts (where's the mode? is it bimodal? is it skew? why might it have the properties it does? etc.)

ggplot(data=data.frame(response), mapping=aes(x=response)) + geom\_histogram(color="black", fill="paleturquoise1", bin s=60)

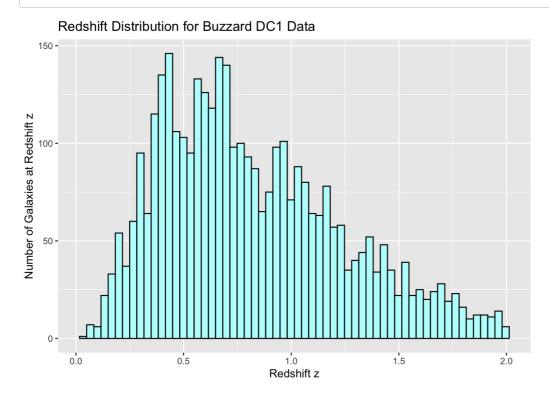


modality: unimodal; skewed right; jaggedness due to randomness, noise

#### Question 2

Repeat Q1, but change the x-axis label to "Redshift z" and the y-axis label to "Number of Galaxies at Redshift z". Also, add a title to the plot: "Redshift Distribution for Buzzard DC1 Data".

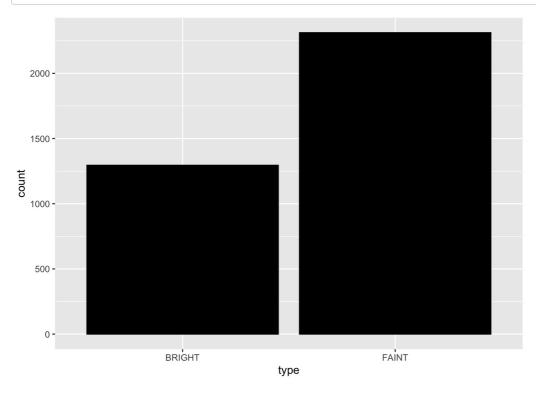
 $ggplot(data=data.frame(response), mapping=aes(x=response)) + geom_histogram(color="black", fill="paleturquoise1", bin s=60) + ggtitle("Redshift Distribution for Buzzard DC1 Data") + xlab("Redshift z") + ylab("Number of Galaxies at Redshift z")$ 



## Question 3

Construct a bar chart that shows the distribution of galaxy type, here defined to be BRIGHT and FAINT. Remember that a bar chart is a representation of a probability mass function, or pmf. As a reminder to yourself, also show the number of counts for each type, coded in some manner that you've learned thus far.

ggplot(data=data.frame(predictors),mapping=aes(x=type)) + geom\_bar(color="black", fill="black")



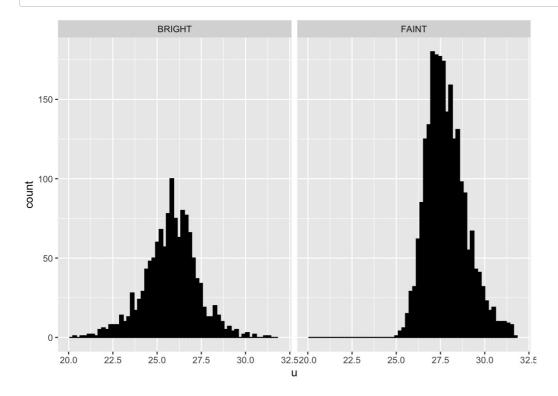
#### Question 4

Utilize facet wrapping to display two histograms side-by-side: one showing u -band magnitudes for BRIGHT data, and one showing u -band magnitudes for FAINT data. Note that to compare distributions, it is helpful to have the data span the same scale. So add a function to the string of functions that sets the x-axis limits to be 20 and 32.

 $ggplot(data=data.frame(predictors), mapping=aes(x=u)) + geom\_histogram(color="black", fill="black", bins=60) + facet \\ \_wrap(~type, scales='free\_x') + xlim(20, 32)$ 

## Warning: Removed 20 rows containing non-finite values (stat\_bin).

## Warning: Removed 4 rows containing missing values (geom\_bar).

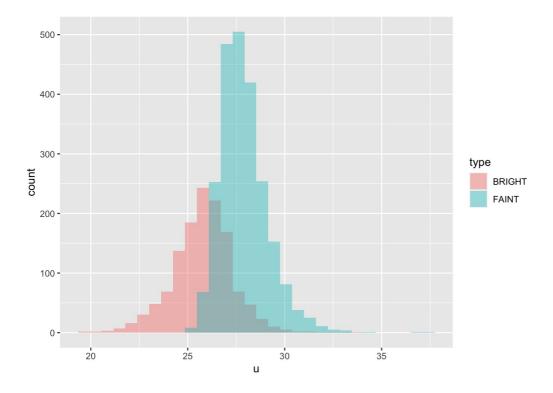


## Question 5

Show the same information as in Q4, except use overlapping, partially transparent histograms. Construct such a histogram (again for u). (Hints: in the call to aes() in the ggplot() call, include fill=type, and in the geom\_histogram() call, include two new arguments: alpha and position="Identity". The alpha argument controls the transparency: 0 for invisible to 1 for totally opaque.

ggplot(data=data.frame(predictors),mapping=aes(x=u, fill=type)) + geom\_histogram( alpha=.4, position="Identity")

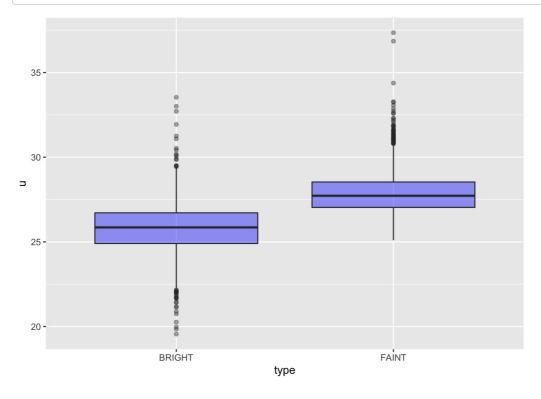
## `stat\_bin()` using `bins = 30`. Pick better value with `binwidth`.



# Question 6

Show the same information as in Q4, except that instead of using side-by-side histograms, use side-by-side boxplots. Unlike in Q4, you do not need to apply  $facet\_wrap()$ . Instead, for the aesthetics, specify that x is type and y is u.

ggplot(data=data.frame(predictors),mapping=aes(x=type, y=u)) + geom\_boxplot(alpha=.4, fill="blue")



#### Question 7

In order to illustrate the gather() function, we are going to remove the type column from predictors:

```
# UNCOMMENT THE LINE BELOW AND RUN CHUNK
pred.notype = predictors[,-1] # keep all rows, remove first column
```

Apply the gather() function to pred.notype, and show the first six rows of output by piping to the head() function. What are the names of the columns of the data frame output by gather()? (To understand more fully what gather() is doing: it is taking a multi-column data frame and reshaping it to have two columns: the first contains the variable name, and the second the variable values.)

```
gather(pred.notype) %>% head()
```

```
## key value
## 1 u 23.5907
## 2 u 27.1252
## 3 u 27.4013
## 4 u 27.9354
## 5 u 27.7733
## 6 u 27.1742
```

Key and Value

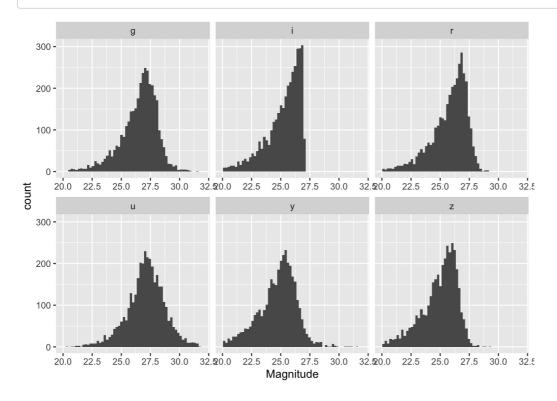
#### Question 8

Create a faceted histogram that takes gather(pred.notype) as the input data frame and value as the aesthetic, and wraps on key . Set the x-axis limits to be 20 and 32. Replace the x-axis label with "Magnitude".

```
ggplot(data=gather(pred.notype), mapping=aes(x=value)) + geom\_histogram(bins=60) + facet\_wrap(\sim key, scales='free\_x') + xlim(20, 32) + xlab("Magnitude")
```

## Warning: Removed 230 rows containing non-finite values (stat\_bin).

## Warning: Removed 12 rows containing missing values (geom\_bar).

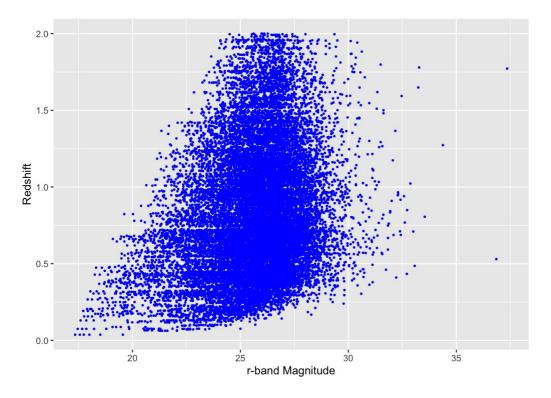


## Question 9

Use <code>geom\_point()</code> to create a scatter plot of redshift vs. r-band magnitude. (Remember, one plots y vs. x, so redshift will go on the y-axis here.) Make the point size 0.5, change the x-axis label to "r-band magnitude", and change the y-axis label to "Redshift". Oh, and add some color!

```
df <- cbind(gather(pred.notype), response)

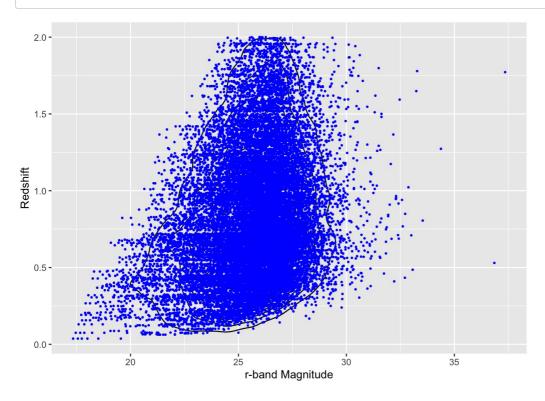
ggplot(data=df,mapping=aes(x=value, y=response)) + geom_point(color="blue", size=.5) + xlab("r-band Magnitude")+ ylab("Redshift")</pre>
```



# Question 10

An issue with scatter plots is that they are not, by themselves, estimates of bivariate distributions; their points are simply samples from such distributions. To try to visualize the underlying distribution, one can do a few things. Here, repeat Q9, and add <code>geom\_density\_2d()</code> so as to overlay a contour plot of the estimated density from which the r-band magnitude and the redshift are sampled.

ggplot(data=df,mapping=aes(x=value, y=response)) + geom\_density\_2d(color="black") + geom\_point(color="blue", size=.
5) + xlab("r-band Magnitude")+ ylab("Redshift")



## Question 11

Repeat Q10, but use geom\_hex() instead. You may have to install the package hexbin first!

```
library(hexbin)

ggplot(data=df,mapping=aes(x=value, y=response)) + geom_hex() + geom_point(color="blue", size=.5) + xlab("r-band Ma
gnitude")+ ylab("Redshift")
```

