

Examples of some uwIntroStats Functions

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1 Introduction

This document is meant to illustrate the use of some of the major **uwIntroStats** functions. The examples will be shown with R code and output. For more options, see the help files provided with the package. Thus, before starting, load both the packages that **uwIntroStats** relies on:

```
> ## uwIntroStats relies on the Exact, survival,  
> ## plyr, and sandwich packages  
> library(Exact)  
> library(survival)  
> library(plyr)  
> library(sandwich)
```

Now load the **uwIntroStats** package (if you don't have it installed, go to "emersonstatistics.com/R" and download the appropriate file).

```
> library(uwIntroStats)
```

We will also be working with the **mri** dataset, so prepare that as well:

```
> data(mri)  
> attach(mri)
```

The reference manual for the **mri** dataset can be found on "emersonstatistics.com/Datasets".

2 Descriptive Statistics

2.1 descrip()

Now that we have our package and dataset loaded, we can delve deeper into the functions. When we first get a dataset, we often want to find some descriptive statistics. The most common are the number of observations a variable has, the number of missing observations, the mean, median, standard deviation, and some others. The default **descrip()** function will give all of this and more to us. Let's say we wanted to get descriptive statistics for **age**:

```
> descrip(age)
```

	N	Msng	Mean	Std Dev	Min	25%	Mdn	75%
age:	735	0	74.57	5.451	65.00	71.000000	74.00	78.00
		Max						
age:	99.00							

However, what if we are interested in the whole dataset? We can do this too!

```
> descrip(mri)
```

	N	Msng	Mean	Std Dev	Min	25%	Mdn
ptid:	735	0	368.0	212.3	1.000	1.845000e+02	368.0
mridate:	735	0	76423	31896	10192	6.664200e+04	80992
age:	735	0	74.57	5.451	65.00	7.100000e+01	74.00
male:	735	0	0.4980	0.5003	0.0000	0.000000e+00	0.0000
race:	735	0	1.318	0.6659	1.000	1.000000e+00	1.000
weight:	735	0	159.9	30.74	74.00	1.385000e+02	158.0
height:	735	0	165.8	9.710	139.0	1.580000e+02	165.9
packyrs:	735	1	19.60	27.11	0.0000	0.000000e+00	6.500
yrsquit:	735	0	9.661	14.10	0.0000	0.000000e+00	0.0000
alcoh:	735	0	2.109	4.852	0.0000	0.000000e+00	0.01920
physact:	735	0	1.922	2.052	0.0000	5.537500e-01	1.312
chf:	735	0	0.05578	0.2297	0.0000	0.000000e+00	0.0000
chd:	735	0	0.3347	0.6862	0.0000	0.000000e+00	0.0000
stroke:	735	0	0.2367	0.6207	0.0000	0.000000e+00	0.0000
diabetes:	735	0	0.1075	0.3099	0.0000	0.000000e+00	0.0000
genhlth:	735	0	2.588	0.9382	1.000	2.000000e+00	3.000
ldl:	735	10	125.8	33.60	11.00	1.020000e+02	125.0
alb:	735	2	3.994	0.2690	3.200	3.800000e+00	4.000
crt:	735	2	1.064	0.3030	0.5000	9.000000e-01	1.000
plt:	735	7	246.0	65.80	92.00	2.017500e+02	239.0
sbp:	735	0	131.1	19.66	78.00	1.180000e+02	130.0
aai:	735	9	1.103	0.1828	0.3171	1.026900e+00	1.112
fev:	735	10	2.207	0.6875	0.4083	1.745000e+00	2.158
dsst:	735	12	41.06	12.71	0.0000	3.200000e+01	40.00
atrophy:	735	0	35.98	12.92	5.000	2.700000e+01	35.00
whgrd:	735	1	2.007	1.410	0.0000	1.000000e+00	2.000
numinf:	735	0	0.6109	0.9895	0.0000	0.000000e+00	0.0000
volinf:	735	1	3.223	17.36	0.0000	0.000000e+00	0.0000
obstime:	735	0	1804	392.3	68.00	1.837000e+03	1879
death:	735	0	0.1810	0.3852	0.0000	0.000000e+00	0.0000
	75%	Max					
ptid:	551.5	735.0					
mridate:	91392	1.232e+05					
age:	78.00	99.00					
male:	1.000	1.000					
race:	1.000	4.000					
weight:	179.0	264.0					
height:	173.2	190.5					
packyrs:	33.75	240.0					
yrsquit:	18.50	56.00					
alcoh:	1.144	35.00					
physact:	2.513	13.81					
chf:	0.0000	1.000					
chd:	0.0000	2.000					
stroke:	0.0000	2.000					

```

diabetes:    0.0000    1.000
  genhlth:    3.000    5.000
    ldl:     147.0    247.0
    alb:      4.200    5.000
    crt:      1.200    4.000
    plt:     285.0    539.0
    sbp:     142.0    210.0
    aai:      1.207    1.728
    fev:      2.649    4.471
    dsst:     50.00    82.00
  atrophy:    44.00    84.00
    whgrd:     3.000    9.000
  numinf:     1.000    5.000
  volinf:    0.09420    197.0
  obstime:     2044    2159
    death:     0.0000    1.000

```

Now we know that the `male` variable can stratify the data. A natural question to ask is: what are the descriptive statistics for age stratified by sex?

```
> descrip(age, strata=male)
```

		N	Msng	Mean	Std Dev	Min	25%	Mdn
age:	All	735	0	74.57	5.451	65.00	71.000000	74.00
age:	Str 0	369	0	74.41	5.258	65.00	71.000000	73.00
age:	Str 1	366	0	74.73	5.642	66.00	71.000000	74.00
		75%		Max				
age:	All	78.00		99.00				
age:	Str 0	78.00		91.00				
age:	Str 1	78.00		99.00				

Other functionality of `descrip()`, as with all of the functions in R, can be found by typing

```
> ?descrip
```

2.2 tableStat()

The next step is to build tables of descriptive statistics. For example, suppose we wish to have a table with count, row percentage, column percentage, standard deviation, and range. This is easy with `tableStat()`! We will build this table using `stroke` as our variable, stratified by `race` and `male`

```
> tableStat(stroke, race, male, stat="count=@count@; row%=@row%@ col%=@col%@; sd=@sd@; range = @min@ - @max@")
```

Tabled descriptive statistics by strata

Call:

```

tableStat.default(variable = stroke, race, male, stat = "count=@count@; row%=@row%@ col%=@col%@; sd=@sd@; range = @min@ - @max@"
- NaN denotes strata with no observations
- NA arises from missing or censored data

```

Format: count=Cnt; row%= % of row col%= % of col; sd=SD; range = Min - Max

```

male.0
race.1 count=286.0; row%= 50.0% col%= 77.5%; sd=0.5184; range = 0.0 - 2.000
race.2 count= 53.0; row%= 51.0% col%= 14.4%; sd=0.4666; range = 0.0 - 2.000
race.3 count= 26.0; row%= 55.3% col%= 7.0%; sd=0.6794; range = 0.0 - 2.000
race.4 count= 4.0; row%= 33.3% col%= 1.1%; sd=0.0000; range = 0.0 - 0.000
race.ALL count=369.0; row%= 50.2% col%=100.0%; sd=0.5219; range = 0.0 - 2.000

```

```

      male.1
race.1  count=286.0; row%= 50.0% col%= 78.1%; sd=0.7032; range = 0.0 - 2.000
race.2  count= 51.0; row%= 49.0% col%= 13.9%; sd=0.7013; range = 0.0 - 2.000
race.3  count= 21.0; row%= 44.7% col%=  5.7%; sd=0.7171; range = 0.0 - 2.000
race.4  count=  8.0; row%= 66.7% col%=  2.2%; sd=0.7071; range = 0.0 - 2.000
race.ALL count=366.0; row%= 49.8% col%=100.0%; sd=0.7010; range = 0.0 - 2.000
      male.ALL
race.1  count=572.0; row%=100.0% col%= 77.8%; sd=0.6210; range = 0.0 - 2.000
race.2  count=104.0; row%=100.0% col%= 14.1%; sd=0.5974; range = 0.0 - 2.000
race.3  count= 47.0; row%=100.0% col%=  6.4%; sd=0.6889; range = 0.0 - 2.000
race.4  count= 12.0; row%=100.0% col%=  1.6%; sd=0.5774; range = 0.0 - 2.000
race.ALL count=735.0; row%=100.0% col%=100.0%; sd=0.6207; range = 0.0 - 2.000

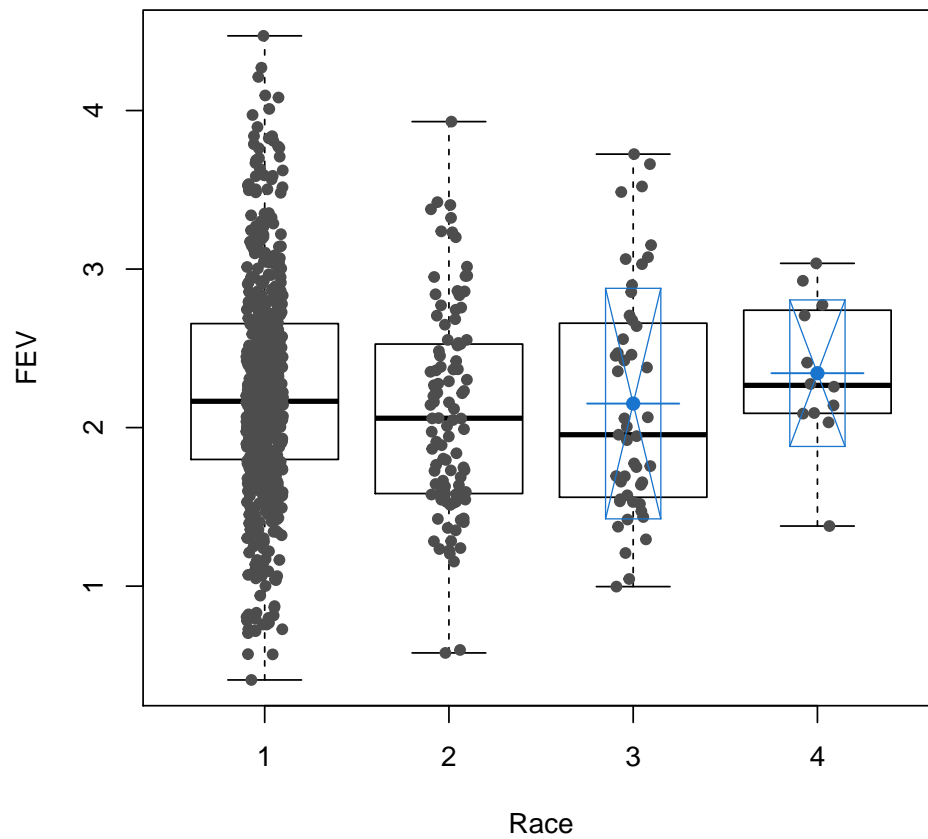
```

3 Plots

3.1 Box plots

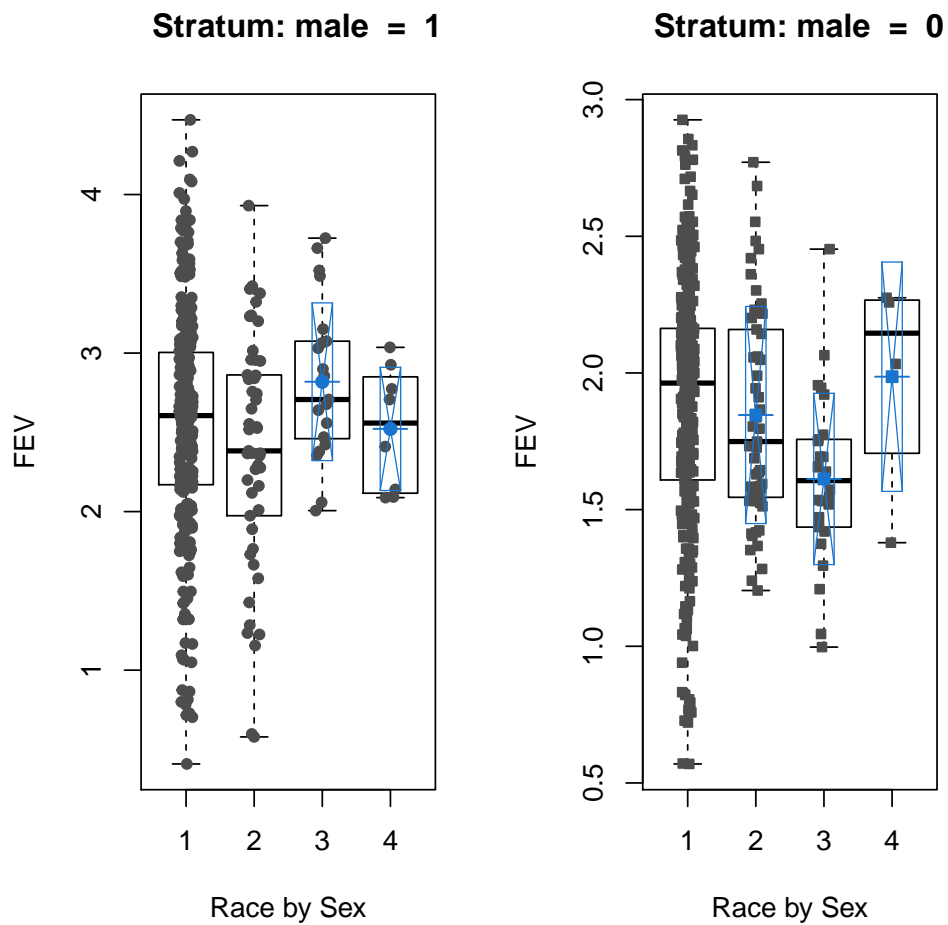
As we cover in our document “An Introduction to R”, box plots can be controversial as a descriptive plot of the data. However, we aimed to mitigate some of those concerns with our box plot function. It is straightforward to add jittered data to the plot (allowing us to see all of the data allows us to see where “outliers” really are) and overlay the sample mean and standard deviation - giving us a much better picture of the data. Let’s create a boxplot of fev by race:

```
> bplot(fev, race, xlab="Race", ylab="FEV")
```



Notice that by default the jittered data is added to the plot, and the plots are overlaid with sample mean and standard deviation. Now we can also stratify by sex:

```
> bplot(fev, race, strata=male, xlab="Race by Sex", ylab="FEV")
```



3.2 Scatter plots

We also often wish to view a scatter plot of the data.

4 Inference

4.1 `tabulate()`