

2-way ANOVA

FMU Biology Department

Analysis

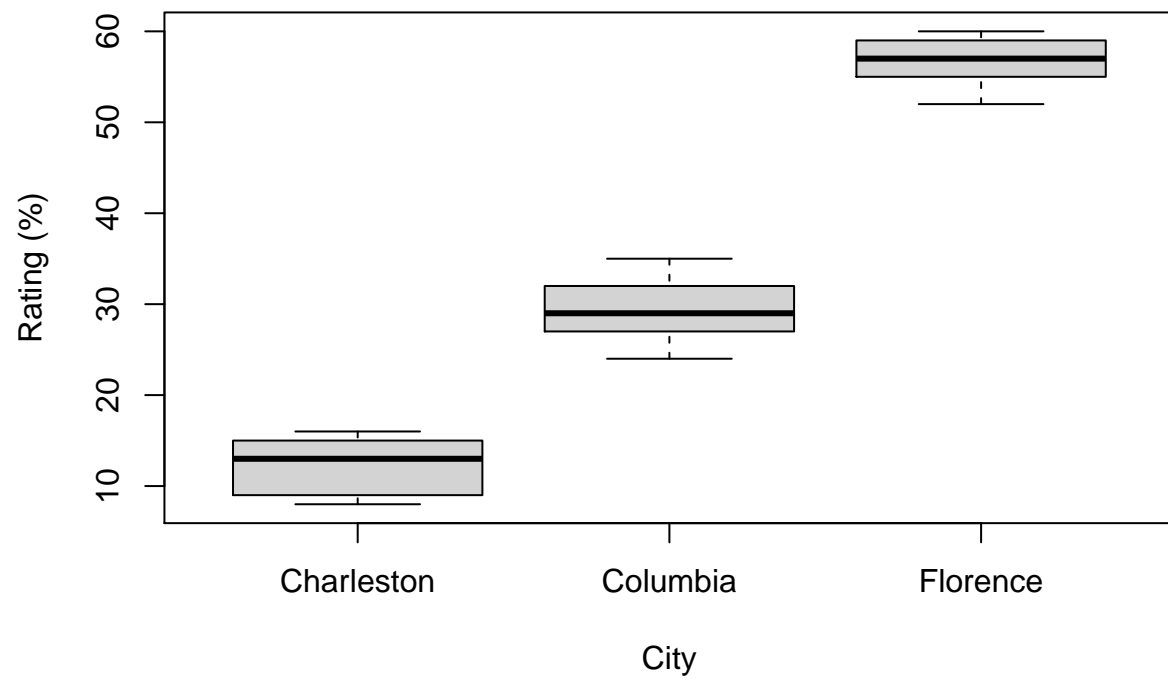
This documents includes code for creating a data object in R, creating an informative plot, and running an 2-way Analysis of Variance (ANOVA).

```
#Create data frame
dat1 = data.frame(rating = c(13,16,8,15,9,
                             29,35,24,27,32,
                             57,59,52,55,60),
                  city = c("Charleston","Charleston","Charleston","Charleston","Charleston",
                           "Columbia","Columbia","Columbia","Columbia","Columbia",
                           "Florence","Florence","Florence","Florence","Florence"),
                  politics = c("R","R","D","D","I",
                              "R","R","D","D","I",
                              "R","R","D","D","I"))

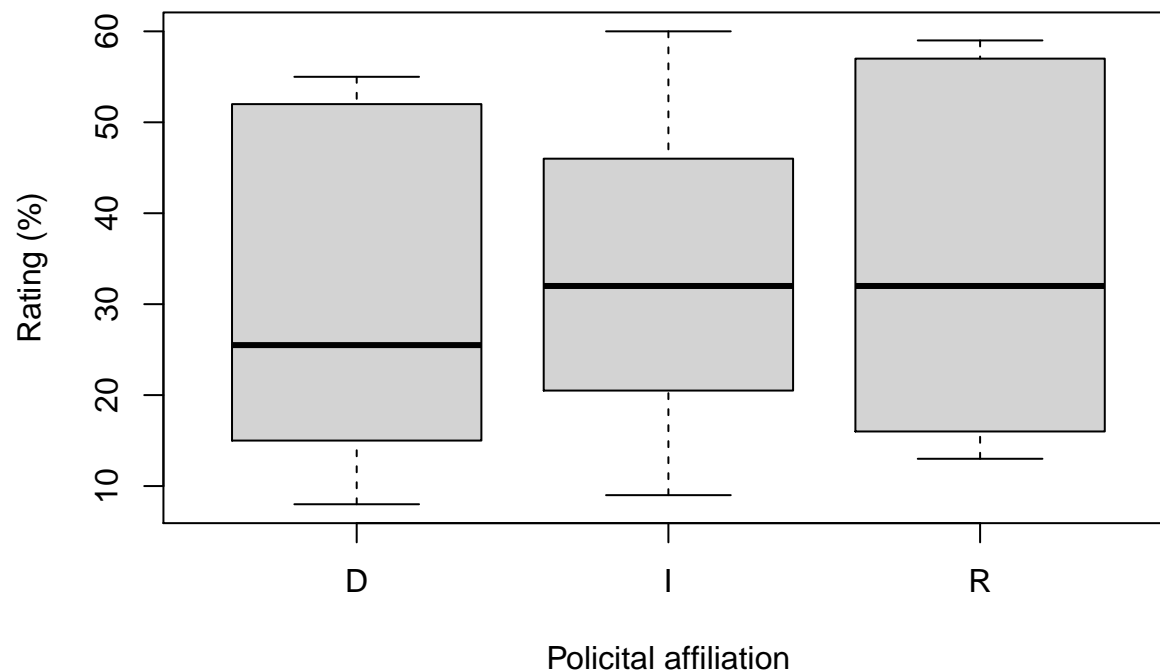
#Print data frame
dat1
```

```
##      rating      city politics
## 1         13 Charleston        R
## 2         16 Charleston        R
## 3          8 Charleston        D
## 4         15 Charleston        D
## 5          9 Charleston        I
## 6         29  Columbia        R
## 7         35  Columbia        R
## 8         24  Columbia        D
## 9         27  Columbia        D
## 10        32  Columbia        I
## 11        57  Florence        R
## 12        59  Florence        R
## 13        52  Florence        D
## 14        55  Florence        D
## 15        60  Florence        I
```

```
#Create Boxplot
boxplot(dat1$rating~dat1$city,
        ylab = "Rating (%)", xlab = "City")
```



```
#Create Boxplot  
boxplot(dat1$rating~dat1$politics,  
        ylab = "Rating (%)", xlab = "Political affiliation")
```



```
#ANOVA
```

```
res = aov(dat1$rating~dat1$city * dat1$politics)
```

```
#Display ANOVA results
```

```
summary(res)
```

```
##               Df Sum Sq Mean Sq F value    Pr(>F)
## dat1$city      2   5012   2505.9  259.228 1.5e-06 ***
## dat1$politics  2     69     34.3    3.548 0.0962 .
## dat1$city:dat1$politics 4     39      9.6    0.998 0.4759
## Residuals      6     58      9.7
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
#Tukey post-hoc comparisons
```

```
TukeyHSD(res)
```

```
##    Tukey multiple comparisons of means
##      95% family-wise confidence level
##
## Fit: aov(formula = dat1$rating ~ dat1$city * dat1$politics)
##
## $'dat1$city'
##               diff       lwr       upr      p adj
## Columbia-Charleston 17.2 11.16659 23.23341 0.0003036
## Florence-Charleston 44.4 38.36659 50.43341 0.0000013
```

```

## Florence-Columbia    27.2 21.16659 33.23341 0.0000217
##
## $'dat1$politics'
##      diff      lwr      upr      p adj
## I-D 3.500000 -3.2455532 10.24555 0.3189666
## R-D 4.666667 -0.8410545 10.17439 0.0899164
## R-I 1.166667 -5.5788866  7.91222 0.8596204
##
## $'dat1$city:dat1$politics'
##                                diff      lwr      upr      p adj
## Columbia:D-Charleston:D      1.400000e+01  0.1073131 27.8926869 0.0483445
## Florence:D-Charleston:D      4.200000e+01 28.1073131 55.8926869 0.0001492
## Charleston:I-Charleston:D -2.500000e+00 -19.5149971 14.5149971 0.9977692
## Columbia:I-Charleston:D      2.050000e+01  3.4850029 37.5149971 0.0212615
## Florence:I-Charleston:D      4.850000e+01 31.4850029 65.5149971 0.0002077
## Charleston:R-Charleston:D  3.000000e+00 -10.8926869 16.8926869 0.9767291
## Columbia:R-Charleston:D      2.050000e+01  6.6073131 34.3926869 0.0077410
## Florence:R-Charleston:D      4.650000e+01 32.6073131 60.3926869 0.0000834
## Florence:D-Columbia:D        2.800000e+01 14.1073131 41.8926869 0.0014579
## Charleston:I-Columbia:D     -1.650000e+01 -33.5149971  0.5149971 0.0571095
## Columbia:I-Columbia:D        6.500000e+00 -10.5149971 23.5149971 0.7302962
## Florence:I-Columbia:D        3.450000e+01 17.4850029 51.5149971 0.0014101
## Charleston:R-Columbia:D     -1.100000e+01 -24.8926869  2.8926869 0.1283358
## Columbia:R-Columbia:D        6.500000e+00 -7.3926869 20.3926869 0.5452766
## Florence:R-Columbia:D        3.250000e+01 18.6073131 46.3926869 0.0006337
## Charleston:I-Florence:D     -4.450000e+01 -61.5149971 -27.4850029 0.0003373
## Columbia:I-Florence:D       -2.150000e+01 -38.5149971 -4.4850029 0.0168815
## Florence:I-Florence:D        6.500000e+00 -10.5149971 23.5149971 0.7302962
## Charleston:R-Florence:D     -3.900000e+01 -52.8926869 -25.1073131 0.0002263
## Columbia:R-Florence:D       -2.150000e+01 -35.3926869 -7.6073131 0.0060460
## Florence:R-Florence:D        4.500000e+00 -9.3926869 18.3926869 0.8461629
## Columbia:I-Charleston:I      2.300000e+01  3.3527737 42.6472263 0.0243864
## Florence:I-Charleston:I      5.100000e+01 31.3527737 70.6472263 0.0003519
## Charleston:R-Charleston:I  5.500000e+00 -11.5149971 22.5149971 0.8473585
## Columbia:R-Charleston:I      2.300000e+01  5.9850029 40.0149971 0.0120885
## Florence:R-Charleston:I      4.900000e+01 31.9850029 66.0149971 0.0001960
## Florence:I-Columbia:I        2.800000e+01  8.3527737 47.6472263 0.0092527
## Charleston:R-Columbia:I     -1.750000e+01 -34.5149971 -0.4850029 0.0441816
## Columbia:R-Columbia:I       -3.552714e-15 -17.0149971 17.0149971 1.0000000
## Florence:R-Columbia:I        2.600000e+01  8.9850029 43.0149971 0.0064595
## Charleston:R-Florence:I     -4.550000e+01 -62.5149971 -28.4850029 0.0002975
## Columbia:R-Florence:I       -2.800000e+01 -45.0149971 -10.9850029 0.0043756
## Florence:R-Florence:I       -2.000000e+00 -19.0149971 15.0149971 0.9995168
## Columbia:R-Charleston:R      1.750000e+01  3.6073131 31.3926869 0.0171408
## Florence:R-Charleston:R      4.350000e+01 29.6073131 57.3926869 0.0001224
## Florence:R-Columbia:R        2.600000e+01 12.1073131 39.8926869 0.0021900

```