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Paste the URL (now in the clipboard) between the quotes in the command source("")

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You summarize the data by group with the command tapply(y,as.factor(x),summary)

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To start, we will examine the structure of the data frame with the command str(onewayanoval)

This should show us that onewayanoval is a data frame with 60 rows, and contains two numeric arrays called x and y.

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If we examine aov1 with str(aov1), we see that it is a list containing information about the linear model, much like the output from the lm function. aov() is customized for ANOVA while lm() is customized for regression.

The overall *F* test is significant if Pr(>F) is smaller than our chosen alpha value (usually 0.05).

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Once the overall *F* test is determined to be significant, we can do a follow-up test to determine which groups are different. Enter TukeyHSD(aov1) to display this.

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If zero does not lie between lwr and upr, we declare the difference between those two levels to be significant. Otherwise, we say it is not sifnificant.