

STA303/1004 - Week 3 R Markdown

January 21-25, 2019

Case Study 1: The Spock Conspiracy Trial Data

Get the data (from R library):

```
#load Sleuth3 R data library; see case0502  
library(Sleuth3)  
#Juries data  
jury = case0502  
#attach(jury)  
head(jury)
```

1st 6 rows

##	Percent	Judge
## 1	6.4	Spock's
## 2	8.7	Spock's
## 3	13.3	Spock's
## 4	13.6	Spock's
## 5	15.0	Spock's
## 6	15.2	Spock's

```
Percent=jury$Percent  
Judge=jury$Judge
```

Case Study 1: How many venires for each Judge?

factor

```
table(Judge)
```

```
## Judge
##      A      B      C      D      E      F Spock's
##      5      6      9      2      6      9      9
```

R function

```
with(jury, tapply(Percent, Judge, mean))
```

factor

```
##      A      B      C      D      E      F Spock's
## 34.12000 33.61667 29.10000 27.00000 26.96667 26.80000 14.62222
```

\bar{y}_A

\bar{y}_B

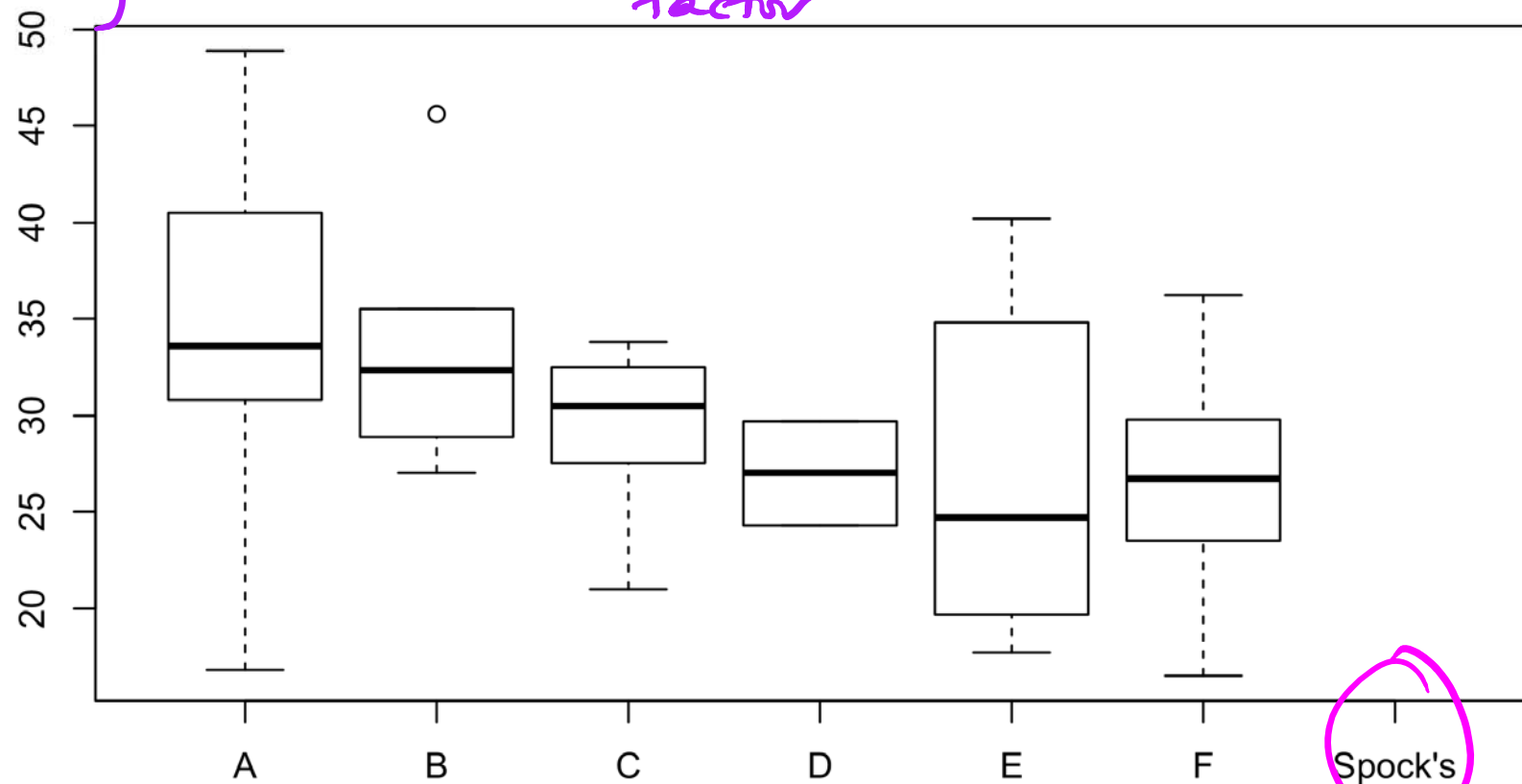
$$\hat{\beta}_B = \bar{y}_B - \bar{y}_A$$

$$E(y_i) = \beta_0 + \beta_i$$

$$\hat{y}_i = \bar{y}_i = b_0 + b_i$$

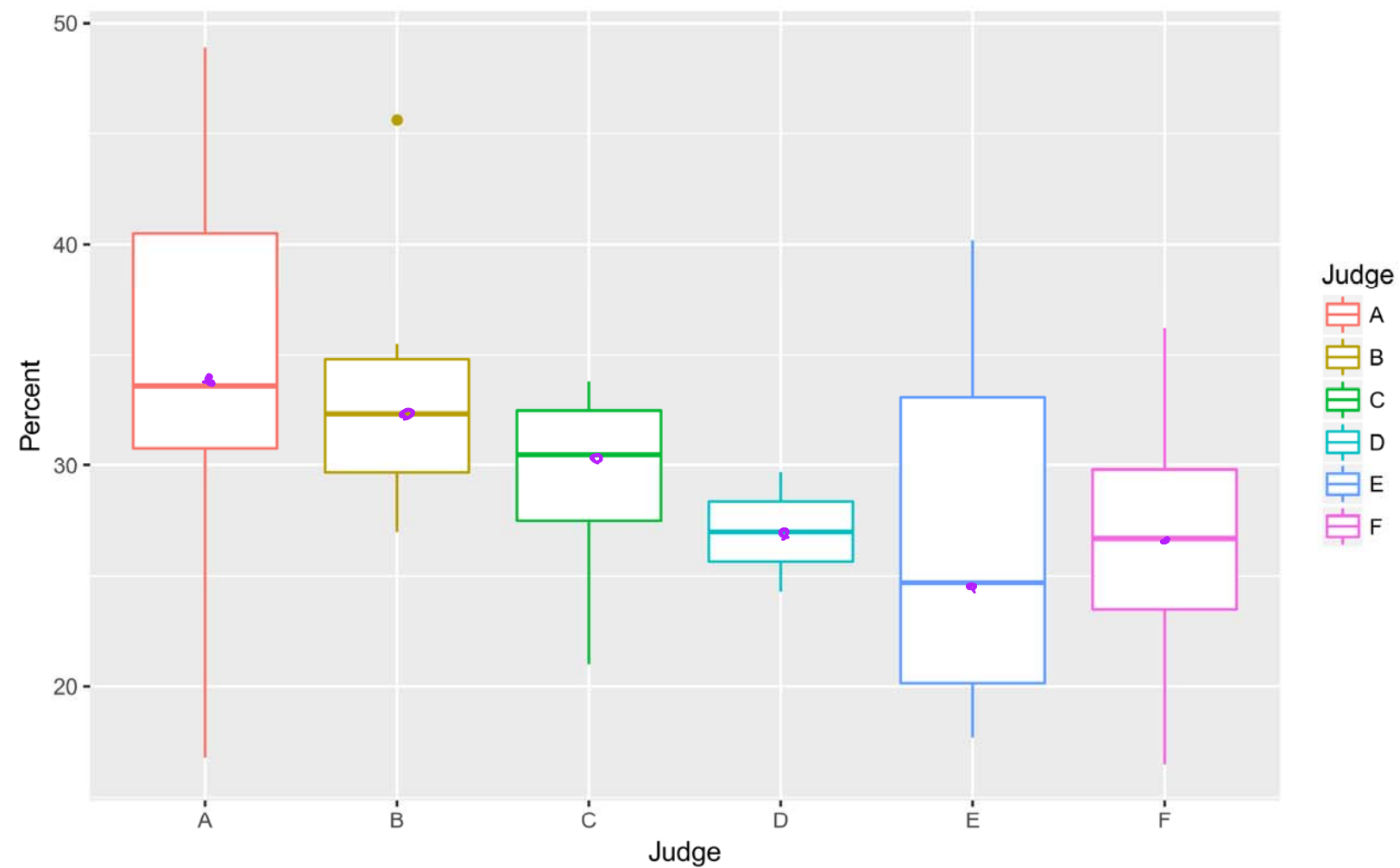
Case Study 1: Boxplot of Judges

```
# Get data subset of other judges  
Others <- subset(jury, Judge != "Spock's")  
boxplot(Percent ~ Judge, data=Others)
```



Case Study 1: Boxplot of Judges

```
#install.packages("ggplot2")  
library(ggplot2)  
ggplot(Others, aes(x=Judge, y=Percent, color=Judge)) + geom_boxplot()
```



Case Study 1: Q2-Compare the 6 other judges

`summary(aov(Percent ~ Judge, data=Others))`

##		Df	Sum Sq	Mean Sq	F value	Pr(>F)
##	Judge	5	326.5	65.29	1.218	0.324
##	Residuals	31	1661.3	53.59		

① $H_0: \mu_A = \mu_B = \mu_C = \mu_D = \mu_E = \mu_F$
 H_a : at least 2 means differ from each other

$${}^6C_2 = \binom{6}{2} = \frac{6(5)}{2} = 15$$

② Test Statistic = 1.218

$$\sim F_{5,31}$$

④ Conclusion

Evidence to support equality of means among the 6 other judges.

③ $p\text{-value} = 0.324$
 large $p\text{-value}$

Compare variances of 6 other judges: Rule of thumb

```
sss<-with(Others, tapply(Percent, Judge, sd))
sss
```

```
##           A           B           C           D           E           F      Spock'
## 11.941817  6.582224  4.592929  3.818377  9.010142  5.968878          N
```

```
dim(sss)
```

```
## [1] 7
```

```
max(sss, na.rm=T)
```

```
## [1] 11.94182
```

```
min(sss, na.rm=T)
```

```
## [1] 3.818377
```

```
isTRUE((max(sss, na.rm=T)/min(sss, na.rm=T))>2)
```

```
## [1] TRUE
```

$$\frac{11.94}{3.81} > 2$$

Compare variances of 6 other judges: Bartlett's

```
bartlett.test(Percent~Judge, data=Others)
```

```
##
```

```
## Bartlett test of homogeneity of variances
```

```
##
```

```
## data: Percent by Judge
```

```
## Bartlett's K-squared = 6.3125, df = 5, p-value = 0.277
```

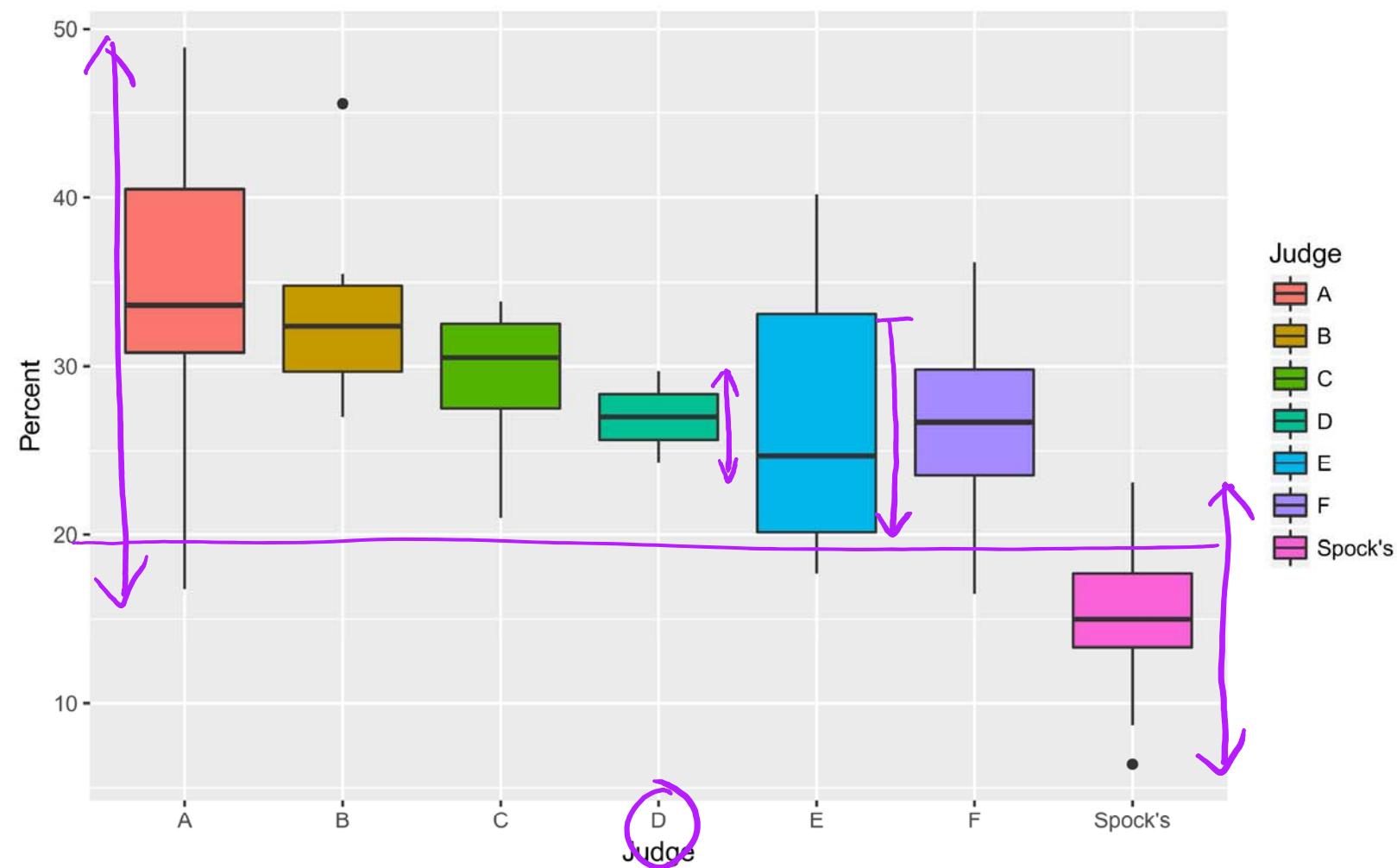
Note: Group sizes are uneven and some are very small

Evidence of equal variance.

Q3

Compare all 7 judges

```
#boxplot(Percent~Judge)
library(ggplot2)
ggplot(jury, aes(x=Judge,y=Percent, fill=Judge))+geom_boxplot()
```



Compare means of all 7 judges: One-way ANOVA

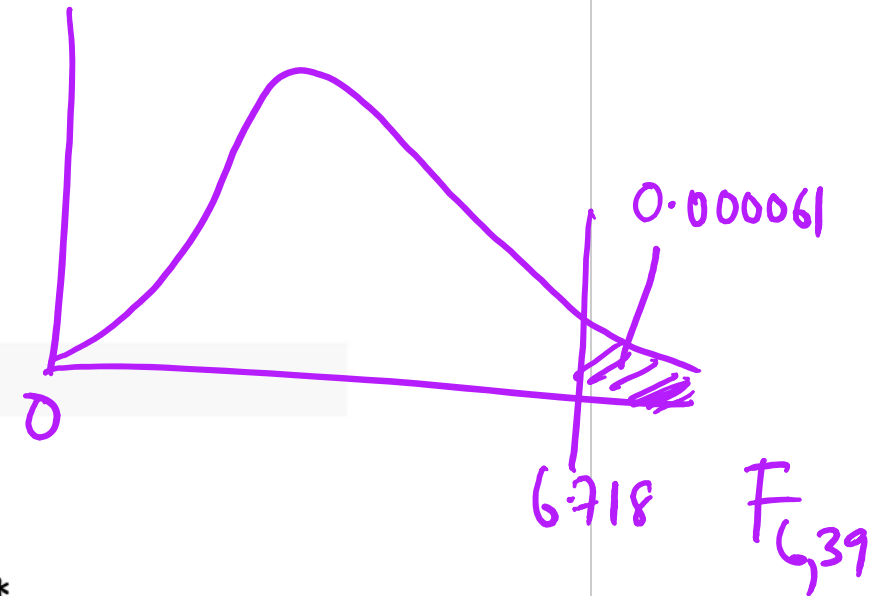
```
summary(aov(Percent ~ Judge))
```

	Df	Sum Sq	Mean Sq	F value	Pr(>F)						
Judge	6	1927	321.2	6.718	6.1e-05 ***						
Residuals	39	1864	47.8								

Signif. codes:	0	'***'	0.001	'**'	0.01	'*'	0.05	'.'	0.1	' '	1

7-1

46-7



Small p-value

Compare means of all 7 judges: Gen Linear Model

```
summary(lm(Percent ~ Judge))
```

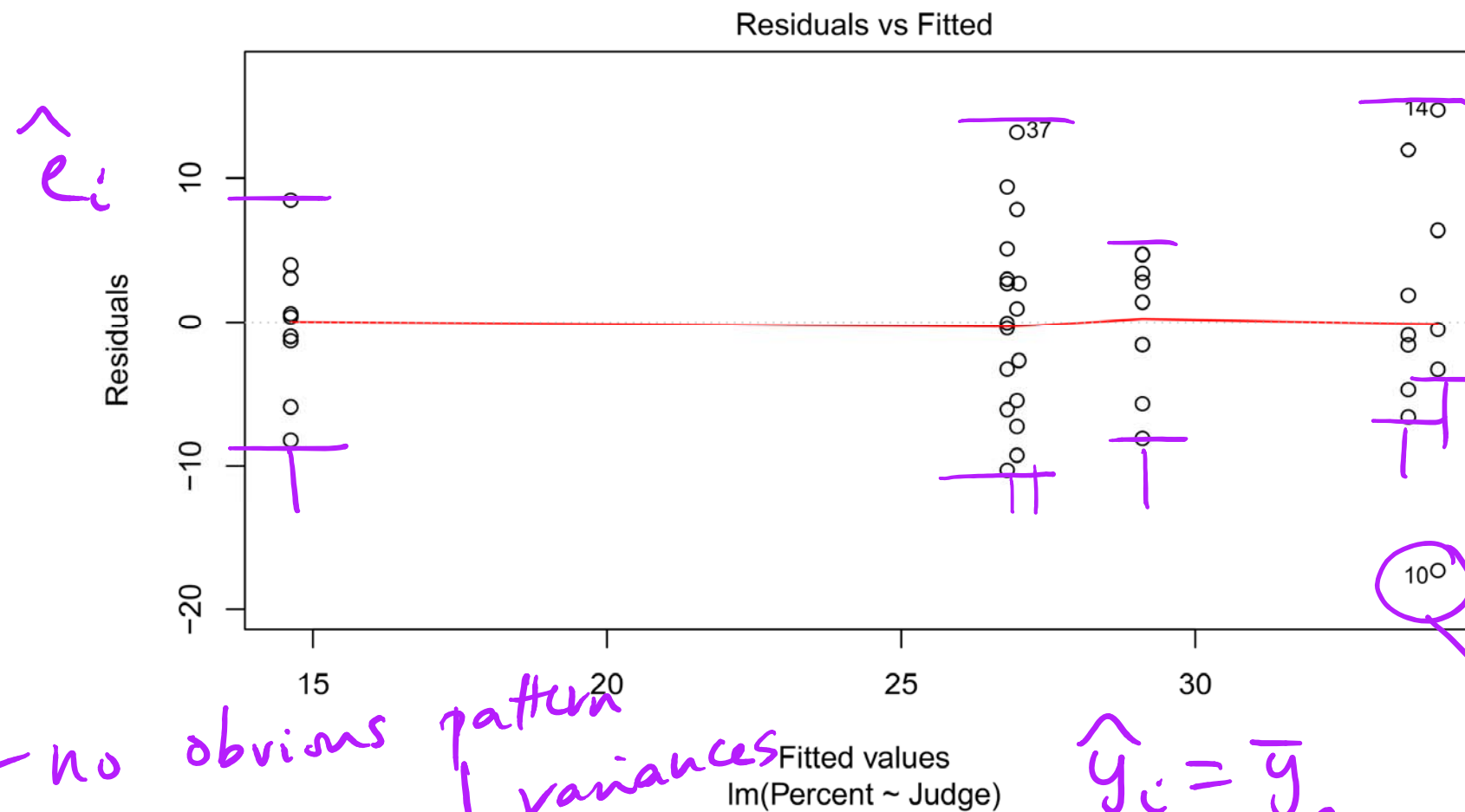
```
##
## Call:
## lm(formula = Percent ~ Judge)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -17.320  -4.367  -0.250   3.319  14.780
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   34.1200     3.0921  11.034 1.47e-13 ***
## JudgeB        -0.5033     4.1868  -0.120  0.9049
## JudgeC        -5.0200     3.8566  -1.302  0.2007
## JudgeD        -7.1200     5.7848  -1.231  0.2258
## JudgeE        -7.1533     4.1868  -1.709  0.0955 .
## JudgeF        -7.3200     3.8566  -1.898  0.0651 .
## JudgeSpock's -19.4978     3.8566  -5.056 1.05e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 6.914 on 39 degrees of freedom
```

$$\bar{y}_C = \frac{7(6)}{2} = 21.$$

$$y_g - \bar{y}_A$$

Check Normality: Linear Model

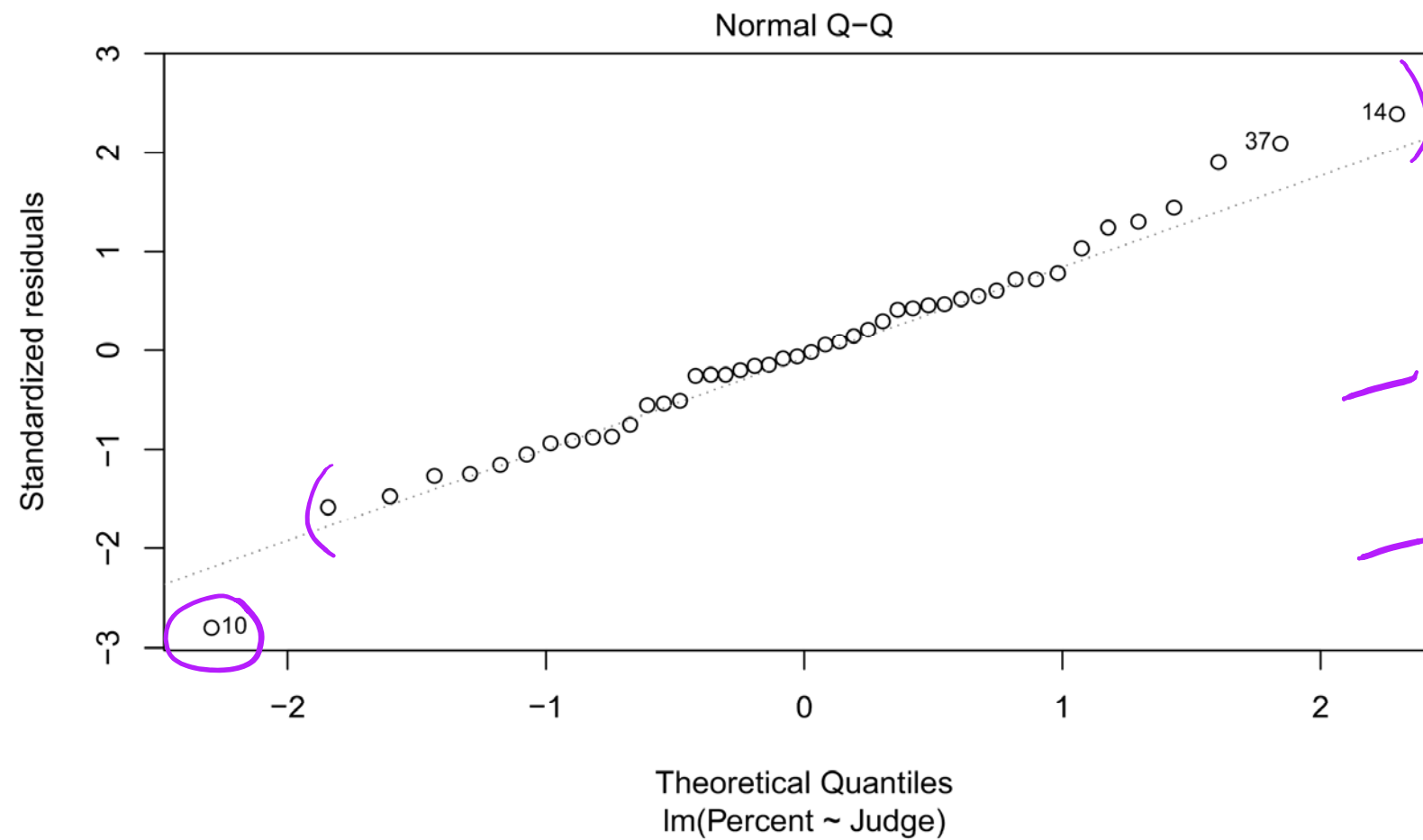
```
plot(lm(Percent ~ Judge), which=1)
```



large -ve residual

Check Normality: Linear Model

```
plot(lm(Percent ~ Judge), which=2)
```



Compare variances of all 7 judges: RoT

```
ssa=with(jury, tapply(Percent, Judge, sd))
ssa
```

```
##           A           B           C           D           E           F           Spock'
## 11.941817  6.582224  4.592929  3.818377  9.010142  5.968878  5.03879
```

```
isTRUE((max(ssa, na.rm=T)/min(ssa, na.rm=T))>2)
```

```
## [1] TRUE
```

$$\frac{11.94}{3.81} \approx 3 > 2.$$

Compare variances of all 7 judges: Bartlett's

```
bartlett.test(Percent~Judge, data=jury)
```

```
##
```

```
## Bartlett test of homogeneity of variances
```

```
##
```

```
## data: Percent by Judge
```

```
## Bartlett's K-squared = 7.7582, df = 6, p-value = 0.2564
```

Assume equal variance

$$S_p = \sqrt{MSE}$$

Case Study 1: Bonferroni's

```
Judge=relevel(Judge, ref="Spock's")
pairwise.t.test(Percent, Judge, p.adj="bonf")
```

$$k = G_2$$

$$k = \binom{7}{2}$$

$$N = 46$$

$$G = 7$$

```
##
## Pairwise comparisons using t tests with pooled SD
##
## data: Percent and Judge
##
## Spock's A B C D E
## A 0.00022 - - - -
## B 0.00013 1.00000 - - -
## C 0.00150 1.00000 1.00000 - -
## D 0.57777 1.00000 1.00000 1.00000 -
## E 0.03408 1.00000 1.00000 1.00000 1.00000
## F 0.01254 1.00000 1.00000 1.00000 1.00000 1.00000
##
## P value adjustment method: bonferroni
```

$$\alpha/k \text{ vs P-value}$$

$$\alpha \text{ vs } \min(k \times \text{P-value}, 1)$$

$$\textcircled{1} \begin{aligned} H_0: \mu_i - \mu_j &= 0 \\ H_a: \mu_i - \mu_j &\neq 0 \\ &\text{(2-sided)} \end{aligned}$$

$$\textcircled{2} t = \frac{\bar{x}_i - \bar{x}_j}{S_p \sqrt{\frac{1}{n_i} + \frac{1}{n_j}}}$$

$$\textcircled{3} \begin{aligned} \text{P-value} \\ &= 2 P(T_{n-9} > |t|) \end{aligned}$$

$$\textcircled{4} \text{Bonf. (Adj. P-value)} \\ k(\text{P-value})$$

Case Study 1: Bonferroni's CIs

```
lmod=lm(Percent~Judge)
nlevels(jury$Judge)
```

```
## [1] 7
```

```
confint(lmod, level=1-0.05/nlevels(jury$Judge))
```

```
##           0.357 % 99.643 %
## (Intercept) 8.078085 21.16636
## JudgeA      8.547341 30.44821
## JudgeB      8.647255 29.34163
## JudgeC      5.222970 23.73259
## JudgeD     -2.969585 27.72514
## JudgeE      1.997255 22.69163
## JudgeF      2.922970 21.43259
```

$$|\bar{y}_i - \bar{y}_j| \pm t_{\alpha/2, k} s_p \sqrt{\frac{1}{n_i} + \frac{1}{n_j}}$$

C.I for μ_{spocks}

Inc. 0

$$\mu_i - \mu_{spock} = \beta_i = 0$$

Case Study 1: Bonferroni's CIs

```
summary(lmod)
```

```
##
## Call:
## lm(formula = Percent ~ Judge)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -17.320  -4.367  -0.250   3.319  14.780
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   14.622     2.305   6.344 1.72e-07 ***
## JudgeA       19.498     3.857   5.056 1.05e-05 ***
## JudgeB       18.994     3.644   5.212 6.39e-06 ***
## JudgeC       14.478     3.259   4.442 7.15e-05 ***
## JudgeD       12.378     5.405   2.290 0.027513 *
## JudgeE       12.344     3.644   3.388 0.001623 **
## JudgeF       12.178     3.259   3.736 0.000597 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 6.914 on 39 degrees of freedom
```

Case Study 1: Tukey's CIs

```
amod=aov(Percent~Judge)
TukeyHSD(amod,"Judge")
```

```
## Tukey multiple comparisons of means
```

```
## 95% family-wise confidence level
```

```
##
```

```
## Fit: aov(formula = Percent ~ Judge)
```

```
##
```

```
## $Judge
```

```
##          diff          lwr          upr          p adj
```

```
## A-Spock's 19.49777778  7.514686 31.480870 0.0001992
```

```
## B-Spock's 18.99444444  7.671487 30.317402 0.0001224
```

```
## C-Spock's 14.47777778  4.350216 24.605339 0.0012936
```

```
## D-Spock's 12.37777778 -4.416883 29.172438 0.2744263
```

```
## E-Spock's 12.34444444  1.021487 23.667402 0.0248789
```

```
## F-Spock's 12.17777778  2.050216 22.305339 0.0098340
```

```
## B-A      -0.50333333 -13.512422 12.505755 0.9999997
```

```
## C-A      -5.02000000 -17.003092  6.963092 0.8470097
```

```
## D-A      -7.12000000 -25.094638 10.854638 0.8777485
```

```
## E-A      -7.15333333 -20.162422  5.855755 0.6146238
```

```
## F-A      -7.32000000 -19.303092  4.663092 0.4936379
```

```
## C-B      -4.51666667 -15.839625  6.806291 0.8742030
```

```
## D-B      -6.61666667 -24.158118 10.924781 0.0003280
```

P-value

- C.I includes 0
• P-value $> (\alpha = 0.05)$

Case Study 1: Tukey's CI's

```
plot(TukeyHSD(amod, "Judge"))
```

