Additional Hypothesis Test Applications Solutions

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1. HFI

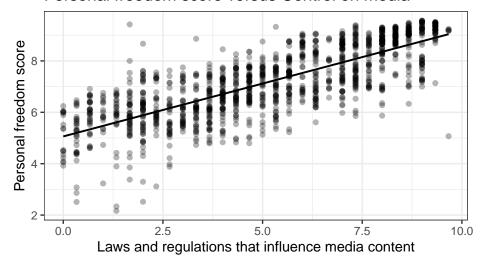
Choose another freedom variable and a variable you think would strongly correlate with it. The openintro package contains the data set hfi. Type ?openintro::hfi in the Console window in RStudio to learn more about the variables.

```
hfi<-tibble(read_csv("data/hfi.csv"))
```

a. Produce a scatterplot of the two variables.

We selected pf_expression_influence as it is a measure of laws and regulations that influence media content. We kept pf_score because it is a measure of personal freedom in a country. Our thought is these should be correlated.

Personal freedom score versus Control on media



b. Quantify the strength of the relationship with the correlation coefficient.

```
hfi %>%
  summarise(cor(pf_expression_influence, pf_score, use = "complete.obs"))
## # A tibble: 1 x 1
     'cor(pf_expression_influence, pf_score, use = "complete.obs")'
##
                                                                <dbl>
## 1
                                                                0.787
```

c. Fit a linear model. At a glance, does there seem to be a linear relationship?

```
m2 <- lm(pf_score ~ pf_expression_influence, data = hfi)</pre>
```

```
summary(m2)
```

```
##
## Call:
## lm(formula = pf_score ~ pf_expression_influence, data = hfi)
##
## Residuals:
##
      Min
                1Q Median
                                3Q
                                       Max
  -3.9688 -0.5830 0.1681 0.5903 3.6730
##
## Coefficients:
                           Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                            5.06135
                                       0.05064
                                                 99.95
                                                         <2e-16 ***
## pf_expression_influence 0.41150
                                       0.00869
                                                 47.36
                                                         <2e-16 ***
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
##
## Residual standard error: 0.8482 on 1376 degrees of freedom
     (80 observations deleted due to missingness)
## Multiple R-squared: 0.6197, Adjusted R-squared: 0.6195
## F-statistic: 2243 on 1 and 1376 DF, p-value: < 2.2e-16
```

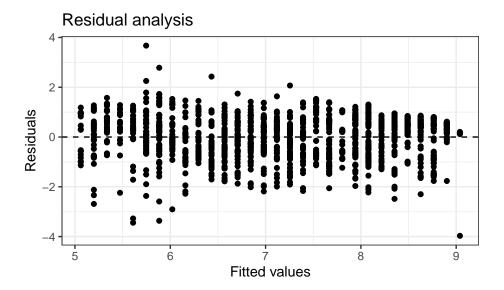
d. How does this relationship compare to the relationship between pf_expression_control and pf score? Use the R^2 values from the two model summaries to compare. Does your independent variable seem to predict your dependent one better? Why or why not?

The adjuste R^2 is a little smaller so the fit is not as good.

e. Display the model diagnostics for the regression model analyzing this relationship.

Linearity:

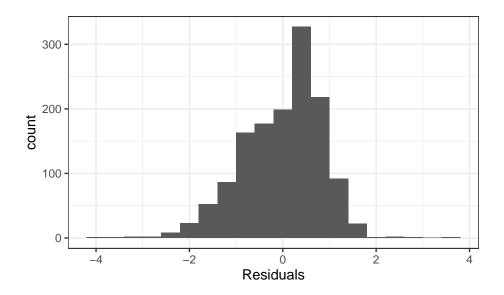
```
ggplot(data = m2, aes(x = .fitted, y = .resid)) +
  geom_point() +
  geom_hline(yintercept = 0, linetype = "dashed") +
  labs(x="Fitted values",y="Residuals",title="Residual analysis") +
  theme_bw()
```



There does appear to be some type of fluctuation but so the linear model may not be appropriate.

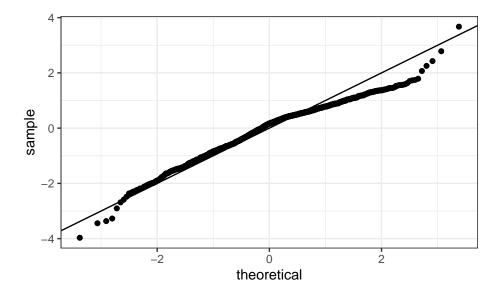
Nearly normal residuals:

```
ggplot(data = m2, aes(x = .resid)) +
  geom_histogram(binwidth = .4) +
  xlab("Residuals") +
  theme_bw()
```



or a normal probability plot of the residuals.

```
ggplot(data = m2, aes(sample = .resid)) +
  stat_qq() +
  theme_bw() +
  geom_abline(slope=1,intercept = 0)
```



No, the sample is small but it appears the residual are skewed to the left.

Constant variability:

Exercise: Based on the residuals vs. fitted plot, does the constant variability condition appear to be met?

Yes, the width of the plot seems constant with the exception of some extreme points.

f. Predict the response from your explanatory variable for a value between the median and third quartile. Is this an overestimate or an underestimate, and by how much?

```
## Min. 1st Qu. Median Mean 3rd Qu. Max. NA's
## 0.000 3.000 5.333 5.200 7.333 9.667 80

predict(m2,newdata=data.frame(pf_expression_influence=6))

## 1
## 7.53036
```

We thus predict a value of 7.53 for the pf_score.

The observed value is 7.96, an average of 42 data points. We tend to underestimate the observed value.

```
library(broom)
```

```
augment(m2) %>%
  filter(pf_expression_influence==6) %>%
   summarize(ave=mean(pf_score),n=n())
```

```
## # A tibble: 1 x 2
## ave n
## <dbl> <int>
## 1 7.96 42
```

File Creation Information

• File creation date: 2020-10-30

 \bullet Windows version: Windows 10 x64 (build 18362)

R version 3.6.3 (2020-02-29)
mosaic package version: 1.7.0
tidyverse package version: 1.3.0
openintro package version: 2.0.0
broom package version: 0.7.0