# R Notebook

## Loading of Data

```
library(tidyverse)
beauty <- read_csv("beauty.csv")</pre>
Filtering to only female respondents:
beauty_data <- beauty %>% filter(gender == "Female")
Demographic characteristics:
fivenum(beauty_data$age)
## [1] 18 20 20 21 30
unique(beauty_data$faculty)
## [1] "Business" "CHS"
                               "CDE"
                                          "Law"
                                                      "SOC"
                                                                  "Medicine"
fivenum(beauty_data$tiktok_hours)
## [1] 0.0 0.5 1.5 2.0 6.0
beauty_data %>% mutate(beauty_index = tiktok_hours*frequency) -> beauty_data
```

#### T-tests between IV and DVs

First, split respondents into two groups: those who do not consume beauty content on Tiktok (ie. Watched Hours = 0 + Tiktoks related to beauty answer "not at all"), and those who do.

```
beauty_false <- beauty_data %>% filter(tiktok_hours == 0 | frequency == 1)
beauty_true <- setdiff(beauty_data, beauty_false)</pre>
```

Then, compare means of dependent variables.

For self-esteem, our hypothesis is that the self-esteems of those who consume beauty content are lower than those who do not. t=1.2579, p=0.8907. Hence, the longer-term self-esteems of those who consume beauty content is not significantly lesser than those who do not.

```
t.test(beauty_true$high_self_esteem, beauty_false$high_self_esteem, alternative = "less")
```

For appearance anxiety, our hypothesis is that those who consume beauty content experience more appearance anxiety than those who do not. t=1.0299, , p-value = 0.156. Hence, the longer-term appearance anxiety of those who consume beauty content is not significantly greater than those who do not.

```
t.test(beauty_true$anxiety, beauty_false$anxiety, alternative = "greater")
```

For mood, our hypothesis is that those who consume beauty content have their moods more negatively affected by such content than those who do not. t = 2.9167, p-value = 0.003428. Hence, the moods of those who consume beauty content are significantly more negatively affected than those who do not.

```
t.test(beauty_true$negative_mood, beauty_false$negative_mood, alternative = "greater")
```

## Simple Linear Regression between IV and DVs

Beauty Index on Mood: correlation is estimated to be 0.11257, p-value is 0.0003816, and there is a significant positive correlation.

```
beauty_mood <- lm(negative_mood ~ beauty_index, data = beauty_data)
summary(beauty_mood)</pre>
```

```
##
## Call:
## lm(formula = negative_mood ~ beauty_index, data = beauty_data)
##
## Residuals:
##
      Min
               1Q Median
                               3Q
                                      Max
## -3.2078 -1.3990 -0.2912 1.2431 3.7088
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                2.85986
                           0.24223 11.806 < 2e-16 ***
## beauty_index 0.10784
                           0.03372
                                    3.198 0.00185 **
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.685 on 100 degrees of freedom
                                   Adjusted R-squared:
## Multiple R-squared: 0.09279,
## F-statistic: 10.23 on 1 and 100 DF, p-value: 0.001853
```

Beauty Index on Anxiety: correlation is estimated to be 0.04991, p-value is 0.0137, and there is a non-significant positive correlation.

```
beauty_anxiety <- lm(anxiety ~ beauty_index, data = beauty_data)
summary(beauty_anxiety)</pre>
```

```
##
## Call:
## lm(formula = anxiety ~ beauty_index, data = beauty_data)
##
## Residuals:
##
       Min
                 1Q
                      Median
                                   3Q
                                           Max
## -2.35412 -0.82380 0.07101 0.87049
##
## Coefficients:
##
                Estimate Std. Error t value Pr(>|t|)
                2.73021
                           0.14289 19.107
## (Intercept)
                                             <2e-16 ***
## beauty_index 0.04991
                           0.01989
                                     2.509
                                             0.0137 *
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.9938 on 100 degrees of freedom
## Multiple R-squared: 0.05924,
                                   Adjusted R-squared:
## F-statistic: 6.297 on 1 and 100 DF, p-value: 0.0137
```

Beauty Index on Self-esteem: correlation is estimated to be 0.1155, p-value is 0.592, and there is a extremely non-significant positive correlation.

```
beauty_esteem <- lm(high_self_esteem ~ beauty_index, data = beauty_data)
summary(beauty_esteem)</pre>
```

```
##
## Call:
## lm(formula = high_self_esteem ~ beauty_index, data = beauty_data)
## Residuals:
##
      Min
               10 Median
                                3Q
                                       Max
  -2.2994 -0.9223 -0.1261
                                   1.8623
                           0.8277
##
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
                                    20.264
## (Intercept)
                3.12612
                            0.15427
                                              <2e-16 ***
## beauty_index 0.01155
                            0.02147
                                     0.538
                                               0.592
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 1.073 on 100 degrees of freedom
## Multiple R-squared: 0.002884,
                                   Adjusted R-squared:
## F-statistic: 0.2892 on 1 and 100 DF, p-value: 0.5919
```

### Confounders

We have identified consumption of beauty content from other platforms outside of Tiktok as a possible confounder for the correlations between our IV and DVs. Our confounder hence becomes the variable of the number of other platforms used outside of Tiktok, found through mutating a new column from the data in beauty\_data\$platforms.

```
mutate(beauty_data, other_platforms = str_remove_all(beauty_data$platforms, "Tiktok/ Douyin;")) -> beau
mutate(beauty_data, other_platforms = str_remove_all(beauty_data$other_platforms, "Tiktok/ Douyin")) ->
mutate(beauty_data, other_platforms = str_count(beauty_data$other_platforms, ";")) -> beauty_data
```

The effect of consuming beauty Tiktok content on appearance anxiety, when controlled for consumption of content from other platforms, shows no change in terms of significance. Estimated correlation = 0.05009, p = 0.04532.

```
beauty_anxiety_conf <- lm(anxiety ~ beauty_index + other_platforms, data = beauty_data)
summary(beauty_anxiety_conf)</pre>
```

```
##
## lm(formula = anxiety ~ beauty_index + other_platforms, data = beauty_data)
##
## Residuals:
       Min
                  1Q
                       Median
                                    3Q
                                            Max
## -2.35195 -0.81773 0.06876 0.87344
##
## Coefficients:
                   Estimate Std. Error t value Pr(>|t|)
                               0.21959 12.147
                    2.66747
                                                 <2e-16 ***
## (Intercept)
```

```
0.05009
                               0.01998
                                         2.507
                                                 0.0138 *
## beauty index
                                                 0.7066
## other_platforms
                   0.05840
                               0.15471
                                         0.377
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.9981 on 99 degrees of freedom
## Multiple R-squared: 0.06059,
                                    Adjusted R-squared:
## F-statistic: 3.193 on 2 and 99 DF, p-value: 0.04532
```

However, when examining the effect of consuming beauty Tiktok content on self-esteem, when controlled for consumption of content from other platforms, shows that while beauty Tiktok content does not have a significant effect, consumption of content from other platforms indeed has a significant effect at a p = 0.05 level.

```
beauty_esteem_conf <- lm(high_self_esteem ~ beauty_index + other_platforms, data = beauty_data)
summary(beauty_esteem_conf)</pre>
```

```
##
## Call:
## lm(formula = high_self_esteem ~ beauty_index + other_platforms,
##
       data = beauty_data)
##
## Residuals:
##
      Min
                10 Median
                                3Q
                                       Max
##
  -2.6716 -0.7129 0.1972
                           0.8263
                                    1.8339
##
## Coefficients:
##
                   Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                    2.71396
                               0.23085 11.756
                                                 <2e-16 ***
## beauty_index
                    0.01269
                               0.02101
                                         0.604
                                                 0.5470
                                         2.359
                                                 0.0203 *
## other_platforms 0.38363
                               0.16264
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.049 on 99 degrees of freedom
## Multiple R-squared: 0.05594,
                                    Adjusted R-squared:
## F-statistic: 2.933 on 2 and 99 DF, p-value: 0.05787
```

We do not repeat the test for confounders on the single linear regression model between beauty Tiktok consumption and mood because the question for that set of observations surveys for a cause and effect relationship between Tiktok consumption specifically and mood ("negative changes in your mood (e.g., increased sadness or self consciousness) *after* watching beauty TikTok video") for non-neutral answers.

#### Conclusions from t-tests and linear regressions: