Final Project

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```
tiktok_data <- read.csv("tiktok4.csv")</pre>
tiktok_observations <- nrow(tiktok_data)</pre>
tiktok_variables <- ncol(tiktok_data)</pre>
print(paste("number of observations:", tiktok_observations))
## [1] "number of observations: 950"
print(paste("number of variables:", tiktok_variables))
## [1] "number of variables: 8"
print(names(tiktok_data))
## [1] "Rank"
                         "Tiktoker.name" "Tiktok.name"
                                                            "followers"
## [5] "views.avg."
                         "likes.avg.."
                                          "comments.avg.." "shares.avg.."
# Function to convert values with 'M' and 'K' suffixes to numeric
convert_followers <- function(x) {</pre>
  # Remove 'M' and 'K' suffixes and convert to numeric
  num <- as.numeric(sub("[^0-9.]", "", x))</pre>
  # Multiply by the necessary value to get the number
  if (endsWith(x, "M")) {
   num <- num * 1e6
  } else if (endsWith(x, "K")) {
    num <- num * 1e3
 }
 num
}
tiktok_data$followers_numeric <- sapply(tiktok_data$followers, convert_followers)
tiktok_data$views_numeric <- sapply(tiktok_data$views.avg., convert_followers)
tiktok_data$likes_numeric <- sapply(tiktok_data$likes.avg.., convert_followers)</pre>
tiktok_data$comments_numeric <- sapply(tiktok_data$comments.avg.., convert_followers)
tiktok_data$shares_numeric <- sapply(tiktok_data$shares.avg.., convert_followers)
library(dplyr)
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##
       filter, lag
```

```
##
##
       intersect, setdiff, setequal, union
sorted_tiktok <- arrange(tiktok_data, desc(views_numeric))</pre>
head(sorted_tiktok, 10)
##
      Rank
               Tiktoker.name
                                   Tiktok.name followers views.avg. likes.avg..
## 1
        27 nicollefigueroaa Nicolle Figueroa
                                                     15.5M
                                                                               1.6M
                                                                   13M
## 2
                                   Fadil Jaidi
                                                      9.2M
                                                                 12.7M
                                                                               1.7M
                  fadiljaidi
## 3
                                                     19.9M
                                                                 11.5M
                                                                               1.3M
        26
                 surthycooks
                                   Surthycooks
## 4
                                   Lexi Rivera
        24 lexibrookerivera
                                                     24.8M
                                                                 11.3M
                                                                               1.5M
## 5
        25
                hotspanishmx
                                    HotSpanish
                                                     11.9M
                                                                 11.2M
                                                                                 1M
## 6
        34
             bayashi.tiktok
                                 Bayashi
                                              38.8M
                                                          10.9M
                                                                        1.1M
## 7
                                  slaterkodish
                slaterkodish
                                                                               1.3M
        33
                                                      1.6M
                                                                 10.8M
## 8
        19
                     rosalia
                                    La Rosalia
                                                     28.1M
                                                                 10.6M
                                                                               1.2M
## 9
        49
                  mmmjoemele
                                       Joe Mele
                                                     24.1M
                                                                 10.6M
                                                                                 1M
                                                                            909.6K
## 10
        62
             bigchungus.tik
                                    BigChungus
                                                      9.2M
                                                                 10.3M
##
      comments.avg.. shares.avg.. followers_numeric views_numeric likes_numeric
## 1
                 4.2K
                               1.3K
                                              15500000
                                                             13000000
                                                                              1600000
## 2
                 7.6K
                               4.4K
                                               9200000
                                                             12700000
                                                                              1700000
## 3
                               1.4K
                 6.3K
                                              19900000
                                                             11500000
                                                                              1300000
## 4
                 5.6K
                               2.4K
                                              24800000
                                                             11300000
                                                                              1500000
## 5
                 6.1K
                               5.9K
                                              11900000
                                                             11200000
                                                                              1000000
## 6
                               2.5K
                                              38800000
                                                                              1100000
                   4K
                                                             10900000
## 7
                               3.9K
                 1.9K
                                               1600000
                                                             10800000
                                                                              1300000
## 8
                 6.1K
                               8.5K
                                              28100000
                                                             10600000
                                                                              1200000
## 9
                                                                              1000000
                 2.4K
                               2.1K
                                              24100000
                                                             10600000
## 10
                 2.5K
                               1.3K
                                               9200000
                                                             10300000
                                                                               909600
##
      comments_numeric shares_numeric
## 1
                   4200
                                   1300
## 2
                   7600
                                   4400
## 3
                   6300
                                   1400
## 4
                   5600
                                   2400
## 5
                                   5900
                   6100
## 6
                   4000
                                   2500
## 7
                   1900
                                   3900
## 8
                   6100
                                   8500
## 9
                                   2100
                   2400
## 10
                   2500
                                   1300
wanted_columns <- c("followers_numeric", "views_numeric", "likes_numeric", "comments_numeric", "shares_</pre>
tiktok_subset <- tiktok_data[wanted_columns]</pre>
pairs(tiktok_subset)
```

The following objects are masked from 'package:base':

```
2.0e+06 1.0e+07
                                                  0 4000 10000
                                                                                 8e+07
    followers_numeric
                                                                                 0e+00
                     views numeric
                                     likes_numeric
                                                   comments_numeric
                                                                    shares_numeric
          6e+07
                                       1000000
                                                                     15000
 0e+00
                                  0
                                                                             35000
m1 <- lm(views_numeric ~ followers_numeric + likes_numeric + comments_numeric + shares_numeric, data = '
summary(m1)
##
## Call:
## lm(formula = views_numeric ~ followers_numeric + likes_numeric +
       comments_numeric + shares_numeric, data = tiktok_subset)
##
## Residuals:
        Min
##
                  1Q
                       Median
                                     3Q
                                             Max
  -3065842 -710505 -221989
                                 469605 5557204
##
## Coefficients:
##
                       Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                      5.706e+05 7.376e+04
                                              7.736 2.63e-14 ***
## followers_numeric 1.895e-02
                                  3.711e-03
                                              5.107 3.96e-07 ***
                      6.516e+00
                                  1.904e-01 34.231
                                                     < 2e-16 ***
## likes_numeric
## comments_numeric -4.001e+00
                                  2.234e+01
                                             -0.179
                                                        0.858
## shares_numeric
                      7.876e+00 1.119e+01
                                              0.704
                                                        0.482
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 1074000 on 945 degrees of freedom
## Multiple R-squared: 0.6042, Adjusted R-squared: 0.6025
## F-statistic: 360.6 on 4 and 945 DF, p-value: < 2.2e-16
anova(m1)
## Analysis of Variance Table
## Response: views_numeric
```

```
##
                                Sum Sa
                                           Mean Sq
                                                      F value Pr(>F)
## followers_numeric
                         1 2.2028e+14 2.2028e+14
                                                     191.0690 <2e-16 ***
                         1 1.4422e+15 1.4422e+15 1250.9967 <2e-16 ***
## likes numeric
## comments_numeric
                         1 1.0229e+09 1.0229e+09
                                                       0.0009 0.9762
## shares_numeric
                         1 5.7154e+11 5.7154e+11
                                                        0.4958 0.4815
## Residuals
                       945 1.0895e+15 1.1529e+12
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
par(mfrow = c(2,2))
plot(m1)
                                                  Standardized residuals
                                                                     Q-Q Residuals
                Residuals vs Fitted
     90+99
                                                       9
           O930
Residuals
                                                       \alpha
     -4e+06
                                                       7
           2.0e+06
                      6.0e+06
                                                                    2
                                                                                      2
                                                                                           3
                                  1.0e+07
                     Fitted values
                                                                   Theoretical Quantiles
(Standardized residuals)
                                                  Standardized residuals
                  Scale-Location
                                                                 Residuals vs Leverage
                                                       9
                                                                                               0.5
     1.5
                                        0
                                                       Ņ
     0.0
                                                                          distance
           2.0e+06
                      6.0e+06
                                  1.0e+07
                                                           0.00
                                                                       0.04
                                                                                   0.08
                     Fitted values
                                                                         Leverage
library(car)
## Loading required package: carData
##
## Attaching package: 'car'
## The following object is masked from 'package:dplyr':
##
##
       recode
tiktok_subset$comments_numeric[tiktok_subset$comments_numeric == 0] <- 1</pre>
summary(tranxy <- powerTransform(cbind(views_numeric, followers_numeric, likes_numeric,</pre>
                                           comments_numeric, shares_numeric) ~ 1, data = tiktok_subset))
## bcPower Transformations to Multinormality
##
                       Est Power Rounded Pwr Wald Lwr Bnd Wald Upr Bnd
## views_numeric
                         -0.0887
                                          0.00
                                                     -0.1782
                                                                     0.0008
                           0.1855
                                          0.19
                                                      0.1493
                                                                     0.2218
## followers_numeric
## likes_numeric
                                          0.15
                                                      0.0689
                                                                     0.2215
                           0.1452
```

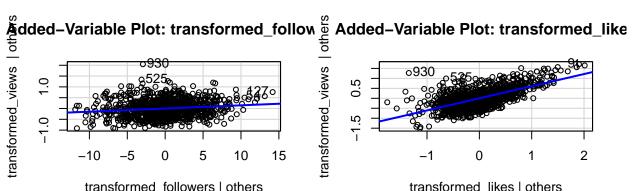
```
0.1925
                                      0.19
                                                 0.1534
                                                               0.2316
## comments numeric
                       -0.0760
                                     -0.08
                                                -0.1277
                                                             -0.0243
## shares_numeric
##
## Likelihood ratio test that transformation parameters are equal to 0
##
    (all log transformations)
##
                                      LRT df
                                                   pval
## LR test, lambda = (0 0 0 0 0) 269.7237 5 < 2.22e-16
## Likelihood ratio test that no transformations are needed
##
                                      LRT df
## LR test, lambda = (1 1 1 1 1) 4817.547
                                           5 < 2.22e-16
tiktok_subset$transformed_followers <- (tiktok_subset$followers_numeric)^(0.19)
tiktok_subset$transformed_likes <- (tiktok_subset$likes_numeric)^(0.15)</pre>
tiktok_subset$transformed_comments <- (tiktok_subset$comments_numeric)^(0.19)
tiktok_subset$transformed_shares <- (tiktok_subset$shares_numeric)^(-0.08)
tiktok_subset$transformed_views <- log(tiktok_subset$views_numeric)</pre>
m2 <- lm(transformed_views ~ transformed_followers +</pre>
                    transformed_likes + transformed_shares + transformed_comments, data = tiktok_subset
summary(m2)
##
## Call:
## lm(formula = transformed_views ~ transformed_followers + transformed_likes +
       transformed_shares + transformed_comments, data = tiktok_subset)
## Residuals:
        Min
                       Median
                  1Q
                                    30
                                            Max
## -0.88112 -0.29365 -0.02945 0.26185
                                       2.08864
##
## Coefficients:
##
                          Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                                     0.233258 43.282 < 2e-16 ***
                         10.095899
## transformed_followers 0.014737
                                     0.003250
                                                4.535 6.51e-06 ***
## transformed_likes
                          0.611160
                                     0.024904 24.541 < 2e-16 ***
                                                        0.0403 *
## transformed_shares
                          0.614070
                                     0.299008
                                                2.054
## transformed_comments -0.008977
                                     0.023663 -0.379
                                                        0.7045
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.3993 on 945 degrees of freedom
## Multiple R-squared: 0.4629, Adjusted R-squared: 0.4606
## F-statistic: 203.6 on 4 and 945 DF, p-value: < 2.2e-16
anova(m2)
## Analysis of Variance Table
## Response: transformed_views
                          Df Sum Sq Mean Sq F value Pr(>F)
                           1 27.951 27.951 175.2730 < 2e-16 ***
## transformed followers
## transformed likes
                           1 101.079 101.079 633.8397 < 2e-16 ***
## transformed shares
                           1
                               0.804
                                       0.804
                                               5.0412 0.02498 *
## transformed_comments
                               0.023
                                       0.023
                                               0.1439 0.70452
                           1
## Residuals
                         945 150.699
                                       0.159
```

```
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
tt_model_empty <- lm(transformed_views ~ 1, data = tiktok_subset)</pre>
tt_forward_model <- step(tt_model_empty, scope = transformed_views ~ transformed_followers + transforme
                         + transformed_comments + transformed_shares, data = tiktok_subset, direction =
## Start: AIC=-1156.71
## transformed_views ~ 1
##
##
                           Df Sum of Sq
                                           RSS
                           1 124.547 156.01 -1712.2
## + transformed_likes
## + transformed_followers 1
                              27.951 252.60 -1254.4
## + transformed_comments
                           1
                                5.464 275.09 -1173.4
## + transformed_shares
                            1
                                 4.578 275.98 -1170.3
## <none>
                                        280.56 -1156.7
##
## Step: AIC=-1712.22
## transformed_views ~ transformed_likes
##
##
                           Df Sum of Sq
                                           RSS
                                                   AIC
## + transformed_followers 1
                                4.4824 151.53 -1737.9
## + transformed_shares
                            1
                                 2.0286 153.98 -1722.7
## <none>
                                        156.01 -1712.2
                                0.1752 155.83 -1711.3
## + transformed_comments
                          1
## Step: AIC=-1737.92
## transformed_views ~ transformed_likes + transformed_followers
##
##
                          Df Sum of Sq
                                          RSS
## + transformed_shares
                              0.80393 150.72 -1741.0
## <none>
                                       151.53 -1737.9
## + transformed_comments 1 0.15429 151.37 -1736.9
##
## Step: AIC=-1740.97
## transformed_views ~ transformed_likes + transformed_followers +
##
       transformed_shares
##
##
                          Df Sum of Sq
                                          RSS
                                                  AIC
## <none>
                                       150.72 -1741.0
## + transformed_comments 1 0.022948 150.70 -1739.1
m3 <- lm(transformed_views ~ transformed_followers +</pre>
                    transformed_likes + transformed_shares, data = tiktok_subset)
summary(m3)
##
## Call:
## lm(formula = transformed_views ~ transformed_followers + transformed_likes +
       transformed shares, data = tiktok subset)
##
## Residuals:
       Min
##
                  1Q
                     Median
                                    3Q
                                            Max
## -0.87867 -0.29251 -0.03013 0.26080 2.09035
##
```

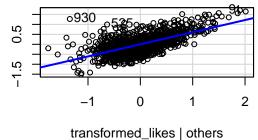
```
## Coefficients:
##
                            Estimate Std. Error t value Pr(>|t|)
                                        0.211779
## (Intercept)
                           10.058889
                                                   47.497 < 2e-16 ***
## transformed_followers 0.014659
                                        0.003242
                                                    4.522 6.91e-06 ***
## transformed_likes
                            0.608857
                                        0.024141
                                                   25.221
                                                            < 2e-16 ***
                                                             0.0249 *
  transformed shares
                            0.645339
                                        0.287292
                                                    2.246
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.3992 on 946 degrees of freedom
## Multiple R-squared: 0.4628, Adjusted R-squared: 0.4611
## F-statistic: 271.6 on 3 and 946 DF, p-value: < 2.2e-16
anova(m3)
## Analysis of Variance Table
##
## Response: transformed_views
##
                            Df Sum Sq Mean Sq F value Pr(>F)
## transformed_followers
                             1
                                27.951 27.951 175.4318 < 2e-16 ***
## transformed_likes
                             1 101.079 101.079 634.4139 < 2e-16 ***
## transformed_shares
                                  0.804
                                          0.804
                                                   5.0458 0.02492 *
                             1
## Residuals
                           946 150.722
                                          0.159
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
par(mfrow = c(2,2))
plot(m3)
                                                 Standardized residuals
                                                                   Q-Q Residuals
                Residuals vs Fitted
                                                      9
              0930
                                                                                        9300
Residuals
     1.0
                                                      \alpha
     -1.0
                                                      7
            13.5
                       14.5
                                   15.5
                                                            -3
                                                                  2
                                                                           0
                                                                                     2
                                                                                         3
                     Fitted values
                                                                  Theoretical Quantiles
Standardized residuals
                                                 Standardized residuals
                  Scale-Location
                                                               Residuals vs Leverage
                                                      9
                                                                         9300
     1.5
                                                      \alpha
     0.0
                                                                                 0.015
            13.5
                       14.5
                                   15.5
                                                          0.000
                                                                 0.005
                                                                         0.010
                                                                                         0.020
                     Fitted values
                                                                        Leverage
par(mfrow=c(2,2))
StanRes2 <- rstandard(m3)</pre>
```

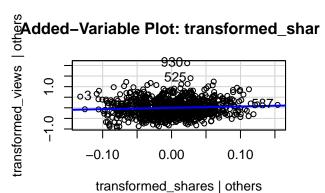
```
plot(tiktok_subset$transformed_followers,StanRes2,ylab="Standardized Residuals")
plot(tiktok_subset$transformed_likes,StanRes2,ylab="Standardized Residuals")
plot(tiktok_subset$transformed_shares,StanRes2,ylab="Standardized Residuals")
plot(m3\fitted.values,StanRes2,ylab="Standardized Residuals",xlab="Fitted Values")
Standardized Residuals
                                                     Standardized Residuals
                                  25
                                        30
                                                                    5
                                                                            6
                                                                                            8
                      15
                            20
          tiktok_subset$transformed_followers
                                                                 tiktok_subset$transformed_likes
Standardized Residuals
                                                     Standardized Residuals
                                                                 13.5
           0.45
                        0.55
                                    0.65
                                                                              14.5
                                                                                          15.5
           tiktok_subset$transformed_shares
                                                                           Fitted Values
library(car)
par(mfrow=c(2,2))
avPlot(m3, variable = "transformed_followers", ask=FALSE)
avPlot(m3, variable = "transformed_likes", ask=FALSE)
```

avPlot(m3, variable = "transformed_shares", ask=FALSE)



transformed_followers | others





vif(m3)

transformed_followers transformed_likes transformed_shares ## 1.175534 1.093710 1.079565