Tobit v.s. Quantile Analysis

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Regressions and Correlation Matrices

```
tobit_bias_regression <- lm(Coefficient ~ Alpha + Omega + Cutoff, data = tobit)
summary(tobit_bias_regression)
##
## Call:
## lm(formula = Coefficient ~ Alpha + Omega + Cutoff, data = tobit)
## Residuals:
##
       Min
                  1Q
                     Median
                                   3Q
                                           Max
## -1.68108 -0.09257 0.00007 0.09167 1.46015
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
                         0.003904 128.813
## (Intercept) 0.502908
                                             <2e-16 ***
## Alpha
              -0.009167
                          0.000235 -39.008
                                             <2e-16 ***
## Omega
               0.000299
                          0.000235
                                    1.272
                                              0.203
              -0.001002
                          0.004700 -0.213
                                              0.831
## Cutoff
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.1946 on 171496 degrees of freedom
## Multiple R-squared: 0.008804,
                                   Adjusted R-squared: 0.008787
## F-statistic: 507.8 on 3 and 171496 DF, p-value: < 2.2e-16
tobit_mse_regression <- lm((Coefficient - .5)^2 ~ Alpha + Omega + Cutoff, data = tobit)
summary(tobit_mse_regression)
##
## Call:
## lm(formula = (Coefficient - 0.5)^2 ~ Alpha + Omega + Cutoff,
      data = tobit)
##
## Residuals:
##
       Min
                 1Q
                      Median
                                   3Q
                                           Max
## -0.09146 -0.03535 -0.00868 0.00997 2.74345
##
## Coefficients:
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) -8.494e-03 1.525e-03 -5.571 2.53e-08 ***
              -1.773e-03 9.177e-05 -19.324 < 2e-16 ***
## Alpha
## Omega
               1.511e-02 9.177e-05 164.658 < 2e-16 ***
## Cutoff
              -1.715e-02 1.835e-03 -9.342 < 2e-16 ***
## ---
```

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

```
##
## Residual standard error: 0.07601 on 171496 degrees of freedom
## Multiple R-squared: 0.1385, Adjusted R-squared: 0.1385
## F-statistic: 9191 on 3 and 171496 DF, p-value: < 2.2e-16
quant_bias_regression <- lm(Coefficient ~ Alpha + Omega + Cutoff, data = quantile)
summary(quant_bias_regression)
##
## Call:
## lm(formula = Coefficient ~ Alpha + Omega + Cutoff, data = quantile)
## Residuals:
##
       Min
                 1Q
                      Median
                                   3Q
## -20.1601 -0.7958 -0.0005
                               0.8047
                                       24.0574
## Coefficients:
##
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) 4.760e-01 3.942e-02 12.074
                                              <2e-16 ***
## Alpha
               1.418e-05 2.373e-03 0.006
                                               0.995
              -1.235e-03 2.373e-03 -0.520
## Omega
                                               0.603
               3.675e-02 4.745e-02
## Cutoff
                                     0.774
                                               0.439
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 1.965 on 171496 degrees of freedom
## Multiple R-squared: 5.076e-06, Adjusted R-squared: -1.242e-05
## F-statistic: 0.2902 on 3 and 171496 DF, p-value: 0.8325
quant_mse_regression <- lm((Coefficient - .5)^2 ~ Alpha + Omega + Cutoff, data = quantile)
summary(quant_mse_regression)
##
## Call:
## lm(formula = (Coefficient - 0.5)^2 ~ Alpha + Omega + Cutoff,
##
      data = quantile)
## Residuals:
     Min
             1Q Median
                           3Q
                                 Max
## -10.96 -3.86 -1.32
                        1.09 570.33
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) -2.44644
                          0.19141 - 12.781
                                            <2e-16 ***
              -0.80995
                          0.01152 -70.299
                                            <2e-16 ***
## Alpha
## Omega
               1.55014
                          0.01152 134.544
                                            <2e-16 ***
## Cutoff
               0.13493
                          0.23043
                                    0.586
                                             0.558
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 9.543 on 171496 degrees of freedom
## Multiple R-squared: 0.1185, Adjusted R-squared: 0.1184
## F-statistic: 7681 on 3 and 171496 DF, p-value: < 2.2e-16
```

kable(cor(quantile))

	Alpha	Omega	Cutoff	Coefficient
Alpha	1.00e+00	0.0000000	0.0000000	0.0000144
Omega	0.00e+00	1.0000000	0.0000000	-0.0012564
Cutoff	0.00e+00	0.0000000	1.0000000	0.0018701
Coefficient	1.44e-05	-0.0012564	0.0018701	1.0000000

kable(cor(tobit))

	Alpha	Omega	Cutoff	Coefficient
Alpha	1.0000000	0.0000000	0.0000000	-0.0937795
Omega	0.0000000	1.0000000	0.0000000	0.0030588
Cutoff	0.0000000	0.0000000	1.0000000	-0.0005127
Coefficient	-0.0937795	0.0030588	-0.0005127	1.0000000

Graphics:

Change in Alpha with cutoffs greater than or equal to top 75% of data

```
# Calculate subsetted averages
tobit_alpha <- subset(tobit, tobit$Cutoff >= .75)
tob <- tapply(tobit_alpha$Coefficient, tobit_alpha$Alpha, mean)</pre>
quantile_alpha <- subset(quantile, quantile$Cutoff >= .75)
quant <- tapply(quantile_alpha$Coefficient, quantile_alpha$Alpha, mean)
# Create dataframe with bias and MSE
data <- data.frame(-3:3, tob - .5, (tob - .5)^2, "Tobit", row.names = 1:7)
data1 <- data.frame(-3:3, quant - .5, (quant - .5)^2, "Quantile", row.names = 1:nrow(data))
colnames(data) <- c("Alpha", "Bias", "MSE" , "Type")</pre>
colnames(data1) <- colnames(data)</pre>
final <- as.data.frame(rbind(data, data1))</pre>
# Bias graph
jpeg('alpha_bias.jpg', quality = 100, width = 12, height = 8, units = "in", res = 300)
ggplot(final, aes(x = Alpha, y = Bias, color = Type)) +
 geom_point() +
 geom_path() +
 geom_abline(mapping = aes(slope = 0, intercept = 0)) +
 ylim(-.1, .1) +
 labs(title = "Tobit and Quantile Bias for levels of Skewness",
       subtitle = "Only Cutoffs above 75th percentile")
dev.off()
```

pdf ## 2

```
# MSE graph
jpeg('alpha_mse.jpg', quality = 100, width = 12, height = 8, units = "in", res = 300)
ggplot(final, aes(x = Alpha, y = MSE, color = Type)) +
  geom point() +
  geom_path() +
  geom_abline(mapping = aes(slope = 0, intercept = 0)) +
 ylim(-.005, .005) +
  labs(title = "Tobit and Quantile MSE for levels of Skewness",
       subtitle = "Only Cutoffs above 75th percentile")
dev.off()
## pdf
## 2
Change in Omega with 75th Percentile and Above and Base Alpha level
# Calculate subsetted averages
tobit_alpha <- subset(tobit, tobit$Cutoff >= .75 & tobit$Alpha == 0)
tob <- tapply(tobit_alpha$Coefficient, tobit_alpha$Omega, mean)</pre>
quantile_alpha <- subset(quantile, quantile$Cutoff >= .75 & quantile$Alpha %in% c(-3,-2,-1))
quant <- tapply(quantile_alpha$Coefficient, quantile_alpha$Omega, mean)
# Create dataframe with bias and MSE
data <- data.frame(1:7, tob - .5, (tob - .5)^2, "Tobit", row.names = 1:nrow(data))
data1 <- data.frame(1:7, quant - .5, (quant - .5)^2, "Quantile", row.names = 1:nrow(data1))</pre>
colnames(data) <- c("Omega", "Bias", "MSE", "Type")</pre>
colnames(data1) <- colnames(data)</pre>
final <- as.data.frame(rbind(data, data1))</pre>
# Bias graph
jpeg('omega_bias.jpg', quality = 100, width = 12, height = 8, units = "in", res = 300)
ggplot(final, aes(x = Omega, y = Bias, color = Type)) +
 geom point() +
 geom_path() +
  geom_abline(mapping = aes(slope = 0, intercept = 0)) +
 ylim(-.1, .1) +
  labs(title = "Tobit and Quantile Bias for levels of Spread",
       subtitle = "Only Cutoffs above 75th percentile and Base Skewness")
dev.off()
## pdf
##
   2
# MSE graph
jpeg('omega_mse.jpg', quality = 100, width = 12, height = 8, units = "in", res = 300)
ggplot(final, aes(x = Omega, y = MSE, color = Type)) +
 geom_point() +
  geom_path() +
  geom_abline(mapping = aes(slope = 0, intercept = 0)) +
 ylim(-.0005, .0005) +
 labs(title = "Tobit and Quantile MSE for levels of Spread",
       subtitle = "Only Cutoffs above 75th percentile and Base Skewness")
```

dev.off()

```
## pdf
## 2
```

Change in Cutoff with Base Alpha level

```
# Calculate subsetted averages
tobit_alpha <- subset(tobit, tobit$Alpha == 0)</pre>
tob <- tapply(tobit_alpha$Coefficient, tobit_alpha$Cutoff, mean)
quantile_alpha <- subset(quantile, quantile$Alpha %in% c(-3,-2,-1))
quant <- tapply(quantile_alpha$Coefficient, quantile_alpha$Cutoff, mean)
# Create dataframe with bias and MSE
data <- data.frame(c(.65, .7, .75, .8, .85, .9, .95), tob - .5, (tob - .5)^2, "Tobit",
                   row.names = 1:nrow(data))
data1 <- data.frame(c(.65, .7, .75, .8, .85, .9, .95), quant - .5, (quant - .5)^2, "Quantile",
                    row.names = 1:nrow(data1))
colnames(data) <- c("Cutoff", "Bias", "MSE", "Type")</pre>
colnames(data1) <- colnames(data)</pre>
final <- as.data.frame(rbind(data, data1))</pre>
# Bias graph
jpeg('cutoff_bias.jpg', quality = 100, width = 12, height = 8, units = "in", res = 300)
ggplot(final, aes(x = Cutoff, y = Bias, color = Type)) +
 geom_point() +
  geom_path() +
 geom_abline(mapping = aes(slope = 0, intercept = 0)) +
 vlim(-.1, .1) +
 labs(title = "Tobit and Quantile Bias for Percentile Cutoff Levels",
       subtitle = "Only Base Skewness")
dev.off()
## pdf
##
   2
# MSE graph
jpeg('cutoff_mse.jpg', quality = 100, width = 12, height = 8, units = "in", res = 300)
ggplot(final, aes(x = Cutoff, y = MSE, color = Type)) +
 geom_point() +
  geom_path() +
 geom_abline(mapping = aes(slope = 0, intercept = 0)) +
 ylim(-.0005, .0005) +
  labs(title = "Tobit and Quantile MSE for Percentile Cutoff Levels",
       subtitle = "Only Base Skewness")
dev.off()
## pdf
##
```