

Measurement_Error_-Apr8.R

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```
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# Course: MSBA 6440
# Session: Selection and Measurement Error
# Topic: Measurement Error

suppressWarnings(suppressPackageStartupMessages({
library(MASS)
library(AER)
library(stargazer)
}))

# X and Y have classical measurement error. The true value are Xt and Yt, but they are meas-
# ured with error.
# X is measured as true X (Xt) plus error (ex)
# Y is measured as true Y (Yt) plus error (ey)
# The mean of Yt, Xt, ey and ex is (10, 7, 0, 0)
# The standard deviation of Yt, Xt, ey, and ex is (4, 8, 3, 6)
# The correlation of Yt and Xt is 0.7; Yt and Xt are uncorrelated with ey and ex; and ey and
# ex are uncorrelated with each other.

set.seed(1234)

Yt_Xt_ey_ex <- (mvrnorm(10000, c(10, 7, 0, 0), matrix(c(16, 22.4, 0.0, 0.0, 22.4, 64, 0, 0, 0,
0, 9, 0, 0, 0, 0, 36), ncol = 4)))

Yt <- Yt_Xt_ey_ex[,1]
Xt <- Yt_Xt_ey_ex[,2]
ey <- Yt_Xt_ey_ex[,3]
ex <- Yt_Xt_ey_ex[,4]

Y <- Yt + ey
X <- Xt + ex

# Check everything worked as expected.

cov(Yt_Xt_ey_ex)

##           [,1]      [,2]      [,3]      [,4]
## [1,] 15.8500152 21.7983191 -0.1185045  0.3525733
## [2,] 21.7983191 62.4120430 -0.3538506  0.4489615
## [3,] -0.1185045 -0.3538506  8.9843536 -0.0209746
## [4,]  0.3525733  0.4489615 -0.0209746 36.4322839
```

```
cor(Yt_Xt_ey_ex)
```

```
##           [,1]      [,2]      [,3]      [,4]
## [1,]  1.000000000  0.693064996 -0.009930629  0.014672070
## [2,]  0.693064996  1.000000000 -0.014943156  0.009415246
## [3,] -0.009930629 -0.014943156  1.000000000 -0.001159330
## [4,]  0.014672070  0.009415246 -0.001159330  1.000000000
```

```
sd(ey)
```

```
## [1] 2.997391
```

```
sd(ex)
```

```
## [1] 6.035916
```

```
sd(Yt)
```

```
## [1] 3.981208
```

```
sd(Xt)
```

```
## [1] 7.900129
```

```
mean (Yt)
```

```
## [1] 10.04655
```

```
mean(Xt)
```

```
## [1] 7.037755
```

```
mean(ey)
```

```
## [1] 0.01225815
```

```
mean(ex)
```

```
## [1] -0.04648214
```

#1. Measurement error in X, underestimates the effect of X on Y. The reliability of mismeasurement is the magnitude of the mismeasurement.

```
summary(lm(Yt-Xt))
```

```
##
## Call:
## lm(formula = Yt ~ Xt)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -11.0613  -1.9560   0.0189   1.9343  10.9935
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  7.588506   0.038439  197.42  <2e-16 ***
## Xt           0.349265   0.003633   96.13  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.87 on 9998 degrees of freedom
## Multiple R-squared:  0.4803, Adjusted R-squared:  0.4803
## F-statistic: 9241 on 1 and 9998 DF, p-value: < 2.2e-16
```

```
summary(lm(Yt~X))
```

```
##
## Call:
## lm(formula = Yt ~ X)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -14.0113  -2.2426  -0.0154   2.2387  13.9810
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  8.493914   0.040360  210.45  <2e-16 ***
## X           0.222081   0.003311   67.08  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.306 on 9998 degrees of freedom
## Multiple R-squared:  0.3104, Adjusted R-squared:  0.3103
## F-statistic: 4500 on 1 and 9998 DF, p-value: < 2.2e-16
```

```
Reliability <- var(Xt)/var(X)
```

```
Reliability
```

```
## [1] 0.6257333
```

```
summary(lm(Yt~X))$coefficients[2,1]/ summary(lm(Yt~Xt))$coefficients[2,1]
```

```
## [1] 0.6358541
```

#2. Measurement error in Y, does not influence the coefficient of X, but exaggerates the standard error of the regression coefficient.

```
summary(lm(Yt~Xt))
```

```
##
## Call:
## lm(formula = Yt ~ Xt)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -11.0613  -1.9560   0.0189   1.9343  10.9935
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  7.588506   0.038439  197.42  <2e-16 ***
## Xt           0.349265   0.003633   96.13  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.87 on 9998 degrees of freedom
## Multiple R-squared:  0.4803, Adjusted R-squared:  0.4803
## F-statistic: 9241 on 1 and 9998 DF, p-value: < 2.2e-16
```

```
summary(lm(Y~Xt))
```

```
##
## Call:
## lm(formula = Y ~ Xt)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -14.4804  -2.7805   0.0196   2.7619  16.3189
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  7.640666   0.055594  137.44  <2e-16 ***
## Xt           0.343595   0.005255   65.39  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 4.151 on 9998 degrees of freedom
## Multiple R-squared:  0.2996, Adjusted R-squared:  0.2995
## F-statistic: 4276 on 1 and 9998 DF, p-value: < 2.2e-16
```

*#3. Given that the measurement error in Y is more innocuous than the measurement error in X, we might run the reverse regression.
The coefficient of the regular regression and the inverse of the coefficient of the reverse regression, bracket the true coefficient.*

```
regular_reg <- (lm(Yt~X))
b <- summary(regular_reg)$coefficients[2,1]
```

```
reverse_reg <- (lm(X~Yt))
reverse_reg_coeff <- summary(reverse_reg)$coefficients[2,1]

g <- 1/(summary(reverse_reg)$coefficients[2,1])
```

```
b/g
```

```
## [1] 0.3103656
```

```
summary((lm(Yt~X)))
```

```
##
## Call:
## lm(formula = Yt ~ X)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -14.0113  -2.2426  -0.0154   2.2387  13.9810
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  8.493914   0.040360  210.45  <2e-16 ***
## X            0.222081   0.003311   67.08  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.306 on 9998 degrees of freedom
## Multiple R-squared:  0.3104, Adjusted R-squared:  0.3103
## F-statistic: 4500 on 1 and 9998 DF, p-value: < 2.2e-16
```

*#4. If there is a good instrument for X_t , then the true estimate can be recovered.
 # Lets say that Z is a good instrument for X_t . Like X_t , Z has a mean of 7 and a sd of 8. Z
 # has a correlation of 0.5 with X_t , and Z is uncorrelated with ey and ex .
 # Z has a correlation of 0.35 with Y_t which is the product of 0.7 and 0.5 i.e, the correl-
 # ation between Y_t and X_t and the correlation between X_t and Z .*

```
Yt_Xt_ey_ex_Z <- (mvrnorm(10000, c(10, 7, 0, 0, 7), matrix(c(16, 22.4, 0.0, 0.0, 11.2, 22.4,
64, 0, 0, 32, 0, 0, 9, 0, 0, 0, 0, 36, 0, 11.2, 32, 0, 0, 64), ncol = 5)))
```

```
Yt <- Yt_Xt_ey_ex_Z[,1]
Xt <- Yt_Xt_ey_ex_Z[,2]
ey <- Yt_Xt_ey_ex_Z[,3]
ex <- Yt_Xt_ey_ex_Z[,4]
Z <- Yt_Xt_ey_ex_Z[,5]
```

```
Y <- Yt + ey
X <- Xt + ex
```

```
cov(Yt_Xt_ey_ex_Z)
```

```
##           [,1]      [,2]      [,3]      [,4]      [,5]
## [1,] 16.0047135 22.63816128 -0.00282960 -0.37864749 11.2970827
## [2,] 22.6381613 64.48127066 0.02019033 -1.05799144 32.3030631
## [3,] -0.0028296 0.02019033 8.92867464 -0.08615143 -0.1550035
## [4,] -0.3786475 -1.05799144 -0.08615143 35.60204905 -0.9907565
## [5,] 11.2970827 32.30306312 -0.15500346 -0.99075646 64.8426342
```

```
cor(Xt, Z)
```

```
## [1] 0.4995703
```

```
cor(X, Z)
```

```
## [1] 0.3928657
```

```
cor(Yt, Z)
```

```
## [1] 0.3506808
```

```
cor(Y, Z)
```

```
## [1] 0.277137
```

```
ols_true <- lm((Yt~Xt))
```

```
ivreg1 <- ivreg(formula=Yt ~ X | Z)
```

```
ivreg2 <- ivreg(formula=Y ~ X | Z)
```

```
stargazer(ols_true,ivreg1, ivreg2,type="text",title="True vs.Instrumented",
  column.labels = c("True","IV1", "IV2"))
```

```
##
## True vs.Instrumented
## =====
##                               Dependent variable:
##                               -----
##                               Yt ~ Xt           Yt           Y
##                               OLS           instrumental instrumental
##                               True           IV1           IV2
##                               (1)           (2)           (3)
## -----
## Xt                               0.351***
##                               (0.004)
##
## X                               0.361***           0.356***
##                               (0.009)           (0.012)
##
## Constant                       7.567***           7.516***           7.548***
```

	(0.038)	(0.073)	(0.095)
##			
##			
##	-----		
## Observations	10,000	10,000	10,000
## R2	0.497	0.207	0.136
## Adjusted R2	0.497	0.207	0.136
## Residual Std. Error (df = 9998)	2.839	3.563	4.641
## F Statistic	9,862.679*** (df = 1; 9998)		
##	=====		
## Note:	*p<0.1; **p<0.05; ***p<0.01		