EEP C118 Section 9A: Measurement Error Suppose we want to know the relationship between hours slept and hours worked, but we have concerns about measurement error. We will use sleep.dta which contains the relevant data. Remember to read in .dta files, we need to use the haven package. 0. Data Prep In [4]: library(tidyverse) library(haven) sleepdata <- read dta("sleep75.dta")</pre> In [5]: | colnames(sleepdata) 'age' · 'black' · 'case' · 'clerical' · 'construc' · 'educ' · 'earns74' · 'gdhlth' · 'inlf' · 'leis1' · 'leis2' · 'leis3' · 'smsa' · 'lhrwage' · 'lothinc' · 'male' · 'marr' · 'prot' · 'rlxall' · 'selfe' · 'sleep' · 'slpnaps' · 'south' · 'spsepay' · 'spwrk75' · 'totwrk' · 'union' · 'worknrm' · 'workscnd' · 'exper' · 'yngkid' · 'yrsmarr' · 'hrwage' · 'agesq' Let's make a new variable called *sleephrs* that is hours slept per night. In [6]: sleepdata\$sleephrs<-sleepdata\$sleep/(7\*60)</pre> summary(sleepdata\$sleephrs) Min. 1st Qu. Median Mean 3rd Qu. 1.798 7.179 7.787 7.777 8.410 11.179 Let's do the same thing for *totwrk*, making a variable called *wrkhrs*. In [7]: | sleepdata\$wrkhrs<-sleepdata\$totwrk/(7\*60)</pre> summary(sleepdata\$wrkhrs) Min. 1st Qu. Median Mean 3rd Qu. Max. 0.000 3.699 5.448 5.055 6.409 15.274 Great! Now let's run a regression and look at the results. In [8]: | slr<-lm(sleephrs~wrkhrs, data=sleepdata)</pre> summary(slr) Call: lm(formula = sleephrs ~ wrkhrs, data = sleepdata) Residuals: Min 1Q Median 3Q -5.7856 -0.5720 0.0117 0.5965 3.1898 Coefficients: Estimate Std. Error t value Pr(>|t|) (Intercept) 8.53899 0.09265 92.165 <2e-16 \*\*\* wrkhrs -0.15075 0.01674 -9.005 <2e-16 \*\*\* Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 '' 1 Residual standard error: 1.003 on 704 degrees of freedom Multiple R-squared: 0.1033, Adjusted R-squared: 0.102 F-statistic: 81.09 on 1 and 704 DF, p-value: < 2.2e-16 1. Classical measurement error in an independent variable In [10]: # create error variable: random number between 0 and 4 sleepdata\$error=runif(nrow(sleepdata), 0,4) summary(sleepdata\$error) # Add random noise to work hours sleepdata\$wrkhrse=sleepdata\$wrkhrs+sleepdata\$error summary(sleepdata\$wrkhrse) summary(sleepdata\$wrkhrs) slr cme<-lm(sleephrs~wrkhrse, data=sleepdata)</pre> summary(slr cme) Min. 1st Qu. Median Mean 3rd Qu. 0.005399 0.968199 1.984864 1.982547 3.017711 3.996401 Min. 1st Qu. Median Mean 3rd Qu. Max. 0.2061 5.5205 7.1721 7.0371 8.8014 18.0349 Min. 1st Qu. Median Mean 3rd Qu. Max. 0.000 3.699 5.448 5.055 6.409 15.274 lm(formula = sleephrs ~ wrkhrse, data = sleepdata) Residuals: Min 1Q Median 3Q -5.7925 -0.5726 0.0055 0.6275 3.1953 Coefficients: Estimate Std. Error t value Pr(>|t|) (Intercept) 8.64156 0.11110 77.785 < 2e-16 \*\*\* wrkhrse -0.12285 0.01483 -8.282 6.07e-16 \*\*\* Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1 Residual standard error: 1.011 on 704 degrees of freedom Multiple R-squared: 0.08879, Adjusted R-squared: 0.0875 F-statistic: 68.6 on 1 and 704 DF, p-value: 6.074e-16 What happened to our estimated relationship between work hours and sleep hours? What would happen if the random error in the measurement of work hours was larger? In [11]: # create error variable: random number between 0 and 8 sleepdata\$error=runif(nrow(sleepdata), 0,8) summary(sleepdata\$error) # Add random noise to work hours sleepdata\$wrkhrse=sleepdata\$wrkhrs+sleepdata\$error summary(sleepdata\$wrkhrse) summary(sleepdata\$wrkhrs) slr cme2<-lm(sleephrs~wrkhrse, data=sleepdata)</pre> summary(slr cme2) Min. 1st Qu. Median Mean 3rd Qu. 0.02018 1.81735 3.89180 3.92152 6.03313 7.98151 Min. 1st Qu. Median Mean 3rd Qu. Max. 0.579 6.599 9.035 8.976 11.444 18.098 Min. 1st Qu. Median Mean 3rd Qu. 0.000 3.699 5.448 5.055 6.409 15.274 lm(formula = sleephrs ~ wrkhrse, data = sleepdata) Residuals: Min 1Q Median 3Q Max -5.9401 -0.5899 0.0386 0.6116 3.4902 Coefficients: Estimate Std. Error t value Pr(>|t|) (Intercept) 8.39040 0.11684 71.814 < 2e-16 \*\*\* wrkhrse -0.06833 0.01227 -5.569 3.64e-08 \*\*\* Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 '' 1 Residual standard error: 1.036 on 704 degrees of freedom Multiple R-squared: 0.0422, Adjusted R-squared: 0.04084 F-statistic: 31.02 on 1 and 704 DF, p-value: 3.642e-08 What about coefficient estimates for covariates? Are they affected by classical measurement error? In [12]: | mlr<-lm(sleephrs~wrkhrs+male+age, data=sleepdata)</pre> summary(mlr) mlr cme<-lm(sleephrs~wrkhrse+male+age, data=sleepdata)</pre> summary(mlr cme) Call: lm(formula = sleephrs ~ wrkhrs + male + age, data = sleepdata) Residuals: Min 1Q Median 3Q Max -5.5482 -0.5740 -0.0034 0.6284 3.1975 Coefficients: Estimate Std. Error t value Pr(>|t|) (Intercept) 8.247958 0.161669 51.018 <2e-16 \*\*\* wrkhrs -0.165900 0.018001 -9.216 <2e-16 \*\*\* 0.202731 0.081827 2.478 0.0135 \* male 0.006512 0.003318 1.962 0.0501. ---Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.05 '.' 0.1 '' 1 Residual standard error: 0.9967 on 702 degrees of freedom Multiple R-squared: 0.1166, Adjusted R-squared: 0.1128 F-statistic: 30.88 on 3 and 702 DF, p-value: < 2.2e-16 lm(formula = sleephrs ~ wrkhrse + male + age, data = sleepdata) Residuals: Min 1Q Median 3Q -5.7929 -0.5819 0.0213 0.6250 3.4025 Coefficients: Estimate Std. Error t value Pr(>|t|) (Intercept) 8.064991 0.178491 45.184 < 2e-16 \*\*\* wrkhrse -0.069234 0.012669 -5.465 6.45e-08 \*\*\* male 0.032128 0.081301 0.395 0.6928 ---Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 '' 1 Residual standard error: 1.033 on 702 degrees of freedom Multiple R-squared: 0.05008, Adjusted R-squared: 0.04602 F-statistic: 12.34 on 3 and 702 DF, p-value: 7.15e-08 Why do the coefficient estimates change? Because the coefficient on work hours is biased, and male and age are not independent of work hours. Notice that while the coefficient on wkrhrs is attenuated, the coefficients on the other coviarates can change in a variety of ways -- it depends on their correlation with work hours. 2. Non-classical measurement error in independent variable Suppose  $wrkhrs\_reported1 = \sqrt{wrkhrs} + 5$ : people with low true work hours overreport their hours, and people with high true work hours underreport. The estimated coefficient on work hours becomes much larger! So we do not have attenuation here. Then consider other possible error values that are correlated with true work hours. These could also be correlated with reported work hours, and all bets are off in these cases. In one example, the estimated coefficient is slightly smaller, and in the other it is slightly larger. Note that it is easy to create a scenario where  $cov(x^*, e) = 0$  but somewhat difficult to create a scenario where cov(x, e) = 0. In [23]: # create error variable: random number between 0 and 4 sleepdata\$wrkhrs rp1=sqrt(sleepdata\$wrkhrs)+5 sleepdata\$wrkhrs rp2=sleepdata\$wrkhrs-sqrt(sleepdata\$wrkhrs+4)/2 sleepdata\$wrkhrs rp3=sleepdata\$wrkhrs+log(sleepdata\$wrkhrs+1) summary(sleepdata\$wrkhrs) summary(sleepdata\$wrkhrs rp1) summary(sleepdata\$wrkhrs rp2) summary(sleepdata\$wrkhrs rp3) print("No measurement error") slr<-lm(sleephrs~wrkhrs, data=sleepdata)</pre> summary(slr) print("Measurement error function 1") slr ncme1<-lm(sleephrs~wrkhrs rp1, data=sleepdata)</pre> summary(slr ncme1) print("Measurement error function 2") slr ncme2<-lm(sleephrs~wrkhrs rp2, data=sleepdata)</pre> summary(slr ncme2) print("Measurement error function 3") slr\_ncme3<-lm(sleephrs~wrkhrs\_rp3, data=sleepdata)</pre> summary(slr ncme3) Min. 1st Qu. Median Mean 3rd Qu. Max. 0.000 3.699 5.448 5.055 6.409 15.274 Min. 1st Qu. Median Mean 3rd Qu. Max. 5.000 6.923 7.334 7.149 7.532 8.908 Min. 1st Qu. Median Mean 3rd Qu. Max. -1.000 2.311 3.911 3.563 4.796 13.079 Min. 1st Qu. Median Mean 3rd Qu. Max. 0.000 5.246 7.311 6.751 8.412 18.063 [1] "No measurement error" lm(formula = sleephrs ~ wrkhrs, data = sleepdata) Residuals: Min 1Q Median 3Q -5.7856 -0.5720 0.0117 0.5965 3.1898 Coefficients: Estimate Std. Error t value Pr(>|t|) (Intercept) 8.53899 0.09265 92.165 <2e-16 \*\*\* wrkhrs -0.15075 0.01674 -9.005 <2e-16 \*\*\* Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1 Residual standard error: 1.003 on 704 degrees of freedom Multiple R-squared: 0.1033, Adjusted R-squared: 0.102 F-statistic: 81.09 on 1 and 704 DF, p-value: < 2.2e-16 [1] "Measurement error function 1" Call: lm(formula = sleephrs ~ wrkhrs rp1, data = sleepdata) Residuals: 1Q Median Min 3Q -5.8076 -0.5961 0.0176 0.6223 3.2907 Coefficients: Estimate Std. Error t value Pr(>|t|) wrkhrs rp1 -0.46535 0.05765 -8.072 2.99e-15 \*\*\* Signif. codes: 0 \\*\*\*' 0.001 \\*\*' 0.01 \\*' 0.05 \'.' 0.1 \' 1 Residual standard error: 1.013 on 704 degrees of freedom Multiple R-squared: 0.08472, Adjusted R-squared: 0.08342 F-statistic: 65.16 on 1 and 704 DF, p-value: 2.986e-15 [1] "Measurement error function 2" Call: lm(formula = sleephrs ~ wrkhrs\_rp2, data = sleepdata) Residuals: Min 1Q Median 3Q -5.7860 -0.5733 0.0102 0.6005 3.1872 Coefficients: Estimate Std. Error t value Pr(>|t|) (Intercept) 8.36621 0.07543 110.917 <2e-16 \*\*\* wrkhrs rp2 -0.16536 0.01833 -9.021 <2e-16 \*\*\* Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 '' 1 Residual standard error: 1.003 on 704 degrees of freedom Multiple R-squared: 0.1036, Adjusted R-squared: 0.1023 F-statistic: 81.37 on 1 and 704 DF, p-value: < 2.2e-16 [1] "Measurement error function 3" Call: lm(formula = sleephrs ~ wrkhrs rp3, data = sleepdata) Residuals: 10 Median 3Q -5.7864 -0.5843 0.0183 0.5980 3.2109 Coefficients: Estimate Std. Error t value Pr(>|t|) (Intercept) 8.60003 0.10003 85.974 <2e-16 \*\*\* wrkhrs rp3 -0.12191 0.01372 -8.886 <2e-16 \*\*\* Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 '' 1 Residual standard error: 1.004 on 704 degrees of freedom Multiple R-squared: 0.1008, Adjusted R-squared: 0.09957 F-statistic: 78.96 on 1 and 704 DF, p-value: < 2.2e-16 3. Classical measurement error in the outcome variable Suppose sleep is measured with classical random error of between 0-4 hours. What happens to our estimated coefficients? In [24]: sleepdata\$error=runif(nrow(sleepdata), 0,4) sleepdata\$sleephrse=sleepdata\$sleephrs+sleepdata\$error summary(sleepdata\$sleephrse) summary(sleepdata\$sleephrs) mlr<-lm(sleephrs~wrkhrs+male, data=sleepdata)</pre> summary(mlr) mlr2<-lm(sleephrse~wrkhrs+male, data=sleepdata)</pre> summary(mlr2) Min. 1st Qu. Median Mean 3rd Qu. Max. 3.433 8.741 9.895 9.844 11.016 14.817 Min. 1st Qu. Median Mean 3rd Qu. 1.798 7.179 7.787 7.777 8.410 11.179 lm(formula = sleephrs ~ wrkhrs + male, data = sleepdata) Residuals: 1Q Median 3Q Max -5.6432 -0.5801 -0.0053 0.6291 3.2851 Coefficients: Estimate Std. Error t value Pr(>|t|) (Intercept) 8.50763 0.09307 91.409 < 2e-16 \*\*\* 0.01800 -9.349 < 2e-16 \*\*\* wrkhrs -0.16825 male 0.21151 0.08187 2.584 0.00998 \*\* Signif. codes: 0 \\*\*\*' 0.001 \\*\*' 0.01 \\*' 0.05 \'.' 0.1 \' 1 Residual standard error: 0.9987 on 703 degrees of freedom Multiple R-squared: 0.1117, Adjusted R-squared: 0.1092 F-statistic: 44.21 on 2 and 703 DF, p-value: < 2.2e-16 Call: lm(formula = sleephrse ~ wrkhrs + male, data = sleepdata) Residuals: Min 1Q Median 3Q Max -6.0438 -1.0885 0.0386 1.2204 4.3336 Estimate Std. Error t value Pr(>|t|) 0.02792 -6.896 1.19e-11 \*\*\* wrkhrs -0.19257 male 0.21060 0.12703 1.658 0.0978 . Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1 Residual standard error: 1.55 on 703 degrees of freedom Multiple R-squared: 0.06464, Adjusted R-squared: 0.06198 F-statistic: 24.29 on 2 and 703 DF, p-value: 6.294e-11 The random error might be correlated with the X variables by random chance, leading to bias, but we note that the coefficients are pretty similar. In both cases, however, the standard errors have increased: our estimates are less precise! Let's look at what happens if we repeat the process of generating random error multiple times. We'll do 1000 iterations of generating random error, regress sleep hours measured with error on work hours, store the estimated coefficients, then plot the distribution. In [26]: library(ggplot2) In [59]: estimates <- data.frame(matrix(NA, # Create empty data frame nrow = 1000,ncol = 1))estimates\$coef<-0 estimates\$tstat<-0 for(i in 1:1000) { set.seed(i) sleepdata\$error=runif(nrow(sleepdata), 0,4) sleepdata\$sleephrse=sleepdata\$sleephrs+sleepdata\$error clm y<-lm(sleephrse~wrkhrs, data=sleepdata)</pre> estimates[i,'coef']<-summary(clm y)\$coefficients[2,1]</pre> estimates[i,'tstat']<-summary(clm y)\$coefficients[2,3]</pre> In [60]: | slr<-lm(sleephrs~wrkhrs, data=sleepdata)</pre> truth<-summary(slr)\$coefficients[2,1]</pre> p <- ggplot(estimates, aes(x=coef)) +</pre> geom density(fill = "blue", alpha = .5) + geom vline( xintercept = truth) р -0.12 coef The estimates are centered around the 'true' coefficient value! What about the t-stats? In [61]: | slr<-lm(sleephrs~wrkhrs, data=sleepdata)</pre> truth<-summary(slr)\$coefficients[2,3]</pre> p <- ggplot(estimates, aes(x=tstat)) +</pre> geom density(fill = "blue", alpha = .5) + geom vline( xintercept = truth) р 0.5 0.4 0.3 density The 'true' t-stat is around -9, but the t-stats with classical error in the outcome variable are centered around -6, with almost none as large (in absolute value) as -9. This shows how the estimates become less precise. What if y is measured with non-classical measurement error? Let's consider where the error varies first by sex and then by sleep hours. In [63]: sleepdata\$sleephrse1=sleepdata\$sleephrs+sqrt (sleepdata\$error+1) \*sleepdata\$male sleepdata\$sleephrse2=sleepdata\$sleephrs+sqrt(sleepdata\$sleephrs+1)\*sleepdata\$error mlr<-lm(sleephrs~wrkhrs+male, data=sleepdata)</pre> summary(mlr) mlr1<-lm(sleephrse1~wrkhrs+male, data=sleepdata)</pre> summary(mlr1) mlr2<-lm(sleephrse2~wrkhrs+male, data=sleepdata)</pre> summary(mlr2) Call: lm(formula = sleephrs ~ wrkhrs + male, data = sleepdata) Residuals: 1Q Median 3Q Max -5.6432 -0.5801 -0.0053 0.6291 3.2851 Coefficients: Estimate Std. Error t value Pr(>|t|) (Intercept) 8.50763 0.09307 91.409 < 2e-16 \*\*\* wrkhrs -0.16825 0.01800 -9.349 < 2e-16 \*\*\* 0.21151 0.08187 2.584 0.00998 \*\* male Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1 Residual standard error: 0.9987 on 703 degrees of freedom Multiple R-squared: 0.1117, Adjusted R-squared: 0.1092 F-statistic: 44.21 on 2 and 703 DF, p-value: < 2.2e-16 lm(formula = sleephrse1 ~ wrkhrs + male, data = sleepdata) Residuals: 1Q Median 3Q Max -5.6469 -0.6115 0.0190 0.6271 3.2858 Coefficients: Estimate Std. Error t value Pr(>|t|) (Intercept) 8.50099 0.09699 87.646 <2e-16 \*\*\* wrkhrs -0.16662 0.01876 -8.884 <2e-16 \*\*\* 1.90969 0.08532 22.383 <2e-16 \*\*\* male Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1 Residual standard error: 1.041 on 703 degrees of freedom Multiple R-squared: 0.4162, Adjusted R-squared: 0.4146 F-statistic: 250.6 on 2 and 703 DF, p-value: < 2.2e-16 lm(formula = sleephrse2 ~ wrkhrs + male, data = sleepdata) Residuals: 1Q Median 3Q -8.3809 -3.1322 0.1856 3.0473 9.0247 Coefficients: Estimate Std. Error t value Pr(>|t|) wrkhrs -0.24996 0.06746 -3.705 0.000228 \*\*\* male Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1 Residual standard error: 3.744 on 703 degrees of freedom Multiple R-squared: 0.01928, Adjusted R-squared: 0.01649 F-statistic: 6.911 on 2 and 703 DF, p-value: 0.001066 All bets are off!