

PSY 653 Module 1: Missing Data

Jan 29, 2020

Neil Yetz & Gemma Wallace

Part 1: In class Demo

Load Libraries

```
library(tidyverse)
```

```
## Warning: package 'ggplot2' was built under R version 3.6.3
## Warning: package 'tibble' was built under R version 3.6.3
## Warning: package 'tidyr' was built under R version 3.6.3
## Warning: package 'dplyr' was built under R version 3.6.3
## Warning: package 'forcats' was built under R version 3.6.3
```

```
library(mice)
```

```
## Warning: package 'mice' was built under R version 3.6.3
```

```
library(olsrr)
```

```
## Warning: package 'olsrr' was built under R version 3.6.3
```

Read in data

```
mice_data1 <- read_csv("mice_data1.csv")
```

Simple Linear regression model $X_1 \sim X_2$

Using pairwise deletion for missing data by default

```
mod1 <- lm(X1 ~ X2, data = mice_data1)
ols_regress(mod1)
```

```
##                               Model Summary
## -----
## R                               0.026      RMSE           0.714
## R-Squared                       0.001      Coef. Var      39.353
## Adj. R-Squared                  -0.002      MSE            0.510
## Pred R-Squared                  -0.009      MAE            0.592
## -----
## RMSE: Root Mean Square Error
## MSE: Mean Square Error
## MAE: Mean Absolute Error
##
```

```
##                                ANOVA
## -----
##              Sum of
##              Squares          DF      Mean Square      F      Sig.
## -----
## Regression      0.137           1          0.137      0.27      0.6039
## Residual       204.939         402          0.510
## Total          205.077         403
## -----
##
##                                Parameter Estimates
## -----
##      model      Beta      Std. Error      Std. Beta      t      Sig.      lower      upper
## -----
## (Intercept)    1.828        0.044                41.152    0.000        1.741        1.916
## X2             -0.011        0.022         -0.026     -0.519    0.604       -0.055        0.032
## -----
```

Impute the dataset 5 times (using mice)

```
imputed_data <- mice(mice_data1, m=5, maxit = 50, method = 'pmm', seed = 500)
```

```
##
## iter imp variable
## 1 1 X1
## 1 2 X1
## 1 3 X1
## 1 4 X1
## 1 5 X1
## 2 1 X1
## 2 2 X1
## 2 3 X1
## 2 4 X1
## 2 5 X1
## 3 1 X1
## 3 2 X1
## 3 3 X1
## 3 4 X1
## 3 5 X1
## 4 1 X1
## 4 2 X1
## 4 3 X1
## 4 4 X1
## 4 5 X1
## 5 1 X1
## 5 2 X1
## 5 3 X1
## 5 4 X1
## 5 5 X1
## 6 1 X1
## 6 2 X1
## 6 3 X1
## 6 4 X1
## 6 5 X1
```

##	7	1	X1
##	7	2	X1
##	7	3	X1
##	7	4	X1
##	7	5	X1
##	8	1	X1
##	8	2	X1
##	8	3	X1
##	8	4	X1
##	8	5	X1
##	9	1	X1
##	9	2	X1
##	9	3	X1
##	9	4	X1
##	9	5	X1
##	10	1	X1
##	10	2	X1
##	10	3	X1
##	10	4	X1
##	10	5	X1
##	11	1	X1
##	11	2	X1
##	11	3	X1
##	11	4	X1
##	11	5	X1
##	12	1	X1
##	12	2	X1
##	12	3	X1
##	12	4	X1
##	12	5	X1
##	13	1	X1
##	13	2	X1
##	13	3	X1
##	13	4	X1
##	13	5	X1
##	14	1	X1
##	14	2	X1
##	14	3	X1
##	14	4	X1
##	14	5	X1
##	15	1	X1
##	15	2	X1
##	15	3	X1
##	15	4	X1
##	15	5	X1
##	16	1	X1
##	16	2	X1
##	16	3	X1
##	16	4	X1
##	16	5	X1
##	17	1	X1
##	17	2	X1
##	17	3	X1
##	17	4	X1

##	17	5	X1
##	18	1	X1
##	18	2	X1
##	18	3	X1
##	18	4	X1
##	18	5	X1
##	19	1	X1
##	19	2	X1
##	19	3	X1
##	19	4	X1
##	19	5	X1
##	20	1	X1
##	20	2	X1
##	20	3	X1
##	20	4	X1
##	20	5	X1
##	21	1	X1
##	21	2	X1
##	21	3	X1
##	21	4	X1
##	21	5	X1
##	22	1	X1
##	22	2	X1
##	22	3	X1
##	22	4	X1
##	22	5	X1
##	23	1	X1
##	23	2	X1
##	23	3	X1
##	23	4	X1
##	23	5	X1
##	24	1	X1
##	24	2	X1
##	24	3	X1
##	24	4	X1
##	24	5	X1
##	25	1	X1
##	25	2	X1
##	25	3	X1
##	25	4	X1
##	25	5	X1
##	26	1	X1
##	26	2	X1
##	26	3	X1
##	26	4	X1
##	26	5	X1
##	27	1	X1
##	27	2	X1
##	27	3	X1
##	27	4	X1
##	27	5	X1
##	28	1	X1
##	28	2	X1
##	28	3	X1

##	28	4	X1
##	28	5	X1
##	29	1	X1
##	29	2	X1
##	29	3	X1
##	29	4	X1
##	29	5	X1
##	30	1	X1
##	30	2	X1
##	30	3	X1
##	30	4	X1
##	30	5	X1
##	31	1	X1
##	31	2	X1
##	31	3	X1
##	31	4	X1
##	31	5	X1
##	32	1	X1
##	32	2	X1
##	32	3	X1
##	32	4	X1
##	32	5	X1
##	33	1	X1
##	33	2	X1
##	33	3	X1
##	33	4	X1
##	33	5	X1
##	34	1	X1
##	34	2	X1
##	34	3	X1
##	34	4	X1
##	34	5	X1
##	35	1	X1
##	35	2	X1
##	35	3	X1
##	35	4	X1
##	35	5	X1
##	36	1	X1
##	36	2	X1
##	36	3	X1
##	36	4	X1
##	36	5	X1
##	37	1	X1
##	37	2	X1
##	37	3	X1
##	37	4	X1
##	37	5	X1
##	38	1	X1
##	38	2	X1
##	38	3	X1
##	38	4	X1
##	38	5	X1
##	39	1	X1
##	39	2	X1

##	39	3	X1
##	39	4	X1
##	39	5	X1
##	40	1	X1
##	40	2	X1
##	40	3	X1
##	40	4	X1
##	40	5	X1
##	41	1	X1
##	41	2	X1
##	41	3	X1
##	41	4	X1
##	41	5	X1
##	42	1	X1
##	42	2	X1
##	42	3	X1
##	42	4	X1
##	42	5	X1
##	43	1	X1
##	43	2	X1
##	43	3	X1
##	43	4	X1
##	43	5	X1
##	44	1	X1
##	44	2	X1
##	44	3	X1
##	44	4	X1
##	44	5	X1
##	45	1	X1
##	45	2	X1
##	45	3	X1
##	45	4	X1
##	45	5	X1
##	46	1	X1
##	46	2	X1
##	46	3	X1
##	46	4	X1
##	46	5	X1
##	47	1	X1
##	47	2	X1
##	47	3	X1
##	47	4	X1
##	47	5	X1
##	48	1	X1
##	48	2	X1
##	48	3	X1
##	48	4	X1
##	48	5	X1
##	49	1	X1
##	49	2	X1
##	49	3	X1
##	49	4	X1
##	49	5	X1
##	50	1	X1

```
## 50 2 X1
## 50 3 X1
## 50 4 X1
## 50 5 X1
```

Regress X1 on X2 on imputed dataset using the “with” function

```
fit.mi <- with(imputed_data, exp= lm(X1 ~ X2))
summary(fit.mi)
```

```
## # A tibble: 10 x 5
##   term      estimate std.error statistic    p.value
##   <chr>      <dbl>    <dbl>    <dbl>    <dbl>
## 1 (Intercept)  1.84      0.0425    43.3 3.13e-160
## 2 X2          -0.0106    0.0207   -0.513 6.08e- 1
## 3 (Intercept)  1.82      0.0423    43.1 9.48e-160
## 4 X2          -0.00106    0.0205   -0.0516 9.59e- 1
## 5 (Intercept)  1.80      0.0429    42.0 1.46e-155
## 6 X2          -0.00443    0.0208   -0.213 8.32e- 1
## 7 (Intercept)  1.81      0.0419    43.3 3.27e-160
## 8 X2          -0.00242    0.0204   -0.119 9.05e- 1
## 9 (Intercept)  1.84      0.0419    43.9 2.47e-162
## 10 X2         -0.0119    0.0204   -0.584 5.60e- 1
```

Pool model estimates across imputed versions of the dataset

```
combimp <- pool(fit.mi)
summary(combimp)
```

```
##           term      estimate std.error statistic    df p.value
## 1 (Intercept)  1.822457999 0.04641326 39.2658866 100.5742 0.00000
## 2           X2 -0.006076888 0.02123498 -0.2861734 289.4252 0.77495
```

Part 2: Try it yourself

Read in data

```
mice_data2 <- read_csv("mice_data2.csv")
```

```
## Warning: Duplicated column names deduplicated: 'X4' => 'X4_1' [5]
```

Multiple Linear regression model $X1 \sim X2 + X3 + X4$

Using pairwise deletion for missing data by default

```
mod2 <- lm(X1 ~ X2 + X3 + X4, data = mice_data2)
ols_regress(mod2)
```

```
##                               Model Summary
## -----
## R                               0.103      RMSE                0.712
## R-Squared                       0.011      Coef. Var          39.254
## Adj. R-Squared                   0.003      MSE                  0.507
## Pred R-Squared                   -0.008      MAE                  0.588
```

```
## -----
## RMSE: Root Mean Square Error
## MSE: Mean Square Error
## MAE: Mean Absolute Error
##
## ANOVA
## -----
## Sum of
## Squares      DF      Mean Square      F      Sig.
## -----
## Regression    2.179        3        0.726    1.432    0.2330
## Residual     202.898      400        0.507
## Total        205.077      403
## -----
##
## Parameter Estimates
## -----
## model      Beta      Std. Error      Std. Beta      t      Sig.      lower      upper
## -----
## (Intercept)  1.681        0.089          18.832    0.000        1.505    1.856
## X2          -0.017        0.022          -0.040    0.431       -0.061    0.026
## X3           0.079        0.040           0.101    0.047         0.001    0.157
## X4          -0.001        0.024          -0.002    0.971       -0.048    0.046
## -----
```

Impute the dataset 5 times (using mice)

```
imputed_data <- mice(mice_data2, m=5, maxit = 50, method = 'pmm', seed = 500)
```

```
##
## iter imp variable
## 1 1 X1
## 1 2 X1
## 1 3 X1
## 1 4 X1
## 1 5 X1
## 2 1 X1
## 2 2 X1
## 2 3 X1
## 2 4 X1
## 2 5 X1
## 3 1 X1
## 3 2 X1
## 3 3 X1
## 3 4 X1
## 3 5 X1
## 4 1 X1
## 4 2 X1
## 4 3 X1
## 4 4 X1
## 4 5 X1
## 5 1 X1
## 5 2 X1
## 5 3 X1
```


##	5	4	X1
##	5	5	X1
##	6	1	X1
##	6	2	X1
##	6	3	X1
##	6	4	X1
##	6	5	X1
##	7	1	X1
##	7	2	X1
##	7	3	X1
##	7	4	X1
##	7	5	X1
##	8	1	X1
##	8	2	X1
##	8	3	X1
##	8	4	X1
##	8	5	X1
##	9	1	X1
##	9	2	X1
##	9	3	X1
##	9	4	X1
##	9	5	X1
##	10	1	X1
##	10	2	X1
##	10	3	X1
##	10	4	X1
##	10	5	X1
##	11	1	X1
##	11	2	X1
##	11	3	X1
##	11	4	X1
##	11	5	X1
##	12	1	X1
##	12	2	X1
##	12	3	X1
##	12	4	X1
##	12	5	X1
##	13	1	X1
##	13	2	X1
##	13	3	X1
##	13	4	X1
##	13	5	X1
##	14	1	X1
##	14	2	X1
##	14	3	X1
##	14	4	X1
##	14	5	X1
##	15	1	X1
##	15	2	X1
##	15	3	X1
##	15	4	X1
##	15	5	X1
##	16	1	X1
##	16	2	X1

##	16	3	X1
##	16	4	X1
##	16	5	X1
##	17	1	X1
##	17	2	X1
##	17	3	X1
##	17	4	X1
##	17	5	X1
##	18	1	X1
##	18	2	X1
##	18	3	X1
##	18	4	X1
##	18	5	X1
##	19	1	X1
##	19	2	X1
##	19	3	X1
##	19	4	X1
##	19	5	X1
##	20	1	X1
##	20	2	X1
##	20	3	X1
##	20	4	X1
##	20	5	X1
##	21	1	X1
##	21	2	X1
##	21	3	X1
##	21	4	X1
##	21	5	X1
##	22	1	X1
##	22	2	X1
##	22	3	X1
##	22	4	X1
##	22	5	X1
##	23	1	X1
##	23	2	X1
##	23	3	X1
##	23	4	X1
##	23	5	X1
##	24	1	X1
##	24	2	X1
##	24	3	X1
##	24	4	X1
##	24	5	X1
##	25	1	X1
##	25	2	X1
##	25	3	X1
##	25	4	X1
##	25	5	X1
##	26	1	X1
##	26	2	X1
##	26	3	X1
##	26	4	X1
##	26	5	X1
##	27	1	X1

##	27	2	X1
##	27	3	X1
##	27	4	X1
##	27	5	X1
##	28	1	X1
##	28	2	X1
##	28	3	X1
##	28	4	X1
##	28	5	X1
##	29	1	X1
##	29	2	X1
##	29	3	X1
##	29	4	X1
##	29	5	X1
##	30	1	X1
##	30	2	X1
##	30	3	X1
##	30	4	X1
##	30	5	X1
##	31	1	X1
##	31	2	X1
##	31	3	X1
##	31	4	X1
##	31	5	X1
##	32	1	X1
##	32	2	X1
##	32	3	X1
##	32	4	X1
##	32	5	X1
##	33	1	X1
##	33	2	X1
##	33	3	X1
##	33	4	X1
##	33	5	X1
##	34	1	X1
##	34	2	X1
##	34	3	X1
##	34	4	X1
##	34	5	X1
##	35	1	X1
##	35	2	X1
##	35	3	X1
##	35	4	X1
##	35	5	X1
##	36	1	X1
##	36	2	X1
##	36	3	X1
##	36	4	X1
##	36	5	X1
##	37	1	X1
##	37	2	X1
##	37	3	X1
##	37	4	X1
##	37	5	X1

##	38	1	X1
##	38	2	X1
##	38	3	X1
##	38	4	X1
##	38	5	X1
##	39	1	X1
##	39	2	X1
##	39	3	X1
##	39	4	X1
##	39	5	X1
##	40	1	X1
##	40	2	X1
##	40	3	X1
##	40	4	X1
##	40	5	X1
##	41	1	X1
##	41	2	X1
##	41	3	X1
##	41	4	X1
##	41	5	X1
##	42	1	X1
##	42	2	X1
##	42	3	X1
##	42	4	X1
##	42	5	X1
##	43	1	X1
##	43	2	X1
##	43	3	X1
##	43	4	X1
##	43	5	X1
##	44	1	X1
##	44	2	X1
##	44	3	X1
##	44	4	X1
##	44	5	X1
##	45	1	X1
##	45	2	X1
##	45	3	X1
##	45	4	X1
##	45	5	X1
##	46	1	X1
##	46	2	X1
##	46	3	X1
##	46	4	X1
##	46	5	X1
##	47	1	X1
##	47	2	X1
##	47	3	X1
##	47	4	X1
##	47	5	X1
##	48	1	X1
##	48	2	X1
##	48	3	X1
##	48	4	X1

```
## 48 5 X1
## 49 1 X1
## 49 2 X1
## 49 3 X1
## 49 4 X1
## 49 5 X1
## 50 1 X1
## 50 2 X1
## 50 3 X1
## 50 4 X1
## 50 5 X1
```

Regress X1 on X2 on imputed dataset using the “with” function

```
fit.mi2 <- with(imputed_data, exp= lm(X1 ~ X2 + X3 + X4))
summary(fit.mi2)
```

```
## # A tibble: 20 x 5
##   term      estimate std.error statistic  p.value
##   <chr>      <dbl>    <dbl>    <dbl>    <dbl>
## 1 (Intercept) 1.68      0.0874   19.2     4.08e-60
## 2 X2        -0.0263    0.0211  -1.25    2.13e- 1
## 3 X3         0.0871    0.0392   2.22    2.67e- 2
## 4 X4         0.000103  0.0238  0.00435 9.97e- 1
## 5 (Intercept) 1.70      0.0862   19.7     3.09e-62
## 6 X2        -0.00724   0.0208  -0.348   7.28e- 1
## 7 X3         0.0709    0.0386   1.84    6.71e- 2
## 8 X4        -0.000183  0.0234  -0.00783 9.94e- 1
## 9 (Intercept) 1.68      0.0849   19.8     1.31e-62
## 10 X2       -0.0261    0.0205  -1.28    2.03e- 1
## 11 X3         0.0838    0.0380   2.20    2.80e- 2
## 12 X4         0.000285  0.0231  0.0124   9.90e- 1
## 13 (Intercept) 1.69      0.0867   19.5     2.79e-61
## 14 X2       -0.0268    0.0209  -1.28    2.00e- 1
## 15 X3         0.0791    0.0389   2.03    4.25e- 2
## 16 X4       -0.00353    0.0236  -0.150   8.81e- 1
## 17 (Intercept) 1.64      0.0867   18.9     1.59e-58
## 18 X2         0.00604   0.0209   0.288   7.73e- 1
## 19 X3         0.0795    0.0389   2.05    4.14e- 2
## 20 X4         0.00378    0.0236   0.160   8.73e- 1
```

Pool model estimates across imputed versions of the dataset

```
combimp2 <- pool(fit.mi2)
summary(combimp2)
```

```
##      term      estimate std.error  statistic    df  p.value
## 1 (Intercept) 1.676699e+00 0.09002441 18.624934170 245.66880 0.0000000
## 2      X2 -1.610209e-02 0.02649093 -0.607834073  25.07851 0.5487629
## 3      X3  8.010165e-02 0.03929906  2.038258926 387.12469 0.0422032
## 4      X4  9.198331e-05 0.02364814  0.003889664 418.47772 0.9968984
```