

# Assignment 1: Data Manipulation - OLS/Discrete Choice

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## Part 1

### Exercise 1 Missing data

```
##                                     [,1]
## number.of.students                340823
## number.of.schools                  842
## number.of.programs                  33
## number.of.choices                   3086
## number.of.missing.test.score 179887
## apply.to.the.same.school           608970
## apply.to.less.than.6.choices 18954
```

### Exercise 2 Data

##		SchPrgm	sssdistrict	ssslong	ssslat	Cutoff	Quality	Size
## 1	100101	General Arts	Wa Municipal	-2.285030	10.03062	198	244.3924	79
## 2	100101	Home Economics	Wa Municipal	-2.285030	10.03062	199	229.4500	40
## 3	100101	Technical	Wa Municipal	-2.285030	10.03062	201	235.1020	49
## 4	100102	Agriculture	Wa Municipal	-2.285030	10.03062	273	292.5556	90
## 5	100102	Business	Wa Municipal	-2.285030	10.03062	283	303.3444	90
## 6	100102	General Arts	Wa Municipal	-2.285030	10.03062	291	311.1111	90
## 7	100102	General Science	Wa Municipal	-2.285030	10.03062	273	298.4333	90
## 8	100102	Home Economics	Wa Municipal	-2.285030	10.03062	262	278.8667	45
## 9	100102	Visual Arts	Wa Municipal	-2.285030	10.03062	250	275.2000	45
## 10	100104	General Arts	Wa Municipal	-2.285030	10.03062	319	337.4444	45
## 11	100104	General Science	Wa Municipal	-2.285030	10.03062	313	334.0000	45
## 12	100104	Home Economics	Wa Municipal	-2.285030	10.03062	282	309.3556	45
## 13	100105	Business	Wa Municipal	-2.285030	10.03062	251	268.0125	80
## 14	100105	General Arts	Wa Municipal	-2.285030	10.03062	258	274.7375	80
## 15	100105	Home Economics	Wa Municipal	-2.285030	10.03062	242	258.1625	80
## 16	100106	Agriculture	Wa Municipal	-2.285030	10.03062	223	240.6250	40
## 17	100106	Business	Wa Municipal	-2.285030	10.03062	238	253.5000	40
## 18	100106	General Arts	Wa Municipal	-2.285030	10.03062	248	268.9750	40
## 19	100201	Business	Lawra	-2.800941	10.54640	288	314.2750	80
## 20	100201	General Arts	Lawra	-2.800941	10.54640	319	339.0250	40

### Exercise 3 Distance

##	HighSchool	SeniorSchool	StuId	jsslong	jsslat
## 1	NA	Bosomtwe/Atwima/Kwanwoma (Kuntanase)	1	-1.5627517	6.559323
## 2	NA	Ho Municipal	2	0.5261422	6.717607
## 3	NA	Kwabre (Mamponteng)	3	-1.5414201	6.806778
## 4	NA	Kassena/Nankani (Navrongo)	4	-1.2174410	10.909423

## 5	NA	Atwima Mponua (Nyinahin)	5	-2.1771805	6.549507
## 6	NA	Kumasi Metro	6	-1.5971872	6.682060
## 7	NA	Nanumba North (Bimbilla)	7	-0.1417642	8.816774
## 8	NA	Jomoro (Half Assini)	8	-2.8032203	5.069508
## 9	NA	East Akim (Kibi)	9	-0.4543442	6.178558
## 10	NA	Ejura/Sekyedumase (Ejura)	10	-1.3679653	7.462874
## 11	NA	Sekyere West (Mampong)	11	-1.1800768	7.199565
## 12	NA	Kassena/Nankani (Navrongo)	12	-1.2174410	10.909423
## 13	NA	Agona Swedru	13	-0.7552425	5.617353
## 14	NA	Tolon Kunbungu (Tolon)	14	-1.1097199	9.527246
## 15	NA	Accra Metropolitan	15	-0.1971153	5.607396
## 16	NA	Mpohor-Wassa East (Daboase)	16	-1.6975694	5.330796
## 17	NA	Ejura/Sekyedumase (Ejura)	17	-1.3679653	7.462874
## 18	NA	Ga West (Amasaman)	18	-0.3975105	5.664688
## 19	NA	Wassa Amenfi (Asankragwa)	19	-2.3020179	5.725518
## 20	NA	Bole	20	-2.2666752	8.629696
##	ssslong	ssslat	JSdist		
## 1	NA	NA	NA		
## 2	NA	NA	NA		
## 3	NA	NA	NA		
## 4	NA	NA	NA		
## 5	NA	NA	NA		
## 6	NA	NA	NA		
## 7	NA	NA	NA		
## 8	NA	NA	NA		
## 9	NA	NA	NA		
## 10	NA	NA	NA		
## 11	NA	NA	NA		
## 12	NA	NA	NA		
## 13	NA	NA	NA		
## 14	NA	NA	NA		
## 15	NA	NA	NA		
## 16	NA	NA	NA		
## 17	NA	NA	NA		
## 18	NA	NA	NA		
## 19	NA	NA	NA		
## 20	NA	NA	NA		

## Exercise 4 Descriptive Characteristics

##	Rank	Cutoff_mean	Cutoff_sd	Quality_mean	Quality_sd	Distance_mean	Distance_sd
## 1	1	284.5812	59.705298	311.1536	52.96497	34.78904	52.23429
## 2	2	277.7861	51.430078	303.6828	44.73330	33.67427	47.81538
## 3	3	262.6396	43.985059	289.9210	37.49325	28.25628	42.75935
## 4	4	249.4498	38.069156	278.4302	31.91191	22.62548	38.28634
## 5	5	210.3753	8.185402	251.9085	12.88347	31.78886	29.18945
## 6	6	210.3297	8.582465	249.4862	11.20343	31.16226	28.54966
##	Interval	Cutoff_mean	Cutoff_sd	Quality_mean	Quality_sd	Distance_mean	
## 1	0-25%	223.77	17.45	245.6	7.55	193.69	
## 2	25-50%	272.99	9.71	270.55	9.47	272.18	
## 3	50-75%	306.67	11.59	307.82	11.61	305.79	
## 4	75-100%	360.07	22.63	366.6	26	360.79	
##	Distance_sd						
## 1	25.23						

```
## 2      11.01
## 3      11.7
## 4      19.75
```

## Part 2

### Exercise 5 Data creation

```
## ydum
##      0      1
## 4382 5618
```

Data have been created.

### Exercise 6 OLS

```
## [1] "corr(Y,X1) = 0.21601496838707"
```

Now, the correlation between Y and X1 is quite different from 1.2, the designated coefficient. But the coefficient of X1 is very close to 1.2. Besides, if we standardize every variables (Y, X1, X2, X3), then the coefficient of new X1 (X1\_1) will be similar to the correlation between Y and X1.

```
## [1] "after standardization: coeff(X1) = 0.205705404881561"
```

```
## [1] "coefficients:"
```

```
##      [,1]
##      2.4907098
## X1  1.1976226
## X2 -0.8970514
## X3  0.0875850
```

```
## [1] "standard errors:"
```

```
##              X1              X2              X3
##      1.649836e-03 -6.048035e-04 -5.043875e-05 -1.439837e-04
## X1 -6.048035e-04  3.012891e-04  5.557060e-07  1.288086e-06
## X2 -5.043875e-05  5.557060e-07  8.273992e-06 -4.787112e-08
## X3 -1.439837e-04  1.288086e-06 -4.787112e-08  4.706056e-04
```

### Exercise 7 Discrete choice

```
## initial value 24689.277344
## iter  2 value 4289.017097
## iter  3 value 4251.184984
## iter  4 value 4209.797985
## iter  5 value 4198.168216
## iter  6 value 3021.365796
## iter  7 value 2521.019589
## iter  8 value 2303.352493
## iter  9 value 2229.544875
## iter 10 value 2214.628584
## iter 11 value 2214.605447
## iter 12 value 2213.755610
## iter 13 value 2213.464263
## iter 14 value 2213.463096
## iter 15 value 2213.334779
## iter 16 value 2213.313310
```

```

## iter 16 value 2213.313307
## iter 16 value 2213.313307
## final value 2213.313307
## converged

## initial value 12337.718958
## iter 2 value 3813.743197
## iter 3 value 3203.783208
## iter 4 value 3139.556932
## iter 5 value 3124.746731
## iter 6 value 2420.100956
## iter 7 value 2293.654430
## iter 8 value 2261.396203
## iter 9 value 2235.625304
## iter 10 value 2224.693119
## iter 11 value 2224.583551
## iter 12 value 2223.255132
## iter 13 value 2223.227459
## iter 14 value 2223.209703
## iter 15 value 2223.081465
## iter 16 value 2223.017353
## iter 16 value 2223.017344
## iter 16 value 2223.017344
## final value 2223.017344
## converged

##      Probit:est  Probit:se  Probit:t-value  Probit:p-value  Logit:est
## cons  3.04275799  0.10007791    30.4038917  2.826563e-194  5.42656128
## X1    1.17235964  0.04292123    27.3142131  1.712261e-158  2.10060104
## X2   -0.90546589  0.01858996   -48.7072561  0.000000e+00 -1.61851270
## X3   -0.01124976  0.04647615    -0.2420544  8.087430e-01 -0.01963054
##      Logit:se  Logit:t-value  Logit:p-value
## cons  0.18557828    29.2413603  1.965090e-180
## X1    0.07936254    26.4684202  2.869053e-149
## X2    0.03670968   -44.0895371  0.000000e+00
## X3    0.08323300   -0.2358504  8.135536e-01

```

Both probit and logit model predict that higher X1 yields higher probability of being ydum=1 and higher X2 or X3 decrease the probability of being ydum=1. According to the p-value under both models, the coefficients of X1,X2 are all significant and the X3 is not significant.

## Exercise 8 Marginal Effects

```

##      margin.probit.avg  margin.probit.mean  margin.logit.avg  margin.logit.mean
## cons      1.709663378      1.495066090      3.04864284      2.657750966
## X1        0.658724865      0.576041589      1.18011794      1.028805195
## X2       -0.508762736     -0.444902736     -0.90928065     -0.792694205
## X3       -0.006321007     -0.005527593     -0.01102844     -0.009614392

```

Note: “avg” means evaluating average marginal effects in the sample; “mean” means evaluating marginal effect at the mean.