

markdown

Part 1

1.1. Find descriptive statistics of the data and summarize them into a table

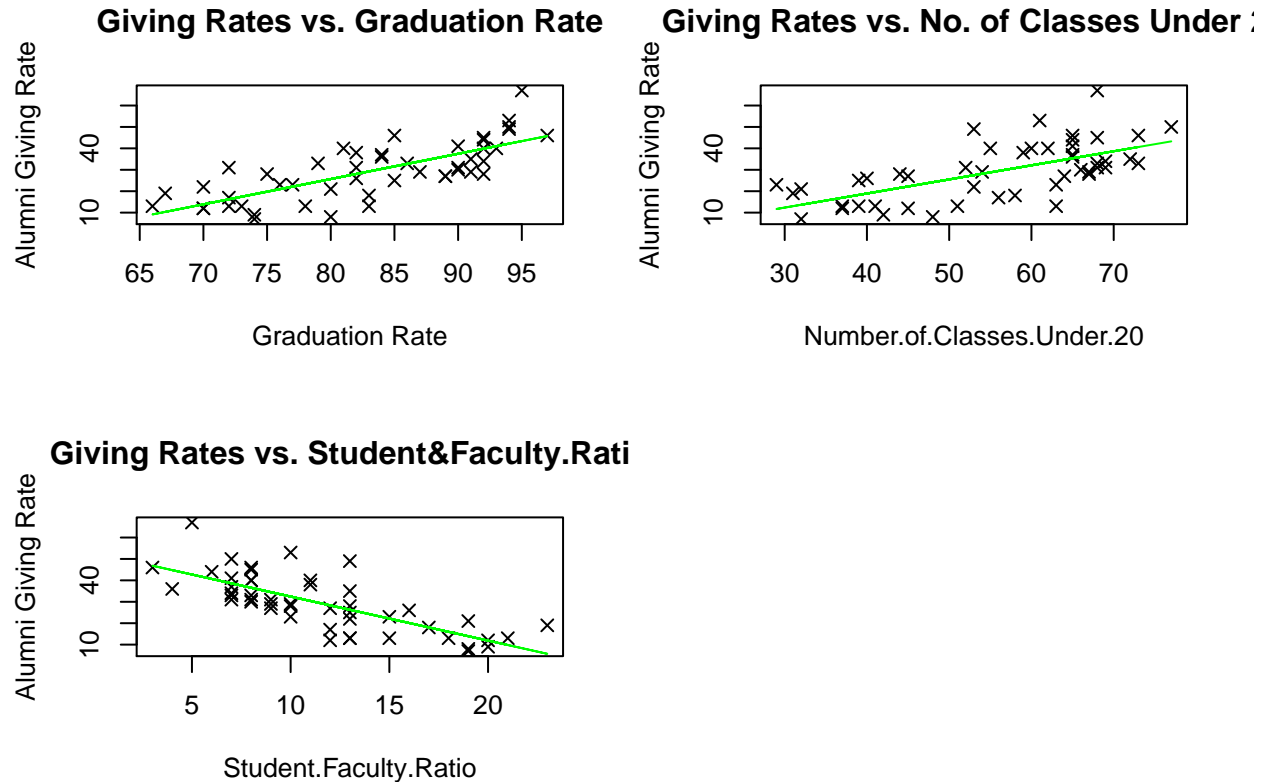
The table is constructed as follows.

##	Graduation.Rate	Number.of.Classes.Under.20	Student.Faculty.Ratio
## 1	85	39	13
## 2	79	68	8
## 3	93	60	8
## 4	85	65	3
## 5	75	67	10
## 6	72	52	8
## 7	89	45	12
## 8	90	69	7
## 9	91	72	13
## 10	94	61	10
## 11	92	68	8
## 12	84	65	7
## 13	91	54	10
## 14	97	73	8
## 15	89	64	9
## 16	81	55	11
## 17	92	65	6
## 18	72	63	13
## 19	90	66	8
## 20	80	32	19
## 21	95	68	5
## 22	92	62	8
## 23	92	69	7
## 24	87	67	9
## 25	72	56	12
## 26	83	58	17
## 27	74	32	19
## 28	74	42	20
## 29	78	41	18
## 30	80	48	19
## 31	70	45	20
## 32	84	65	4
## 33	67	31	23
## 34	77	29	15
## 35	83	51	15
## 36	82	40	16
## 37	94	53	13
## 38	90	65	7
## 39	76	63	10
## 40	70	53	13
## 41	66	39	21
## 42	92	44	13
## 43	70	37	12
## 44	73	37	13
## 45	82	68	9

## 46	82	59	11
## 47	86	73	7
## 48	94	77	7
##	Alumni.Giving.Rate		
## 1	25		
## 2	33		
## 3	40		
## 4	46		
## 5	28		
## 6	31		
## 7	27		
## 8	31		
## 9	35		
## 10	53		
## 11	45		
## 12	37		
## 13	29		
## 14	46		
## 15	27		
## 16	40		
## 17	44		
## 18	13		
## 19	30		
## 20	21		
## 21	67		
## 22	40		
## 23	34		
## 24	29		
## 25	17		
## 26	18		
## 27	7		
## 28	9		
## 29	13		
## 30	8		
## 31	12		
## 32	36		
## 33	19		
## 34	23		
## 35	13		
## 36	26		
## 37	49		
## 38	41		
## 39	23		
## 40	22		
## 41	13		
## 42	28		
## 43	12		
## 44	13		
## 45	31		
## 46	38		
## 47	33		
## 48	50		

1.2. Use graphical analysis to investigate the relationship between Alumni Giving Rate and each of the other variables

Scatter plot is demonstrated as follows. In the first and the third graphs, we can find that points cluster closely around fitted lines. While in the second graph, points seem to drift away from fitted line. From the graphical analysis, we can reasonably assume that alumni giving rate is more closely related to both graduation rate and faculty rate than number of classes under 20.



1.3. Develop a multiple linear regression model that could be used to predict the Alumni Giving Rate using the data provided

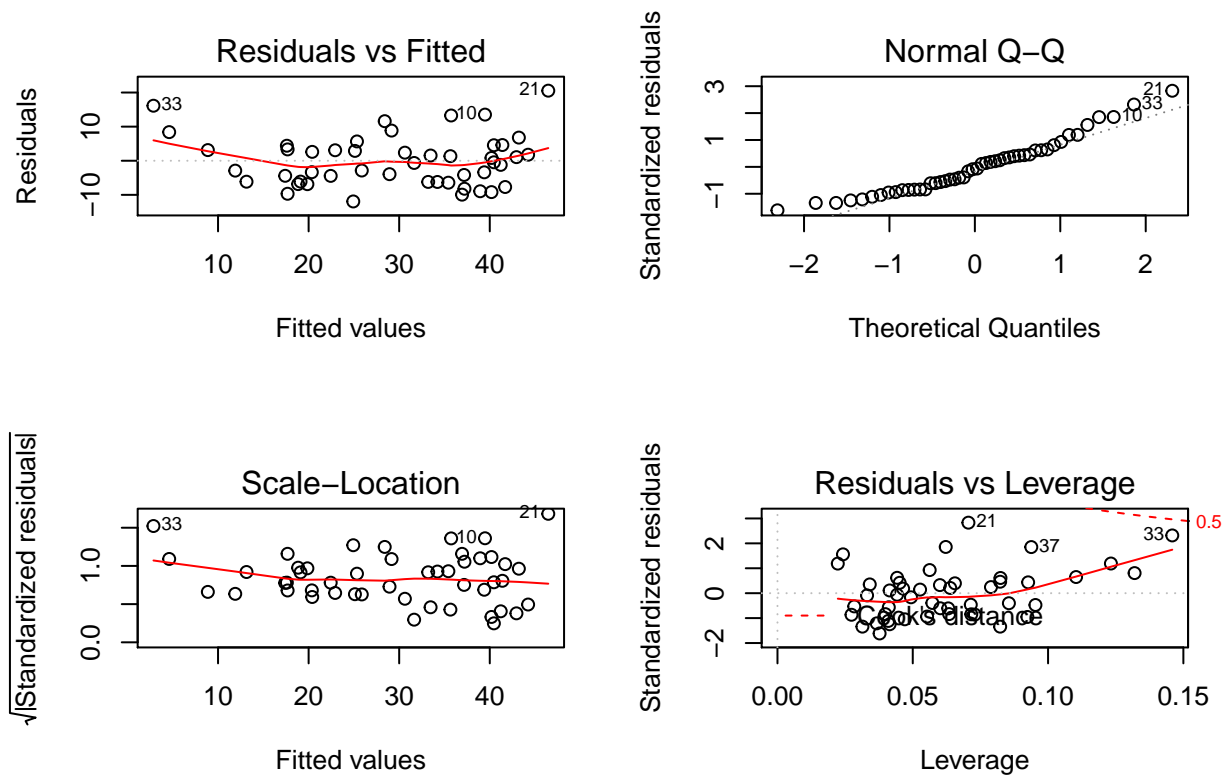
We can use function **stepAIC** to construct the best multi-linear regression model from a set of candidate variables. The experiment result indicates that, AIC value becomes smaller when the variable “Number of classes under 20” is deleted from the regression model. Since smaller AIC value means better fitting effect, we should construct a regression model with “Graduation Rates” and “Student & Faculty Ratios” as independent variables.

```
## Start:  AIC=198.65
## Alumni.Giving.Rate ~ Graduation.Rate + Number.of.Classes.Under.20 +
##      Student.Faculty.Ratio
##
##              Df Sum of Sq  RSS   AIC
## - Number.of.Classes.Under.20  1      2.52 2550.5 196.70
## <none>                                2547.9 198.65
## - Student.Faculty.Ratio          1     550.17 3098.1 206.03
## - Graduation.Rate                 1     1176.92 3724.9 214.88
##
## Step:  AIC=196.7
```

```
## Alumni.Giving.Rate ~ Graduation.Rate + Student.Faculty.Ratio
##
##              Df Sum of Sq    RSS    AIC
## <none>                2550.5 196.70
## - Student.Faculty.Ratio  1    1088.5 3639.0 211.76
## - Graduation.Rate        1    1260.9 3811.4 213.98
##
## Call:
## lm(formula = Alumni.Giving.Rate ~ Graduation.Rate + Student.Faculty.Ratio,
##     data = newdataframe)
##
## Coefficients:
##             (Intercept)      Graduation.Rate  Student.Faculty.Ratio
##             -19.1063              0.7557             -1.2460
##
## Call:
## lm(formula = Alumni.Giving.Rate ~ Graduation.Rate + Student.Faculty.Ratio,
##     data = newdataframe)
##
## Coefficients:
##             (Intercept)      Graduation.Rate  Student.Faculty.Ratio
##             -19.1063              0.7557             -1.2460
##
## Call:
## lm(formula = Alumni.Giving.Rate ~ Graduation.Rate + Student.Faculty.Ratio,
##     data = newdataframe)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -11.9304  -6.1594  -0.5521   3.5910  20.5412
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -19.1063    15.5501  -1.229   0.226
## Graduation.Rate    0.7557     0.1602   4.717 2.35e-05 ***
## Student.Faculty.Ratio -1.2460     0.2843  -4.382 6.95e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 7.528 on 45 degrees of freedom
## Multiple R-squared:  0.6996, Adjusted R-squared:  0.6863
## F-statistic: 52.41 on 2 and 45 DF,  p-value: 1.765e-12
```

1.4. Check the model assumptions

In Q-Q graph, points closely cluster around the line, which proves that the assumption of **normality** is satisfied; there is no reason to assume that graduation ratio and faculty to student ratio is related. Therefore, the assumption of **independence** is satisfied; from graph one, we can observe that residuals is no systematic relationship between residuals and the predicted values. The model well captures systematic variance in the data, thereby proves that the assumption of **Linearity** is satisfied; the Scale-Location graph shows that the points form a random band around the horizontal line. Hence the assumption of **Homoscedasticity** is satisfied.



Part 2

2.1. Calculate the mean of Fertility and partition the provinces into two groups

The mean of fertility in stated provinces is 70.14255. Group1 and group2 is illustrated as follows.

```
## [1] 70.14255
```

	Fertility	Agriculture	Examination	Education	Catholic
## Courtelary	80.2	17.0	15	12	9.96
## Delemont	83.1	45.1	6	9	84.84
## Franches-Mnt	92.5	39.7	5	5	93.40
## Moutier	85.8	36.5	12	7	33.77
## Neuveville	76.9	43.5	17	15	5.16
## Porrentruy	76.1	35.3	9	7	90.57
## Broye	83.8	70.2	16	7	92.85
## Glane	92.4	67.8	14	8	97.16
## Gruyere	82.4	53.3	12	7	97.67
## Sarine	82.9	45.2	16	13	91.38
## Veveyse	87.1	64.5	14	6	98.61
## Grandson	71.7	34.0	17	8	3.30
## Oron	72.5	71.2	12	1	2.40
## Payerne	74.2	58.1	14	8	5.23
## Paysd'enhaut	72.0	63.5	6	3	2.56
## Conthey	75.5	85.9	3	2	99.71
## Herens	77.3	89.7	5	2	100.00
## Martigwy	70.5	78.2	12	6	98.96

## Monthey	79.4	64.9	7	3	98.22
## Sierre	92.2	84.6	3	3	99.46
## Sion	79.3	63.1	13	13	96.83
## Boudry	70.4	38.4	26	12	5.62
## Le Locle	72.7	16.7	22	13	11.22
## Val de Ruz	77.6	37.6	15	7	4.97
##	Infant.Mortality ynprovinces				
## Courtelary	22.2		1		
## Delemont	22.2		1		
## Franches-Mnt	20.2		1		
## Moutier	20.3		1		
## Neuveville	20.6		1		
## Porrentruy	26.6		1		
## Broye	23.6		1		
## Glane	24.9		1		
## Gruyere	21.0		1		
## Sarine	24.4		1		
## Veveyse	24.5		1		
## Grandson	20.0		1		
## Oron	21.0		1		
## Payerne	23.8		1		
## Paysd'enhaut	18.0		1		
## Conthey	15.1		1		
## Herens	18.3		1		
## Martigwy	19.4		1		
## Monthey	20.2		1		
## Sierre	16.3		1		
## Sion	18.1		1		
## Boudry	20.3		1		
## Le Locle	18.9		1		
## Val de Ruz	20.0		1		
##	Fertility Agriculture Examination Education Catholic				
## Aigle	64.1	62.0	21	12	8.52
## Aubonne	66.9	67.5	14	7	2.27
## Avenches	68.9	60.7	19	12	4.43
## Cossonay	61.7	69.3	22	5	2.82
## Echallens	68.3	72.6	18	2	24.20
## Lausanne	55.7	19.4	26	28	12.11
## La Vallee	54.3	15.2	31	20	2.15
## Lavaux	65.1	73.0	19	9	2.84
## Morges	65.5	59.8	22	10	5.23
## Moudon	65.0	55.1	14	3	4.52
## Nyone	56.6	50.9	22	12	15.14
## Orbe	57.4	54.1	20	6	4.20
## Rolle	60.5	60.8	16	10	7.72
## Vevey	58.3	26.8	25	19	18.46
## Yverdon	65.4	49.5	15	8	6.10
## Entremont	69.3	84.9	7	6	99.68
## St Maurice	65.0	75.9	9	9	99.06
## La Chauxdfnd	65.7	7.7	29	11	13.79
## Neuchatel	64.4	17.6	35	32	16.92
## ValdeTravers	67.6	18.7	25	7	8.65
## V. De Geneve	35.0	1.2	37	53	42.34

## Rive Droite	44.7	46.6	16	29	50.43
## Rive Gauche	42.8	27.7	22	29	58.33
##	Infant.Mortality	ynprovinces			
## Aigle	16.5	0			
## Aubonne	19.1	0			
## Avenches	22.7	0			
## Cossonay	18.7	0			
## Echallens	21.2	0			
## Lausanne	20.2	0			
## La Vallee	10.8	0			
## Lavaux	20.0	0			
## Morges	18.0	0			
## Moudon	22.4	0			
## Nyone	16.7	0			
## Orbe	15.3	0			
## Rolle	16.3	0			
## Vevey	20.9	0			
## Yverdon	22.5	0			
## Entremont	19.8	0			
## St Maurice	17.8	0			
## La Chauxdfnd	20.5	0			
## Neuchatel	23.0	0			
## ValdeTravers	19.5	0			
## V. De Geneve	18.0	0			
## Rive Droite	18.2	0			
## Rive Gauche	19.3	0			
##	[,1]				
## [1,]	1				
## [2,]	1				
## [3,]	1				
## [4,]	1				
## [5,]	1				
## [6,]	1				
## [7,]	1				
## [8,]	1				
## [9,]	1				
## [10,]	1				
## [11,]	1				
## [12,]	0				
## [13,]	0				
## [14,]	0				
## [15,]	0				
## [16,]	0				
## [17,]	1				
## [18,]	0				
## [19,]	0				
## [20,]	0				
## [21,]	0				
## [22,]	0				
## [23,]	0				
## [24,]	0				
## [25,]	1				
## [26,]	1				

```
## [27,] 1
## [28,] 0
## [29,] 0
## [30,] 0
## [31,] 1
## [32,] 0
## [33,] 1
## [34,] 1
## [35,] 1
## [36,] 0
## [37,] 1
## [38,] 1
## [39,] 1
## [40,] 0
## [41,] 1
## [42,] 0
## [43,] 1
## [44,] 0
## [45,] 0
## [46,] 0
## [47,] 0
```

2.2. Use logistic regression to show the relationship between y and the other variables and then interpret the regression results

The experiment result indicates that fertility in selected provinces is significantly related to **Agriculture** and **Examination** under significance level of 0.05. It implies that fertility is closely related to agriculture situation and examination circumstance in the provinces.

```
##
## Call:
## glm(formula = y ~ Agriculture + Examination + Education + Catholic +
##      Infant.Mortality, family = binomial(), data = swiss)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -1.85403  -0.45960   0.03648   0.55548   2.32911
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    4.82826    5.25607   0.919  0.3583
## Agriculture    -0.09615    0.04011  -2.397  0.0165 *
## Examination    -0.32116    0.13844  -2.320  0.0203 *
## Education      -0.12078    0.08610  -1.403  0.1607
## Catholic        0.02078    0.01376   1.509  0.1312
## Infant.Mortality 0.29078    0.21051   1.381  0.1672
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 65.135  on 46  degrees of freedom
## Residual deviance: 32.887  on 41  degrees of freedom
## AIC: 44.887
##
```



```
## Number of Fisher Scoring iterations: 6
```

2.3. Choose a model selection criterion, for instances, AIC, BIC, adjusted R square or Cp, and use it to select a reasonable model

We can construct the best multi-linear regression model from a set of candidate variables. Under AIC criterion, regression model with smaller AIC value is considered better. The experiment result indicates that AIC value gets smaller when the variables “Catholic” and “Education” are deleted from the regression model. Hence we should construct a regression model with “**Infant.Mortality**”, “**Agriculture**” and “**Examination**” as independent variables.

```
## Start:  AIC=-84.75
## y ~ Agriculture + Examination + Education + Catholic + Infant.Mortality
##
##              Df Sum of Sq    RSS    AIC
## - Education      1   0.25537  6.2545 -84.792
## - Infant.Mortality 1   0.25579  6.2549 -84.789
## <none>                        5.9991 -84.751
## - Catholic       1   0.40468  6.4038 -83.683
## - Examination    1   0.93170  6.9308 -79.966
## - Agriculture    1   1.06427  7.0634 -79.075
##
## Step:  AIC=-84.79
## y ~ Agriculture + Examination + Catholic + Infant.Mortality
##
##              Df Sum of Sq    RSS    AIC
## - Catholic       1   0.20534  6.4598 -85.274
## <none>                        6.2545 -84.792
## - Infant.Mortality 1   0.36465  6.6191 -84.129
## - Agriculture    1   0.82599  7.0805 -80.962
## - Examination    1   2.43417  8.6887 -71.342
##
## Step:  AIC=-85.27
## y ~ Agriculture + Examination + Infant.Mortality
##
##              Df Sum of Sq    RSS    AIC
## <none>                        6.4598 -85.274
## - Infant.Mortality 1   0.4546   6.9144 -84.077
## - Agriculture    1   0.7943   7.2542 -81.823
## - Examination    1   3.7161  10.1760 -65.916
##
## Call:
## lm(formula = y ~ Agriculture + Examination + Infant.Mortality,
##     data = swiss)
##
## Coefficients:
##      (Intercept)      Agriculture      Examination  Infant.Mortality
##           1.05056          -0.00811           -0.05017            0.03501
##
## Call:
## lm(formula = y ~ Agriculture + Examination + Infant.Mortality,
##     data = swiss)
##
## Coefficients:
```

##	(Intercept)	Agriculture	Examination	Infant.Mortality
##	1.05056	-0.00811	-0.05017	0.03501