# **Project Report**

(MATH 584 Applied Statistics)

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## Part 1

#### 1) Use OLS to estimate Regression coefficients.

 Based on the summary output using OLS, Regression coefficients of each of predictors are as follows in the Red box:

			gressio =====				
Dep. Varia	ble:		BIO R	-squa	red:		0.823
Model:			OLS A	dj. R-	-squared:		0.734
Method:		Least Squa	res F	-stati	istic:		9.270
Date:	TI	hu, 07 Dec 2	023 P	rob (f	-statisti	.c):	4.03e-07
Γime:		11:46	:30 L	Log-Likelihood:			-302.70
No. Observ	ations:		43 A	IC:			635.4
of Residua	ls:		28 B	IC:			661.8
f Model:			14				
Covariance	Type:	nonrob	ust				
	coef	std err		t	P> t	[0.025	0.975]
onst	3475.9507	3441.050	1.0	 10	0.321	-3572.720	1.05e+04
125	1.1544	3.048	0.3	79	0.708	-5.089	7.398
SAL	-19.2305	26.581	-0.7	23	0.475	-73.679	35.218
h7	2.4120	1.964	1.2	28	0.230	-1.612	6.435
Н	149.1615	330.050	0.4	52	0.655	-526.915	825.238
BUF	-19.6909	121.063	-0.1	63	0.872	-267.676	228.295
	-6.1819	3.854	-1.6	04	0.120	-14.077	1.713
(	-1.0168	0.474	-2.1	44	0.041	-1.988	-0.045
Ca	-0.0657	0.125	-0.5	24	0.604	-0.323	0.191
1g	-0.3667	0.273	-1.3	43	0.190	-0.926	0.192
la .	0.0100	0.024	0.4	11	0.684	-0.040	0.060
1n	-3.6814	5.513	-0.6	68	0.510	-14.975	7.612
<u>'</u> n	-8.0818	21.989	-0.3	68	0.716	-53.125	36.961
Cu	373.8948	110.351	3.3	88	0.002	147.852	599.938
IH4	-1.5510	3.219	-0.4	82	0.634	-8.145	5.043
mnibus:		10.	===== 120 D	urbin-	 -Watson:		1.791
Prob(Omnib	us):				-Bera (JB)	:	14.888
kew:			602 P				0.000585
urtosis:		5.	619 C	ond. N	No.		1.22e+06

#### 2) Run Collinearity diagnostics (VIF, Condition Index)

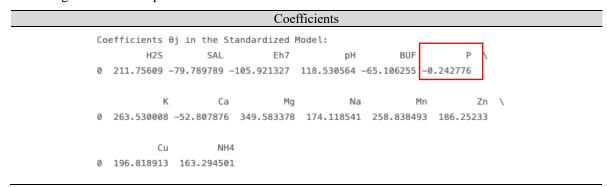
- Looking at the result from the two Collinearity diagnostics method, it is observed from VIF that there are 6 predictors that exceeds 10, meaning there presents serious multicollinearity (pH, BUF, Ca, Mg, Na, Zn). From Condition Index, since there are no predictors that exceeds 30, we see no serious collinearity, but it suggests that there exists collinearity from the predictor NH4 for it is exceeding 15.
- Below is the code output of 'VIF' and 'Condition Index' to check collinearity of the predictors:

VIF			Condition Index		
Predictors	VIF		Predictors	Condition Index	
const	4350.771896		H2S	1.000000	
H2S	3.136506		SAL	1.184103	
SAL	3.361283		Eh7	1.791479	
Eh7	1.964076		pH	2.048414	
pH	62.564383		BUF	2.733624	
BUF	33.478422				
Р	2.884226		P	3.241408	
K	7.432133		K	3.696482	
Ca	17.343432		Ca	4.447720	
Mg	24.476419		Mg	5.687906	
Na	10.372624		Na	6.009633	
Mn	6.737786		Mn	7.842823	
			Zn	10.674971	
Zn	12.391033		Cu	13.561188	
Cu	4.866983		NH4	23.308398	
NH4	8.586275				
Thresholds: - VIF = 1 : Best case - VIF between 4 & 10 : Needs further investigation - VIF > 10 : Serious Multicollinearity			Thresholds (Where, $k$ - $k \ge 15$ : Collinearity - $k \ge 30$ : Serious Mu	y exists	

#### Part 2

#### 1) Use Principal Components Regression (PCR) method with collinearity reduction.

- The collinearity reduction was conducted based on value of Coefficients, Explained Variance, and Condition Index.
  - i. If the Coefficient is close to zero, it means there is no significance in the model, so it is better to be excluded.
  - ii. Explained Variance is the proportion of the total variability in a dataset that is accounted for by the statistical model, similar concept as R-squared. If the Explained Variance is low, it can be also considered to be excluded from the model.
  - iii. If Condition Index is greater than 15, it means there exists collinearity, hence the predictor will be excluded from the model. (Explained Variance here is used only to double check but solely used to select predictor)
- Following is the code output:



Explained Variance	Condition Index
Explained Variance  Explained Variance  H2S	Condition Index of Each Dringing Component.

- Looking at the result, it is observed that predictor 'P' has insignificant coefficient and its Explained Variance is also low. There are other predictors whose Explained Variance are also very low, but their coefficients are not considered as insignificant. Also, predictor 'NH4' condition index is 23 which is greater than 15, there can be collinearity exists. so will exclude.
- Excluded predictors: P, NH4
   Remaining predictors: H2S, SAL, Eh7, pH, BUF, K, Ca, Mg, Na, Mn, Zn, Cu
- 2) Compare the standard error sum  $\sum_i s. e(\widehat{\beta}_i)$  and SSE with their counterparts in Part I
  - From Part 1, we have  $\sum_i s. e(\widehat{\beta}_i)$  and SSE for Full model as follows:

```
Standard Error Sum (SSE): 3276740.2803900684
Sum of Standard Errors of Coefficients: 628.529003575566
```

• If we apply the previous result to exclude predictor to make Reduced model, we can get the value of  $\widehat{\beta}_{j}$ , and SSE as follows:

```
Sum of Squared Errors (SSE): 3287657.542488552 Standard Errors of the Coefficients: s.e.(\hat{\beta}_{-}1): 18.541127672701244 s.e.(\hat{\beta}_{-}2): 21.954607722598457 s.e.(\hat{\beta}_{-}3): 33.21604517681828 s.e.(\hat{\beta}_{-}4): 37.979912316268276 s.e.(\hat{\beta}_{-}5): 50.68446649294327 s.e.(\hat{\beta}_{-}6): 68.53694945916295 s.e.(\hat{\beta}_{-}7): 82.46574301595605 s.e.(\hat{\beta}_{-}8): 105.46018376016357 s.e.(\hat{\beta}_{-}9): 111.42536934798409 s.e.(\hat{\beta}_{-}10): 145.4147795921936 s.e.(\hat{\beta}_{-}11): 197.92599540395935 s.e.(\hat{\beta}_{-}12): 251.4397114448181
```

Let's compare standard error sum  $\sum_{j} s. e(\widehat{\beta}_{j})$  and SSE each both from Full model and Reduced model:

```
Full Model - SSE: 3276740.280390065 , Sum of Standard Errors: 1614.6207173853195
Reduced Model - SSE: 3287657.542488552 , Sum of Standard Errors: 1125.0448914055673
```

• We can conclude that SSE from Full model is smaller than that of Reduced model after excluding predictor 'P' and 'NH4'. On the other hand,  $\sum_j s. e(\widehat{\beta}_j)$  from Full model is larger than that of Reduced model.

#### Part 3

#### **Part 3.1)**

# 3.1.1) Build stepwise regression method (using $\alpha_E = \alpha_R = 0.1$ ) and report each step explicitly.

- The process of Stepwise regression method:
  - i. Start from empty model, with no predictors included.
  - ii. Forward selection: Evaluate all available predictors (SAL, pH, K, Na, Zn) by adding each one individually to the model and calculating the p-value of its coefficient, and then decide which predictor enters the model based on the p-value.
  - iii. Backward selection: Evaluate all included variables to ensure they still have p-values below the threshold, 0.1.
  - iv. Iterate until no more variables meet the criteria for inclusion or exclusion.
- The stepwise regression process output:

Iteration	Notes		
1st Forward step		- Models:	
Add pH	with p-value 1.61671e-09	Y∼pH	
Compared to other variables:		$Y \sim Zn$	
pH 1.616712e-09		Y ~ Na	
Zn 4.126330e-06		Y ~ K	
Na 9.500137e-02		$Y \sim SAL$	
K 2.082624e-01		'mII' autous the medel with levest m	
SAL 6.365223e-01		- 'pH' enters the model with lowest p-value of 1.61671e-09.	
1st Backward step			
No variable dropped. Current mode	- No change, p-value of pH < 0.1.		
pH 1.616712e-09			
2 <sup>nd</sup> Forward step		- Models:	
Add Na	with p-value 0.0142458	$Y \sim pH + Na$	
Compared to other variables:		$Y \sim pH + K$	
Na 0.014246		$Y \sim pH + Zn$	
K 0.026971		$Y \sim pH + SAL$	
Zn 0.272026			
SAL 0.608300		- 'Na' enters the model with lowest p-value of 0.0142458.	
2 <sup>nd</sup> Backward step			
No variable dropped. Current	model p-values:		
pH 4.731149e-10		- No change, p-value of pH and Na < 0.1.	
Na 1.424576e-02			

3 <sup>rd</sup> Forw	vard step	
Best p	ditional predictors enter the model because their p-v p-value among excluded variables: 0.4302895702889037 ded variables and their p-values:	- The iteration stops here, since there is no p-value from predictors below 0.1.
Zn K SAL	0.430290 0.641029 0.844553	- Resulting predictors: 'pH' and 'Na'

• After Including the selected predictors chosen from stepwise function to the model, the p-value of both 'pH' and 'Na' is below the threshold 0.1, hence these two are the final model.

	coef	std err	t	P> t	[0.025	0.975]	
const	-466.3748	279.219	-1.670	0.103	-1030.698	97.948	
рН	400.4547	49.046	8.165	0.000	301.329	499.580	
Na	-0.0227	0.009	-2.563	0.014	-0.041	-0.005	

Therefore, the model we get from this stepwise function is as follows  $(X_1 = pH, X_2 = Na)$ :

$$\hat{Y} = -466.3748 + 400.4547X_1 - 0.0227X_2$$

#### 3.1.2) Check collinearity diagnostics (VIF)

Comparison of VIF from Full model vs. Reduced model (stepwise regression)

Full	model	Reduced model (Stepwise regression)		
Predictors const SAL pH K Na Zn	VIF 420.277700 2.099364 3.327339 2.982513 3.311625 4.309322	Predictors const pH Na	VIF	

 Looking at the result above, the VIF after selecting predictors with stepwise regression shows much smaller value compared to that of predictors from the Full model.

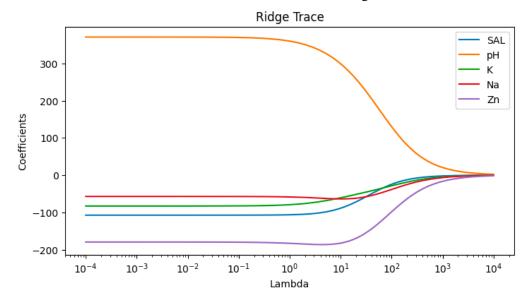
#### **Part 3.2)**

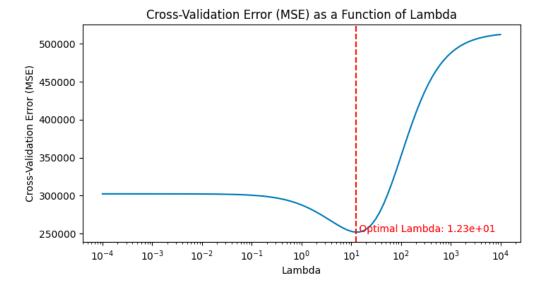
#### 3.2.1) Conduct Ridge Trace and Cross-validate calculating MSE to get optimal $\lambda$ value.

- The process of selecting variable based on Ridge Trace:
  - i. Standardize the predictor variables (X)
  - ii. Compute Ridge Regression for various  $\lambda$  values
  - iii. Plot the Ridge Trace

- iv. Select an appropriate  $\lambda$  based on cross-validation error which is Mean Squared Error(MSE) for each  $\lambda$  values that minimizes the MSE
- Ridge Trace and optimal λ value output:

As seen in the two plots below, we can see that the optimal  $\lambda$  is approximately 12.33. This value represents the best balance between bias and variance for the model, according to the MSE metric used in cross-validation.





#### 3.2.2) Variable Selection:

- Variables can be selected by comparing Ridge coefficients and OLS coefficients. If coefficients both from Ridge and OLS shows noticeable difference, those predictors could be excluded from the model.
- Looking at the result below after applying chosen λ value and calculating back for OLS coefficients, it is observed that predictor 'Na' and 'Zn' shows less difference each other compared to other predictors, hence these two predictors are selected to be included in the model.

Predictors	Ridge Coefficients	OLS Coefficients
SAL	-84.358184	-107.539104
pH	287.982203	370.842749
K	-58.204827	-83.019174
Na	-63.851492	-57.225485
Zn	-179.922429	-179.789802

#### 3.2.3) Check Collinearity Diagnostics (VIF) for selected model:

• Looking at the result above, the VIF after selecting predictors Ridge Trace shows smaller value compared to that of predictors from the Full model.

Full	model	Reduced model (Ridge Trace)		
Predictors const SAL pH K	VIF 420.277700 2.099364 3.327339 2.982513 3.311625	Predictors Constant Na	VIF 1.000000 1.015028 1.015028	
Zn	4.309322			

## **Part 3.3**)

#### 3.3.1) Build Subset Selection method (using BIC and VIF) and report each step explicitly.

- The process of Subset Selection method:
  - i. For a two-variable model, generate all possible pairs of predictors.
  - ii. For each pair of predictors, fit a OLS regression model and calculate BIC and VIF.
  - iii. Identify the model with the lowest BIC value as it suggests a good balance between model complexity and fit. (In case of tie BIC, use VIF)
- The Subset Selection process result:

There are total of 10 subsets (5 combination 2) as possible pairs of predictors. Looking at the result below, we can conclude that the best two variable subset model is 'pH + Na' with the lowest BIC value of 645.8937.

```
The best two-variable model based on the lowest BIC is: pH + Na With BIC: 645.8937195289844 and Max VIF: 1.0005584029303853
```

• More details of the Subset Selection process from the code are shown below output:

#### BIC and VIF value of all possible pairs of predictors

Model: SAL + pH

BIC: 652.1459756701684

Max VIF: 1.0008328341478487

\_\_\_\_\_

Model: SAL + K

BIC: 689.0726713443028

Max VIF: 1.0004265994130863

\_\_\_\_\_

Model: SAL + Na

BIC: 688.0009853893404

Max VIF: 1.0200335426235634

\_\_\_\_\_

Model: SAL + Zn

BIC: 656.5687438839859

Max VIF: 1.2172307994771843

\_\_\_\_\_

Model: pH + K

BIC: 647.106807265751

Max VIF: 1.0008165409299896

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Model: pH + Na

BIC: 645.8937195289844

Max VIF: 1.0005584029303853

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Model: pH + Zn

BIC: 651.1186233056428

Max VIF: 2.1484671152888484

\_\_\_\_\_

Model: K + Na

BIC: 688.0579006245046

Max VIF: 2.7248709843797716

\_\_\_\_\_

Model: K + Zn

BIC: 666.8280215308667

Max VIF: 1.0049129920618014

\_\_\_\_\_

Model: Na + Zn

BIC: 666.0664388199644

Max VIF: 1.0150282539083788

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