2022-1 – Upstart – Research Scientist Phone Interview

Question 1.

When doing the multiple linear regression problem, one column B is wrongly copied as column A. What is the consequence?

Table 1 has the results with correct parameters and Table 2 has the results with incorrect parameter, Administration = Marketing Spend.

Coefficients on Table 2 of columns A and B are very closed to each other, but not identical. Notice the variance matrix has the smallest eigenvalue closed to zero, that means there exists strong multicollinearity.

Table 1.

 $runfile ('/Users/zli/Desktop/Multiple-Linear-Regression/multiple_linear_regression.py', wdir='/Users/zli/Desktop/Multiple-Linear-Regression')$

Intercept:

42554.16761773238

Coefficients:

[7.73467193e-01 3.28845975e-02 3.66100259e-02 -9.59284160e+02 6.99369053e+02]

OLS Regression Results

Profit R-squared: 0.950 Dep. Variable: Model: OLS Adj. R-squared: 0.943 Method: Least Squares F-statistic: 129.7 Fri, 07 Jan 2022 Prob (F-statistic): Date: 3.91e-21 Time: 21:36:42 Log-Likelihood: -421.10 No. Observations: 40 AIC: 854.2 Df Residuals: 34 BIC: 864.3

Df Model: 5

Covariance Type: nonrobust

coef std err

t P>|t| [0.025 0.975]

=====

	COCI	Jiu C	• •		- 141	[0.023	0.575]	
const	4.255	e+04	835	8.538	5.091	0.000	2.56e+04	5.95e+04
R&D Spend		0.773	5 (0.055	14.025	0.000	0.661	0.886
Administrati	on	0.032	9	0.066	0.495	0.624	-0.102	0.168
Marketing S _l	pend	0.03	366	0.019	1.884	4 0.068	-0.003	0.076
Florida	-959.	2842	4038	8.108	-0.238	0.814	-9165.706	7247.138

New York 699.3691 3661.563 0.191 0.850 -6741.822 8140.560

Omnibus: 15.823 Durbin-Watson: 2.468 Prob(Omnibus): 0.000 Jarque-Bera (JB): 23.231

Skew: -1.094 Prob(JB): 9.03e-06 Kurtosis: 6.025 Cond. No. 1.49e+06

Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 1.49e+06. This might indicate that there are strong multicollinearity or other numerical problems.

Table 2.

runfile('/Users/zli/Desktop/Multiple-Linear-Regression/multiple_linear_regression.py', wdir='/Users/zli/Desktop/Multiple-Linear-Regression')

Intercept:

46329.06017854024

Coefficients:

 $[\ 7.85141717e-01\ 1.69781929e-02\ 1.69782094e-02\ -8.26468159e+02$

5.54657333e+02]

OLS Regression Results

Dep. Variable: Profit R-squared: 0.950 Model: OLS Adj. R-squared: 0.944 Method: Least Squares F-statistic: 165.6 Date: Fri, 07 Jan 2022 Prob (F-statistic): 3.19e-22 Time: 21:37:55 Log-Likelihood: -421.24 No. Observations: 40 AIC: 852.5 Df Residuals: 35 BIC: 860.9

Df Model: 4

Covariance Type: nonrobust

=====

coef std err t P>|t| [0.025 0.975]

 const
 4.633e+04
 3375.873
 13.724
 0.000
 3.95e+04
 5.32e+04

 R&D Spend
 0.7851
 0.049
 15.924
 0.000
 0.685
 0.885

 Administration
 0.0170
 0.009
 1.839
 0.074
 -0.002
 0.036

 Marketing Spend
 0.0170
 0.009
 1.838
 0.075
 -0.002
 0.036

 Florida
 -826.4682
 3985.466
 -0.207
 0.837
 -8917.395
 7264.459

New York 554.6573 3610.268 0.154 0.879 -6774.576 7883.891

Omnibus: 14.873 Durbin-Watson: 2.511 Prob(Omnibus): 0.001 Jarque-Bera (JB): 21.150

Skew: -1.038 Prob(JB): 2.56e-05 Kurtosis: 5.895 Cond. No. 1.11e+16

Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The smallest eigenvalue is 3.82e-20. This might indicate that there are strong multicollinearity problems or that the design matrix is singular.

Question 2.

When doing the linear regression, if the dataset is wrongly copied twice. What is the consequence?

The estimated parameter won't change but the confidence interval (or the c.i. range) could shrink by approx sqrt(2). In addition, the R squared won't change but the adjusted R squared changes.

Table 1.

runfile('/Users/zli/Desktop/Multiple-Linear-Regression/multiple_linear_regression.py', wdir='/Users/zli/Desktop/Multiple-Linear-Regression')

Intercept:

45299.49140836343

Coefficients: [0.51986565]

OLS Regression Results

Profit R-squared: Dep. Variable: 0.111 Model: OLS Adj. R-squared: 0.087 Method: Least Squares F-statistic: 4.726 Date: Fri, 07 Jan 2022 Prob (F-statistic): 0.0360 Time: 21:50:10 Log-Likelihood: -478.7440 AIC: No. Observations: 961.5 Df Residuals: 38 BIC: 964.9

Df Model: 1

Covariance Type: nonrobust

====

coef std err t P>|t| [0.025 0.975]

const 4.53e+04 3.02e+04 1.502 0.141 -1.57e+04 1.06e+05 Administration 0.5199 0.239 2.174 0.036 0.036 1.004

Omnibus: 0.124 Durbin-Watson: 1.946 Prob(Omnibus): 0.940 Jarque-Bera (JB): 0.070

Skew: -0.081 Prob(JB): 0.966 Kurtosis: 2.874 Cond. No. 6.14e+05

Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 6.14e+05. This might indicate that there are strong multicollinearity or other numerical problems.

Table 2.

runfile('/Users/zli/Desktop/Multiple-Linear-Regression/multiple_linear_regression.py', wdir='/Users/zli/Desktop/Multiple-Linear-Regression')

Intercept:

45299.491408363414

Coefficients: [0.51986565]

OLS Regression Results

Dep. Variable: Profit R-squared: 0.111 0.099 Model: OLS Adj. R-squared: Method: Least Squares F-statistic: 9.700 Fri, 07 Jan 2022 Prob (F-statistic): Date: 0.00258 21:51:41 Log-Likelihood: Time: -957.48 No. Observations: 80 AIC: 1919. Df Residuals: 78 BIC: 1924.

Df Model: 1

Covariance Type: nonrobust

====

coef std err t P>|t| [0.025 0.975]

const 4.53e+04 2.1e+04 2.153 0.034 3403.607 8.72e+04

Administration 0.5199 0.167 3.115 0.003 0.188 0.852

Omnibus: 0.109 Durbin-Watson: 1.976 Prob(Omnibus): 0.947 Jarque-Bera (JB): 0.140

Skew: -0.081 Prob(JB): 0.932 Kurtosis: 2.874 Cond. No. 6.14e+05

Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 6.14e+05. This might indicate that there are strong multicollinearity or other numerical problems.

2022-3-17 Meta 买它 infra research data scientist 电面

ML/STATISTICS: credit fraud,

Q1: given amount and distance as features, what algorithm you will use?

Answer: Build a classification model to predict probability of fraud.

Q2: what other algorithms you can think of and what are the pro and cons compared to the one you proposed in Q1.

Answer:

2 features -> desicion tree/boosting/deep learning is not adequate.

Decision Tree:

* Not be efficient because lots of data but very few features

KNN:

- * Frauds change over time, not a good patterns as new tech used in the new fraud cases
- * Save all the data but not training needed

Anormaly Dection (to be reviewed):

* Distribution of individual features

Logistic regression:

- * Good interpretability
- * Score fast
- * Training is relatively slow
- * Its relative simplicity makes it a high-bias and low-variance model, so it may not performance well when the decision boundary is not linear.

Q3: coefficient of amount to fraudulence if 0.10 with standard error 0.02, what's the relationship between amount and fraudulence? Is it statistically significant? How do you prove it?

Answer: (See ESL Page 124) Each unit increase in the distance accounts for an increase in the odds of fraudulence of $exp(0.10)^{\sim}=1.105$ or 10.5% (alternatively the increase in the log-odds of

fraud of 0.1 or 10%). The Z score is 0.10/0.02=5 which means the coefficient is significant. The proved by the CLT.	e is