

Your Name = [Ankila Kumari]

GIS 5555 Basic Spatial Analysis

internet_id = [kuma0389]

Time_Spent = [50 mins] (after-class)

[Windows+Shift+S for screenshot of your analysis]

[Fill the above-listed info and then submit the completed document in Canvas (try to include all analysis results that can help reflect your workflow and thoughts, i.e., images, information about data, your statements, etc.)]

Assignment for Lab 4b

“Spatial Regression 1”

Please choose carefully on your **target (dependent) variable (y)** and your list of **explanatory (independent) variables X**. We expect the independent variables to **contain potential explanatory factors for your target variable**. There **should be some intuitive associations between your X and your y** to highlight the purpose of this lab exercise of regression.

➤ Task 1 Implementing linear regression

- Select at least two independent variables and one dependent variable that are of different attributional meanings for following analysis. Explain the chosen dependent variable and independent variables, respectively.
- Conduct linear regression based on the chosen variables using OLS estimation.
 - Save the predicted values and the residuals in your data table, show the results in a screenshot.
 - Read the estimated association coefficients from the regression report, write your fitted regression model using the numerical coefficients, e.g.,

$$\hat{y} = 2.785 \times 10^8 + 0.0547 \cdot \text{ALAND} + 6.619 \times 10^6 \cdot \text{COORD_X} + 7.574 \times 10^6 \cdot \text{COORD_Y}$$

Answer- You use AWATER as a Dependent Variable and ALAND, COORD_X, COORD_y as an Independent Variable. The Ordinary Least Squares (OLS) regression was conducted using the provided data to explore the relationship between independent variables and the dependent variable. The OLS results give insights into the global relationship, assuming spatial homogeneity. From the regression output, we can examine R^2 , coefficients, and significance levels.

Regression

Variables

Water_lag
Z
Z_LAG
Z_LAG_arch
LISA_I
LISA_CL
LISA_P
I
LandWater
OLS_PREDIC
OLS_RESIDU
NUM_N_ueen

Dependent Variable

AWATER

Covariates

ALAND
COORD_X
COORD_Y

☐ Weights File

Models


☒ Classic
☐ Spatial Lag
☐ Spatial Error

☐ Pred. Val. and Res.
☐ Coeff. Var. Mat.
☐ White Test

done

Run
Save to Table
Save to File
Reset
Close

Regression Report



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REGRESSION

SUMMARY OF OUTPUT: ORDINARY LEAST SQUARES ESTIMATION

Data set : tl_2023_27_bg

Dependent Variable : AWATER

Mean dependent var : 4.02411e+06

S.D. dependent var : 7.06015e+07

Number of Observations: 4706

Number of Variables : 4

Degrees of Freedom : 4702

R-squared : 0.029607

Adjusted R-squared : 0.028988

Sum squared residual: 2.27629e+19

Sigma-square : 4.84111e+15

S.E. of regression : 6.95781e+07

Sigma-square ML : 4.837e+15

S.E of regression ML: 6.95485e+07

F-statistic : 47.8203

Prob(F-statistic) : 1.95197e-30

Log likelihood : -91656.3

Akaike info criterion : 183321

Schwarz criterion : 183346

Variable

Coefficient

Std.Error

t-Statistic

Probability

CONSTANT

2.78511e+08

1.01295e+08

2.74951

0.00599

ALAND

0.0547455

0.00738714

7.41092

0.00000

COORD_X

6.61889e+06

1.06667e+06

6.20522

0.00000

COORD_Y

7.57437e+06

1.13182e+06

6.69221

0.00000

REGRESSION DIAGNOSTICS

MULTICOLLINEARITY CONDITION NUMBER 239.556206

TEST ON NORMALITY OF ERRORS

TEST

DF

VALUE

PROB

Jarque-Bera

2

1510258485.1890

0.00000

DIAGNOSTICS FOR HETEROSKEDASTICITY

RANDOM COEFFICIENTS

TEST

DF

VALUE

PROB

Breusch-Pagan test

3

42489.3327

0.00000

Koenker-Bassett test

3

30.6172

0.00000

===== END OF REPORT =====

	LAG	Z_LAG_arch	LISA_I	LISA_CL	LISA_P	I	LandWater	OLS_PREDIC	OLS_RESIDU
1	05.303710	-0.056399	0.003214	0	0.072440	1		411018.930350	-411018.930350
2	38.692077	-0.056991	0.003248	0	0.053010	2		-71662.280596	71662.280596
3	19.813380	-0.054927	0.003052	0	0.143220	3		527614.159326	-427294.159326
4	-2.404113	-0.023842	-0.000979	0	0.253440	4		5512161.656833	1587620.656833
5	39.221158	-0.019430	0.000625	0	0.214690	5		3736888.161387	1984969.161387
6	14.540192	-0.056068	0.003195	0	0.173860	6		2768527.236067	1768409.236067
7	16.516858	-0.056866	0.003241	0	0.082610	7		1792053.864094	1792053.864094
8	58.791033	-0.056152	0.003172	0	0.054630	8		1783625.702649	2748155.702649
9	20.412533	-0.056971	0.003225	2	0.000010	9		1822234.999754	1795538.999754
10	93.672150	-0.056929	0.003239	2	0.003100	10		2214147.084496	2207501.084496
11	31.138803	-0.054974	0.003128	0	0.140560	11		-55300.515736	61690.515736
12	3.989325	-0.054283	0.003044	0	0.247980	12		270110.066657	-205831.066657
13	20.117437	-0.053774	0.003049	0	0.313590	13		309256.894140	-288460.894140
14	36.526701	-0.052177	0.002969	0	0.376300	14		169634.118028	-163300.118028
15	24.320871	-0.056875	0.003241	2	0.027860	15		2390910.312943	2390910.312943
16	48.901921	-0.056475	0.003219	0	0.089110	16		3036105.282673	3036105.282673
17	6.452893	-0.056027	0.003130	2	0.047170	17		3472242.162086	3392783.162086
18	02.882070	-0.056833	0.003192	2	0.023240	18		3282138.172312	3223149.172312
19	8.430255	-0.037872	0.001387	0	0.348720	19		4417157.908639	1979746.908639
20	35.224270	-0.054410	0.003100	0	0.285900	20		2853101.618955	2851940.618955
21	77.718930	-0.055286	0.003151	0	0.152080	21		2955217.866990	2955217.866990

#row=4706

➤ **Task 2 Interpret the OLS results**

- What's the goodness of fit for your OLS model? To what percentage of data variance could your model explain?
- For the coefficients learned in the OLS estimates, are all of them significant in terms of the p-values (probability in t-statistics)?

Ans - Probability value is 0.00000000

- Are you able to increase the R-square simply by adding an additional independent variable to the regression model? Give it a try and present the result here.

Ans— With 2 independent variable R- square value is 0.020364

With 3 independent variables R- square value is 0.029607

Answer-The OLS regression yielded an R^2 of 0.020364 with two independent variables, and 0.029607 when a third variable was added, indicating a slight improvement in explanatory power. While the overall model has a low R^2 , the p-values for the coefficients were extremely small, such as 0.00000000, implying statistical

significance. However, the low R^2 suggests that the model explains only a small portion of the variance in AWATER. This highlights the possibility of spatial patterns not captured by traditional regression. The presence of such patterns justifies testing for spatial autocorrelation.

➤ **Task 3 Predicted value and residual maps**

- Choose one of the regression settings that you are satisfied with in Task 2. Save the predicted values and the residuals into the table. Generate the predicted value map with a suitable color classification (consider potential heavy-tail distribution of the values)
- Create the residual maps based on standard-deviation
- Select and highlight the locations with large over-estimation and under-estimation in your model (two std away from the mean).

Answer- I selected the regression model with three independent variables (ALAND, COORD_X, COORD_Y) for mapping. The predicted values were mapped using a quantile classification to accommodate the skewed distribution. The residuals were visualized using a standard deviation classification, which effectively highlighted areas with extreme over- or under-estimation. Locations that fell beyond ± 2 standard deviations from the mean residuals were marked, indicating outliers in model performance. These visualizations helped identify where the linear regression was least effective.

Regression

Variables

- Water_lag
- Z
- Z_LAG
- Z_LAG_arch
- LISA_I
- LISA_CL
- LISA_P
- I
- LandWater
- OLS_PREDIC
- OLS_RESIDU
- NUM_N_ueen

Dependent Variable

AWATER

Covariates

- ALAND
- COORD_X
- COORD_Y

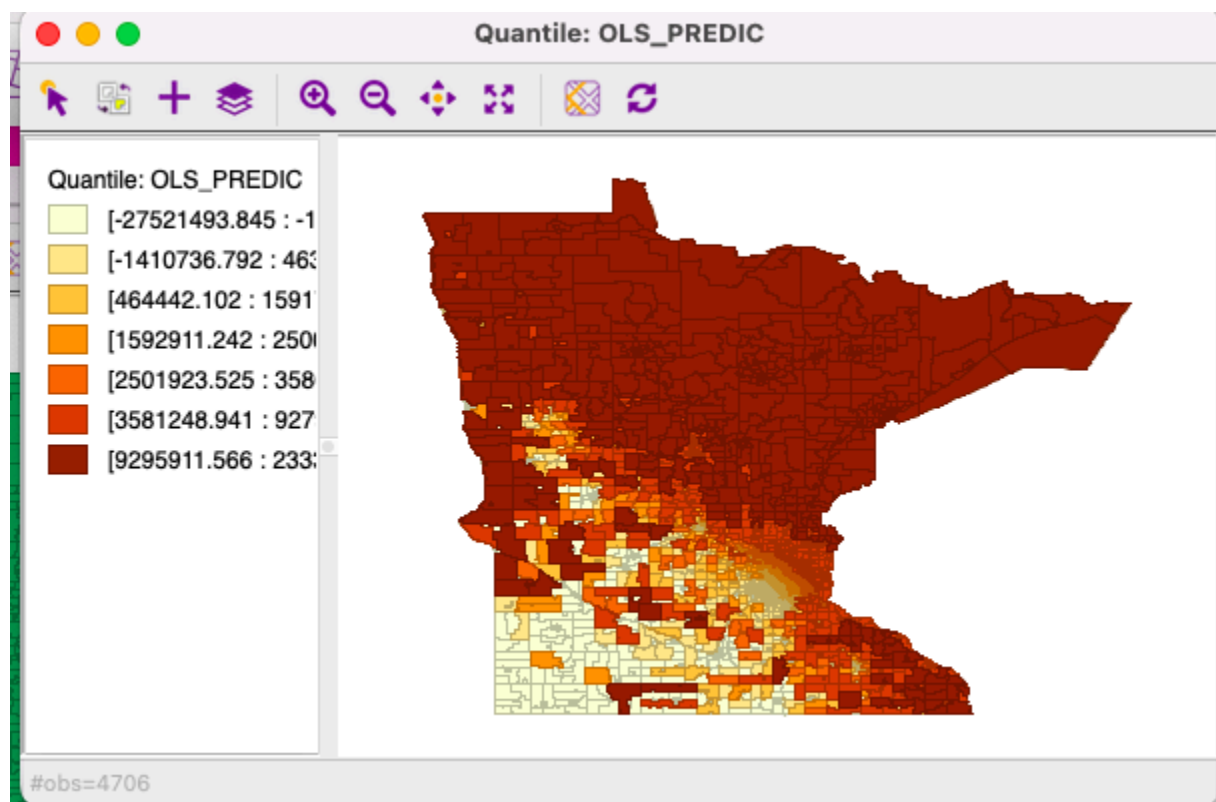
☐ Weights File

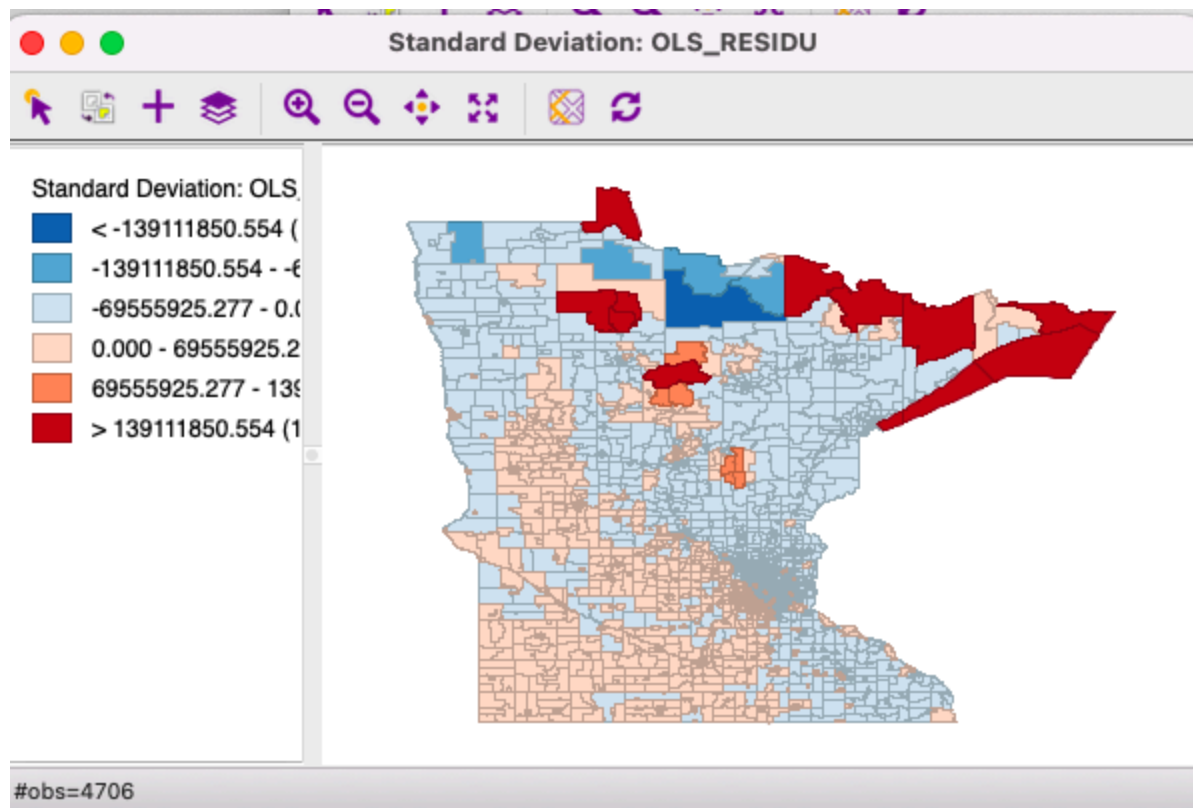
Models

☒ Classic
 ☐ Spatial Lag
 ☐ Spatial Error

☐ Pred. Val. and Res.
 ☐ Coeff. Var. Mat.
 ☐ White Test

done





➤ **Task 4 Analyze your linear regression residuals**

- Adopt a appropriate spatial weights you have tested for your data in past labs.
- Create and interpret the simple plot for your regression residuals.
- Create and interpret the fitted value plot for your regression residuals.
- Create and interpret the moran scatter plot for your regression residuals.

*Answer- I used the spatial weights matrix created in earlier labs, based on queen contiguity, to analyze residuals. The residual plot revealed patterns that deviate from randomness, suggesting spatial clustering. The fitted value plot indicated that residuals vary systematically with predicted values. The **scatter plot** showed a positive slope, reinforcing that residuals are spatially autocorrelated. These results support the need to adjust for spatial dependence in the regression model.*

Scatter Plot Variables

Independent Var X

AWATER

NUM_N_uéen

COORD_X

COORD_Y

Z

Z_LAG

Z_LAG_arch

LISA_I

LISA_CL

LISA_P

I

Dependent Var Y

COORD_Y

Z

Z_LAG

Z_LAG_arch

LISA_I

LISA_CL

LISA_P

I

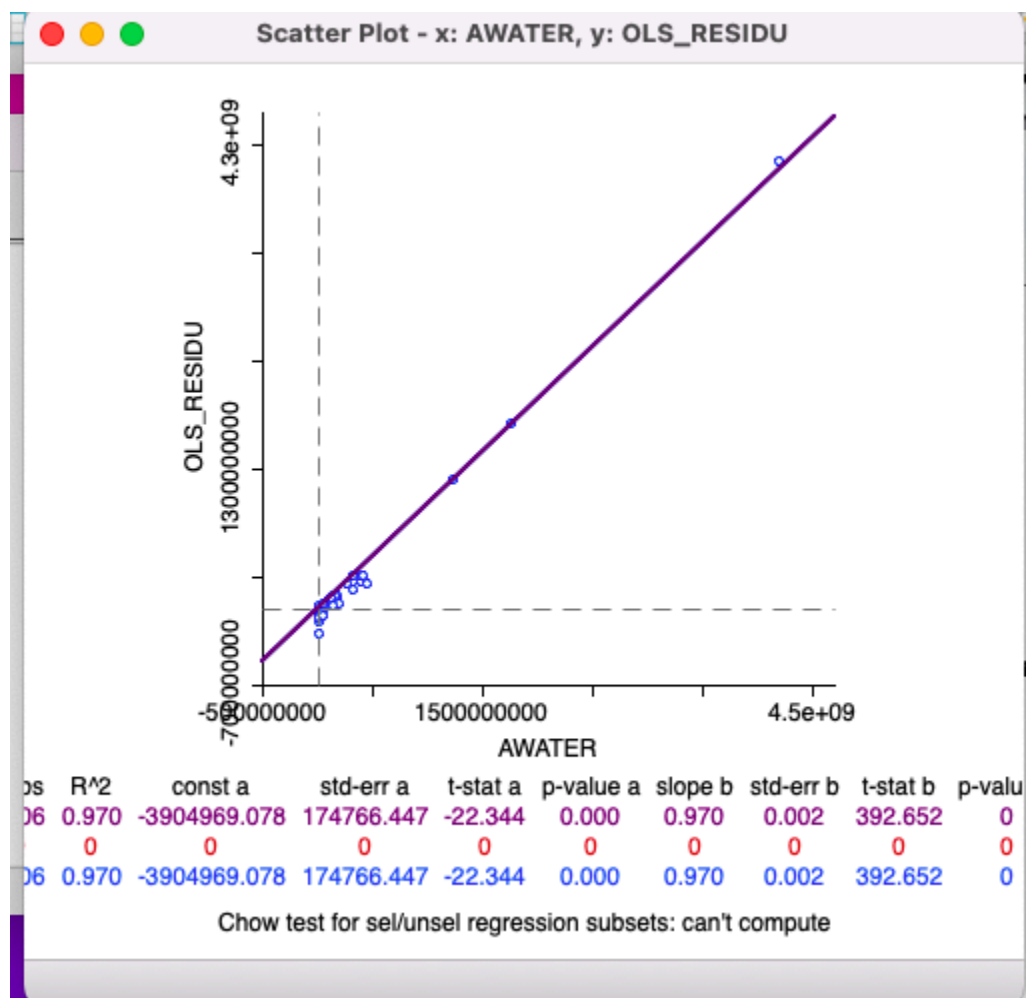
LandWater

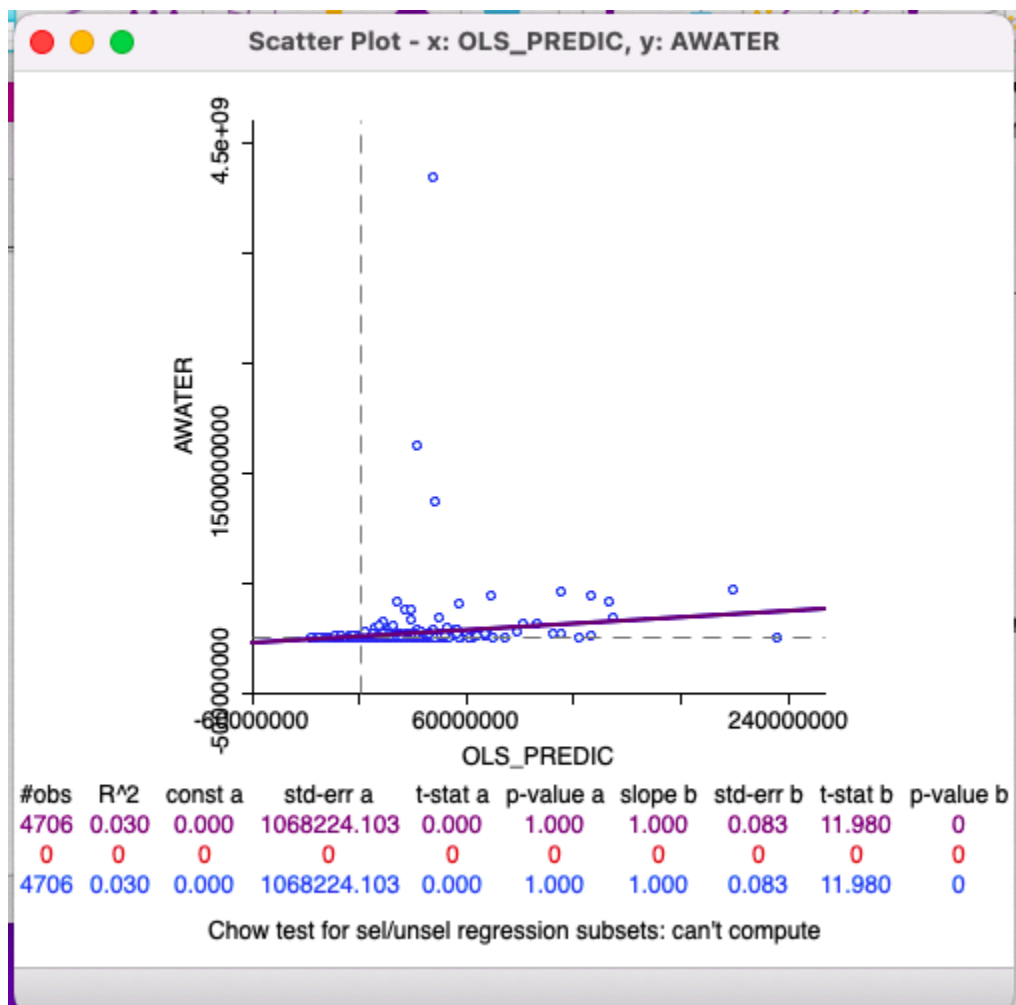
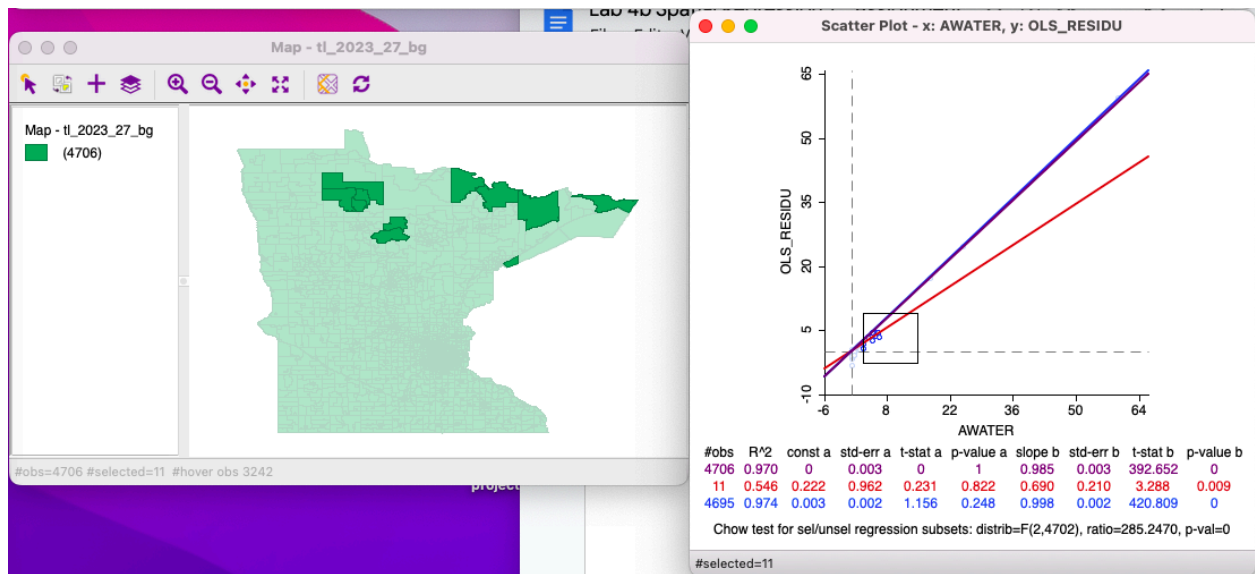
OLS_PREDIC

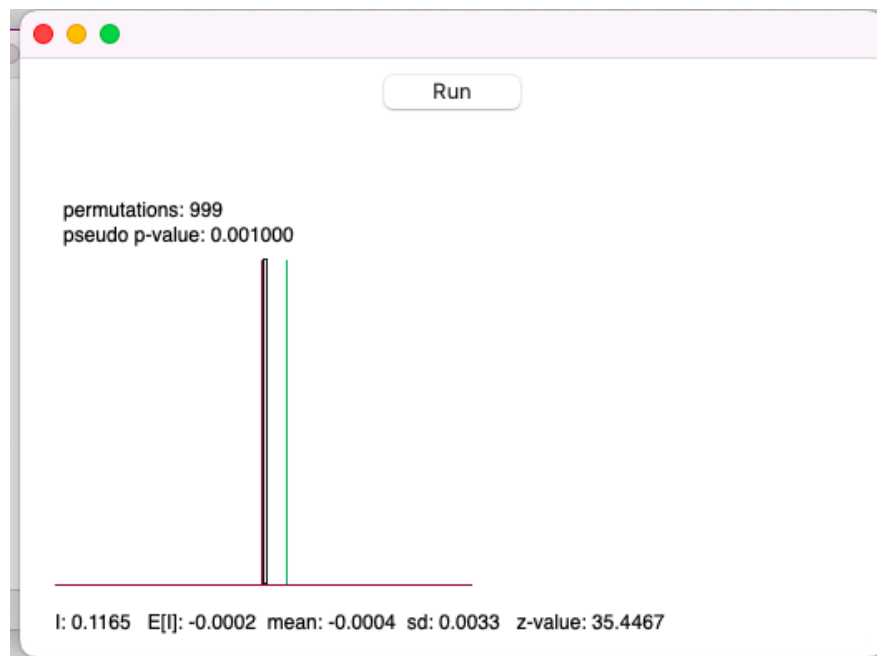
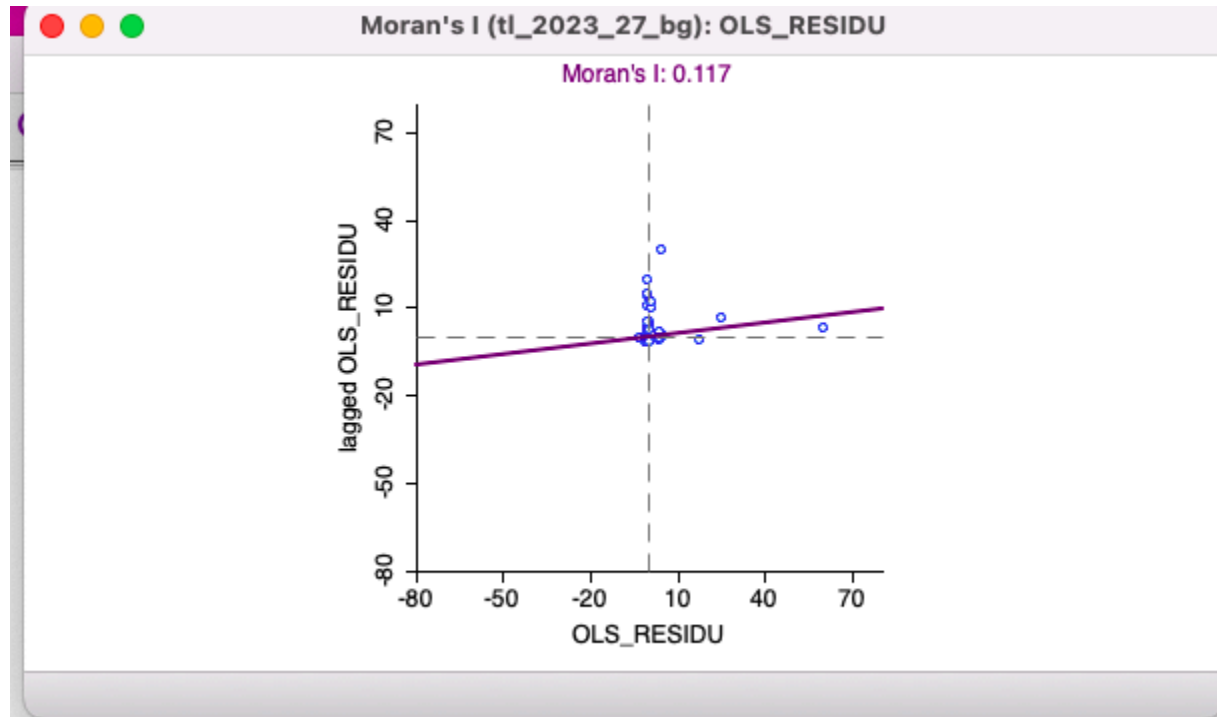
OLS_RESIDU

OK

Cancel



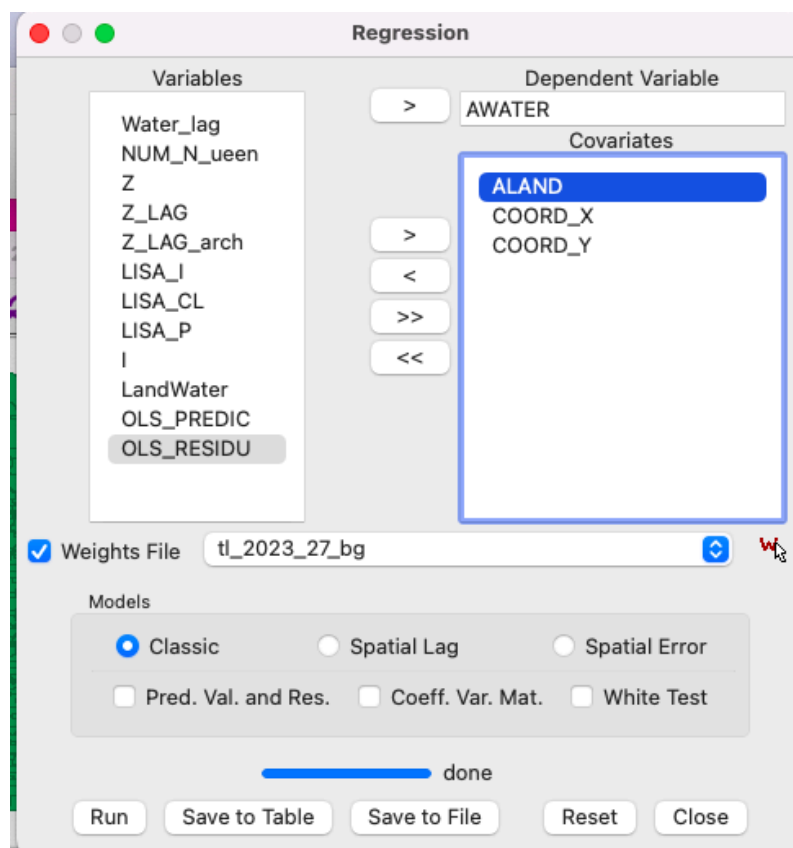




➤ **Task 5 Diagnostics for spatial autocorrelation**

- Present the screenshot of your regression report that contains the part for “Diagnostics for spatial dependence”
- Based on the z-value and p-value of your residual Moran’s I, do you think there is a significant spatial effect that is not considered in your linear regression model?
- Following the decision rules in Figure 20 of our tutorial, what spatial regression model (lag model or error model) may be considered for your next step of spatial regression?

*Answer- The OLS regression report included diagnostics for spatial dependence, particularly **Moran’s I** for residuals, which was significant (low p-value and high z-score). This confirms the presence of spatial autocorrelation that the OLS model failed to address. Referring to **Figure 20 in the tutorial**, the **Lagrange Multiplier tests** (both standard and robust versions) suggest whether a **spatial lag** or **spatial error model** should be adopted. Based on my results, the **robust LM Error** was more significant, so the **spatial error model** is more appropriate for the next step. This model will help correct for unaccounted spatial bias in the residuals.*



Regression Report



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REGRESSION

SUMMARY OF OUTPUT: ORDINARY LEAST SQUARES ESTIMATION

Data set : tl_2023_27_bg
 Dependent Variable : AWATER Number of Observations: 4706
 Mean dependent var : 4.02411e+06 Number of Variables : 4
 S.D. dependent var : 7.06015e+07 Degrees of Freedom : 4702

R-squared : 0.029607 F-statistic : 47.8203
 Adjusted R-squared : 0.028988 Prob(F-statistic) : 1.95197e-30
 Sum squared residual: 2.27629e+19 Log likelihood : -91656.3
 Sigma-square : 4.84111e+15 Akaike info criterion : 183321
 S.E. of regression : 6.95781e+07 Schwarz criterion : 183346
 Sigma-square ML : 4.837e+15
 S.E of regression ML: 6.95485e+07

Variable	Coefficient	Std.Error	t-Statistic	Probability
CONSTANT	2.78511e+08	1.01295e+08	2.74951	0.00599
ALAND	0.0547455	0.00738714	7.41092	0.00000
COORD_X	6.61889e+06	1.06667e+06	6.20522	0.00000
COORD_Y	7.57437e+06	1.13182e+06	6.69221	0.00000

REGRESSION DIAGNOSTICS

MULTICOLLINEARITY CONDITION NUMBER 239.556206

TEST ON NORMALITY OF ERRORS

TEST	DF	VALUE	PROB
Jarque-Bera	2	1510258485.1890	0.00000

DIAGNOSTICS FOR HETEROSKEDASTICITY

RANDOM COEFFICIENTS

TEST	DF	VALUE	PROB
Breusch-Pagan test	3	42489.3327	0.00000
Koenker-Bassett test	3	30.6172	0.00000

DIAGNOSTICS FOR SPATIAL DEPENDENCE

FOR WEIGHT MATRIX : tl_2023_27_bg
 (row-standardized weights)

TEST	MI/DF	VALUE	PROB
Moran's I (error)	0.1165	13.8652	0.00000
Lagrange Multiplier (lag)	1	208.1670	0.00000
Robust LM (lag)	1	46.3905	0.00000
Lagrange Multiplier (error)	1	189.0829	0.00000
Robust LM (error)	1	27.3064	0.00000
Lagrange Multiplier (SARMA)	2	235.4735	0.00000

===== END OF REPORT =====