

Q1 (10 points)

Q1(10 points). Consider the linear regression model

$$y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \epsilon.$$

1(5). The residuals are listed below:

$$\begin{array}{cccccccccccccc} 1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 & 9 & 10 & 11 & 12 & 13 & 14 & 15 \\ 0.2, 0.3, -0.8, -0.8, -0.3, 0.4, 0.1, -0.1, -0.4, -0.7, 0.6, -0.1, -0.1, 0.3, 0.2. \end{array}$$

Do a Durbin-Watson test H_0 : the error terms are not (first-order) autocorrelated. H_a : the error terms are negatively or positively (first-order) autocorrelated. Let $\alpha = 0.1$.

2(5). The residuals are: 0.2, 0.3, -0.8, -0.8, -0.3, calculate $e_{(i)}$, $z_{(i)}$ and construct the normal plot.

$$y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \varepsilon$$

① H_0 : No A.C. (autocorrelated)

H_a : + A.C. or - A.C.

$n = t = 15$, $\alpha = 0.1$, $k-1 = 3-1 = 2$, two Indep. Variables

$$\begin{aligned} T.S. : d &= \frac{\sum_{t=2}^{15} (l_t - l_{t-1})^2}{\sum_{t=1}^{15} l_t^2} \\ &= \frac{(0.3-0.2)^2 + (-0.8-0.3)^2 + \dots + (0.2-0.3)^2}{(0.2)^2 + (0.3)^2 + \dots + (0.2)^2} \\ &= 1.6268 \end{aligned}$$

$$d_{L, \alpha/2} = d_{L, 0.05} = 0.95, d_{U, \alpha/2} = d_{U, 0.05} = 1.54$$

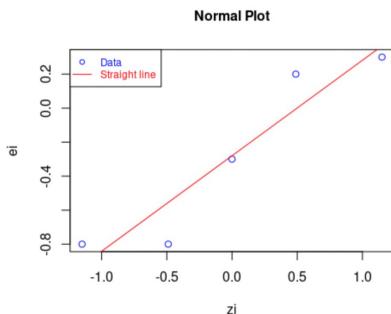
$$4 - d_{U, 0.05} = 4 - 1.54 = 2.46 > d_{U, 0.05}$$

Since $d = 1.6268 > d_{U, 0.05}$ and $4 - d_{U, \alpha/2} > d_{U, 0.05}$, we do not reject H_0 .

②

i	$l_{(i)}$	$(3i-1)/(3n+1)$	$Z_{(i)}$
1	-0.8	$[3(1)-1]/[3(5)+1] = 0.125$	$Z(1) = -1.15$
2	-0.8	$[3(2)-1]/[3(5)+1] = 0.3125$	-0.489
3	-0.3	$[3(3)-1]/[3(5)+1] = 0.5$	0
4	0.2	$[3(4)-1]/[3(5)+1] = 0.6875$	0.489
5	0.3	$[3(5)-1]/[3(5)+1] = 0.875$	1.15

$n=5$



$$P(Z \leq Z(1)) = 0.125$$

$$P(Z \leq Z(2)) = 0.3125$$

$$P(Z \leq Z(3)) = 0.5$$

$$P(Z \leq Z(4)) = 0.6875$$

$$P(Z \leq Z(5)) = 0.875$$

Q2 (20 points)

Q2(20 points). The sample data set is show as following

$x: 1, 0, 3, 2$

$y: 4, 3, 1, 3$

Apply the matrix form to calculate

1(9 points) The least squares estimates of β_0, β_1 .

2(4 points) SS_{Res}, SS_R, SS_T , multiple coefficient of determination R^2 .

3(2 points) Predict y_0 given $x_0 = [1, 2]$.

4(5 points) 90% prediction interval for y_0 given $x_0 = [1, 2]$.

(1)

$$\hat{\tilde{X}} = \begin{bmatrix} 1 & 1 \\ 1 & 0 \\ 1 & 3 \\ 1 & 2 \end{bmatrix}, \quad \tilde{X}' = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 0 & 3 & 2 \end{bmatrix}, \quad \hat{\tilde{y}} = \begin{bmatrix} 4 \\ 3 \\ 1 \\ 3 \end{bmatrix}$$

$$\hat{\beta} = (\tilde{X}' \tilde{X})^{-1} \tilde{X}' \hat{\tilde{y}}$$

$$(\tilde{X}' \tilde{X}) = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 0 & 3 & 2 \end{bmatrix}_{2 \times 4} \cdot \begin{bmatrix} 1 & 1 \\ 1 & 0 \\ 1 & 3 \\ 1 & 2 \end{bmatrix}_{4 \times 2}$$

$$= \begin{bmatrix} 1(1) + 1(1) + 1(1) + 1(1) & 1(1) + 1(0) + 1(3) + 1(2) \\ 1(1) + 0(1) + 3(1) + 2(1) & 1(1) + 0(0) + 3(3) + 2(2) \end{bmatrix}_{2 \times 2}$$

$$= \begin{bmatrix} 4 & 6 \\ 6 & 14 \end{bmatrix}$$

$$(\tilde{X}' \tilde{X})^{-1} = \begin{bmatrix} 4 & 6 \\ 6 & 14 \end{bmatrix}^{-1}$$

$$= \frac{1}{4(4) - 6(6)} \begin{bmatrix} 14 & -6 \\ -6 & 4 \end{bmatrix}$$

$$= \frac{1}{20} \begin{bmatrix} 14 & -6 \\ -6 & 4 \end{bmatrix}$$

$$= \begin{bmatrix} 14/20 & -6/20 \\ -6/20 & 4/20 \end{bmatrix}$$

$$= \begin{bmatrix} 7/10 & -3/10 \\ -3/10 & 1/5 \end{bmatrix}$$

$$= \begin{bmatrix} 0.7 & -0.3 \\ -0.3 & 0.2 \end{bmatrix}$$

$$\hat{\beta} = (\tilde{X}' \tilde{X})^{-1} \tilde{X}' \tilde{y}$$

$$= \begin{bmatrix} 7/10 & -3/10 \\ -3/10 & 1/5 \end{bmatrix} \cdot \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 0 & 3 & 2 \end{bmatrix} \cdot \begin{bmatrix} 4 \\ 3 \\ 1 \\ 3 \end{bmatrix}$$

$$= \begin{bmatrix} 3.8 \\ -0.7 \end{bmatrix}$$

$$\therefore \hat{\beta}_0 = 3.8, \hat{\beta}_1 = -0.7$$

$$\begin{aligned} \textcircled{2} \quad SS_{RES} &= \tilde{y}' \tilde{y} - \hat{\beta}' \tilde{X}' \tilde{y} \\ &= [4 \ 3 \ 1 \ 3] \cdot \begin{bmatrix} 4 \\ 3 \\ 1 \\ 3 \end{bmatrix} - [3.8 \ -0.7] \cdot \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 0 & 3 & 2 \end{bmatrix} \cdot \begin{bmatrix} 4 \\ 3 \\ 1 \\ 3 \end{bmatrix} \\ &= [4(4) + 3(3) + 1(1) + 3(3)] - [3.1 \ 3.8 \ 1.7 \ 2.4] \cdot \begin{bmatrix} 4 \\ 3 \\ 1 \\ 3 \end{bmatrix} \\ &= [35] - [3.1 \ 3.8 \ 1.7 \ 2.4] \cdot \begin{bmatrix} 4 \\ 3 \\ 1 \\ 3 \end{bmatrix} \\ &= 35 - 32.7 \\ &= 2.3 \end{aligned}$$

$$\begin{aligned} SS_R &= \hat{\beta} \tilde{X}' \tilde{y} - \bar{y}^2 \\ &= 32.7 - 4\left(\frac{4+3+1+3}{4}\right)^2 \\ &= 32.7 - 4(2.75)^2 \\ &= 2.45 \end{aligned}$$

$$\begin{aligned} SS_T &= \tilde{y}' \tilde{y} - \bar{y}^2 \\ &= 35 - 4(2.75)^2 \\ &= 4.75 \end{aligned}$$

$$R^2 = \frac{SS_R}{SS_T} = \frac{2.45}{4.75} = 0.5108$$

$$\textcircled{3} \quad \hat{y}_0 = \hat{x}_0 \beta = [1 \ 2] \cdot \begin{bmatrix} 3.8 \\ -0.7 \end{bmatrix} = 2.4$$

$$\textcircled{4} \quad \sigma = \sqrt{\frac{SS_{Res}}{n-k}} = \sqrt{\frac{2.3}{2}} = 1.0724$$

$$h_{00} = \hat{x}' (\hat{x}' \hat{x})^{-1} \hat{x} = [1 \ 2] \cdot \begin{bmatrix} 0.7 & -0.3 \\ -0.3 & 0.2 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 2 \end{bmatrix} = 0.3$$

$$t_{0.05}(2) = 2.92$$

$$\therefore \hat{y}_0 \pm 2.92 (1.0724) \sqrt{1+0.3} = (-1.1704, 5.9704)$$

Q3 (20 points)

Q3(20 points). Apply SAS to work on the following sample data.

i	x ₁	x ₂	y
1	1.43	2.79	1.23
2	7.90	5.59	6.12
3	-3.40	3.58	-1.90
4	52.87	9.73	44.61
5	54.39	9.32	41.28
6	4.70	0.13	10.56
7	39.68	8.91	34.78
8	21.75	7.03	16.57
9	12.62	6.46	10.09
10	-1.85	2.97	2.01

$\alpha = 0.01$ Find the following information from the SAS output.

1. Fit the data to linear regression model $y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_2^2 + \epsilon$.

- (3 points) Specify the least squares point estimators on your output.
- (3 points) Implement an F -test for a portion of the model: $H_0 : \beta_2 = \beta_3 = 0$, H_a : at least one of β_2, β_3 is not 0. Specify the test statistic $F(x_2, x_2^2 | x_1)$ value and P -value on your output, and make conclusion.
- (4 points) Code to display $\tilde{X}'\tilde{X}$, $(\tilde{X}'\tilde{X})^{-1}$, $\tilde{X}'\tilde{y}$, $\tilde{y}'\tilde{y}$, specify these matrices/vectors/numbers on your output.
- (3 points) Implement a Durbin-Watson test for first-order autocorrelation test (H_0 : no autocorrelation, H_a : positive autocorrelation). Specify the value of the test statistic and P -value on your output and give the conclusion.
- (2 points) Use the “Residual by Regressors for y ” panel of the regression model output to validate the constant variance assumption.
- (2 points) Calculate the prediction of y_0 for $x_{01} = 10, x_{02} = 5$.

2 (3 points) Fit the data to linear regression model $y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_1 * x_2 + \epsilon$, compare this with the above model (in sub-question 1.) using multiple coefficients of determination R^2 .

1. Fit the data to linear regression model $y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_2^2 + \epsilon$.

a. (3 points) Specify the least squares point estimators on your output.

Parameter Estimate
$\hat{\beta}_0 = 9.91079$
$\hat{\beta}_1 = 0.30646$
$\hat{\beta}_2 = -5.24023$
$\hat{\beta}_3 = 0.73677$

b. (3 points) Implement an F -test for a portion of the model: $H_0 : \beta_2 = \beta_3 = 0$,

H_a : at least one of β_2, β_3 is not 0. Specify the test statistic $F(x_2, x_2^2 | x_1)$ value and P -value on your output, and make conclusion.

The REG Procedure Model: MODEL1				
Test mytest Results for Dependent Variable y				
Source	DF	Mean Square	F Value	Pr > F
Numerator	2	25.63895	22.63	0.0016
Denominator	6	1.13318		

T.S. $F(x_2, x_2^2 | x_1)$

\downarrow \circlearrowleft \rightarrow p-value

Conclusion: Since $p\text{-value} = 0.0016 < \alpha = 0.01$, we reject H_0 , and conclude that

at least one of x_2 and x_2^2 is significant.

c. (4 points) Code to display $\tilde{X}'\tilde{X}$, $(\tilde{X}'\tilde{X})^{-1}$, $\tilde{X}'\tilde{y}$, $\tilde{y}'\tilde{y}$, specify these matrices/vectors/numbers on your output.

Model Crossproducts $X'X$ $X'Y$ $Y'Y$					
Variable	Intercept	x1	x2	x2sq	y
Intercept	10	190.09	56.51	412.7623	165.35
x1	190.09	8061.8657	1640.4116	14679.655592	6574.034
x2	56.51	1640.4116	412.7623	3323.565419	1348.5262
x2sq	412.7623	14679.655592	3323.565419	28273.561383	12004.502344
y	165.35	6574.034	1348.5262	12004.502344	5438.2429

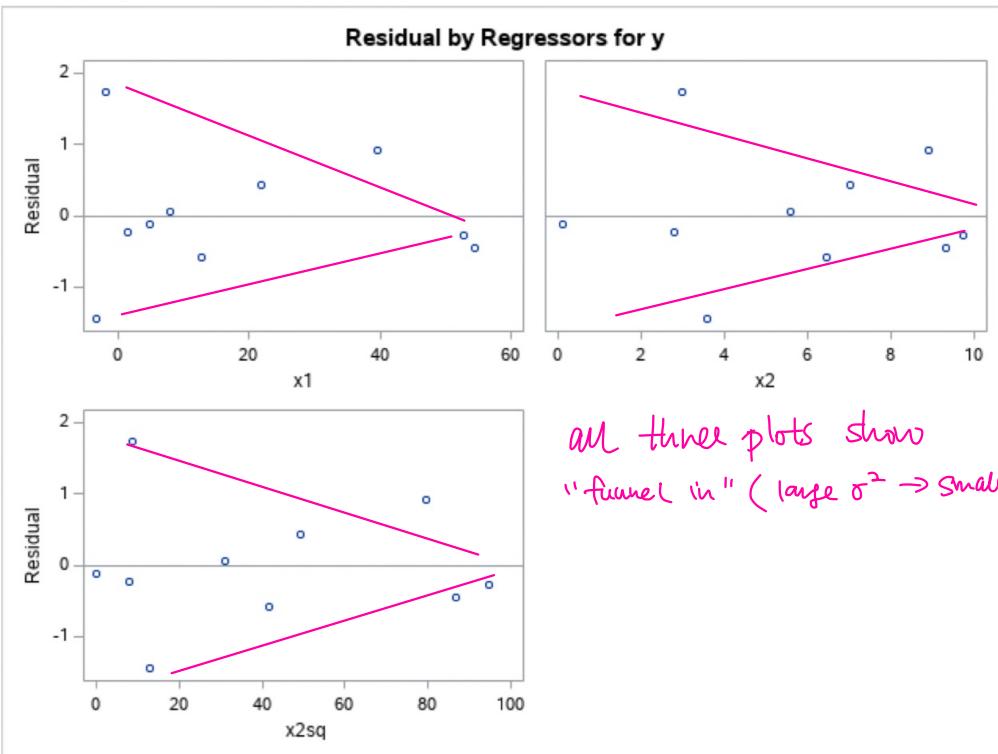
$$(\tilde{X}'\tilde{X})^{-1}$$

X'X Inverse, Parameter Estimates, and SSE					
Variable	Intercept	x1	x2	x2sq	y
Intercept	1.4489080779	-0.085966558	-0.856100865	0.1241163792	9.9107930382
x1	-0.085966558	0.0140719141	0.0854121885	-0.016091361	0.306459682
x2	-0.856100865	0.0854121885	0.684714617	-0.112336367	-5.24022871
x2sq	0.1241163792	-0.016091361	-0.112336367	0.0197832292	0.73677386
y	9.9107930382	0.306459682	-5.24022871	0.73677386	6.7990811384

d. (3 points) Implement a Durbin-Watson test for first-order autocorrelation test (H_0 : no autocorrelation, H_a : positive autocorrelation). Specify the value of the test statistic and P-value on your output and give the conclusion.

Durbin-Watson D		1.702	T.S. : d
P-value + A.C.	Pr < DW	0.2927	$p\text{-values} > \alpha = 0.01 \Rightarrow \text{do not reject } H_0$ (Conclusion)
P-value - A.C.	Pr > DW	0.7073	
Number of Observations		10	
1st Order Autocorrelation		-0.075	

- e. (2 points) Use the “Residual by Regressors for y ” panel of the regression model output to validate the constant variance assumption.



f. (2 points) Calculate the prediction of y_0 for $x_{01} = 10, x_{02} = 5$.

Output Statistics								
Obs	Dependent Variable	Predicted Value	Std Error Mean Predict	99% CL Mean		99% CL Predict		Residual
1	1.23	1.4639	0.6383	-0.9027	3.8305	-3.1379	6.0657	-0.2339
2	6.12	6.0617	0.5012	4.2034	7.9201	1.6995	10.4239	0.0583
3	-1.90	-0.4484	0.5709	-2.5649	1.6681	-4.9267	4.0299	-1.4516
4	44.61	44.8784	0.7555	42.0775	47.6793	40.0389	49.7179	-0.2684
5	41.28	41.7381	0.9124	38.3554	45.1209	36.5402	46.9361	-0.4581
6	10.56	10.6824	0.9792	7.0520	14.3128	5.3200	16.0448	-0.1224
7	34.78	33.8718	0.6465	31.4750	36.2685	29.2544	38.4891	0.9082
8	16.57	16.1495	0.5038	14.2818	18.0172	11.7833	20.5157	0.4205
9	10.09	10.6732	0.5371	8.6820	12.6644	6.2527	15.0937	-0.5832
10	2.01	0.2794	0.4751	-1.4822	2.0410	-4.0425	4.6013	1.7306
11	.	5.1936	0.8748	1.9502	8.4370	0.0852	10.3020	.

$$y_0 \text{ given } \tilde{X}_0' = [1 \ 10 \ 5 \ 25]$$

$$= 5.1936$$

2 (3 points) Fit the data to linear regression model $y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_1 * x_2 + \epsilon$, compare this with the above model (in sub-question 1.) using multiple coefficients of determination R^2 .

Root MSE	1.06451	R-Square	0.9975
Dependent Mean	16.53500	Adj R-Sq	0.9962
Coeff Var	6.43792		

→ R^2 (1st model)

Root MSE	2.38424	R-Square	0.9874
Dependent Mean	16.53500	Adj R-Sq	0.9811
Coeff Var	14.41932		

→ R^2 (2nd model)

Both R^2 are close to 1, so they both indicate a good fit model. The R^2 of the first model in sub-question 1 is greater than the R^2 of the second model ($0.9975 > 0.9874$), so the first model is better.

```
*****  
data A2_Q3; /* Assigns name A2_Q3 to file*/  
input i x1 x2 y; /* assigns variable names i x1 x2 y*/  
x2sq = x2*x2; /*interaction term data*/  
t = _N_;  
cards; /* start to input data */  
1 1.43 2.79 1.23  
2 7.90 5.59 6.12  
3 -3.40 3.58 -1.90  
4 52.87 9.73 44.61  
5 54.39 9.32 41.28  
6 4.70 0.13 10.56  
7 39.68 8.91 34.78  
8 21.75 7.03 16.57  
9 12.62 6.46 10.09  
10 -1.85 2.97 2.01  
  
; /* inputs data, .= missing value*/  
proc reg data = A2_Q3; /* specifies regression procedure */  
model y = x1 x2 x2sq /alpha=0.01 p clm cli clb xpx i dwprob;  
mytest: test x2=0, x2sq=0; /* partial F-test */  
run;
```

The REG Procedure
Model: MODEL1

Model Crossproducts X'X X'Y Y'Y					
Variable	Intercept	x1	x2	x2sq	y
Intercept	10	190.09	56.51	412.7623	165.35
x1	190.09	8061.8657	1640.4116	14679.655592	6574.034
x2	56.51	1640.4116	412.7623	3323.565419	1348.5262
x2sq	412.7623	14679.655592	3323.565419	28273.561383	12004.502344
y	165.35	6574.034	1348.5262	12004.502344	5438.2429

The REG Procedure
Model: MODEL1
Dependent Variable: y

Number of Observations Read	10
Number of Observations Used	10

X'X Inverse, Parameter Estimates, and SSE					
Variable	Intercept	x1	x2	x2sq	y
Intercept	1.4489080779	-0.085966558	-0.856100865	0.1241163792	9.9107930382
x1	-0.085966558	0.0140719141	0.0854121885	-0.016091361	0.306459682
x2	-0.856100865	0.0854121885	0.684714617	-0.112336367	-5.24022871
x2sq	0.1241163792	-0.016091361	-0.112336367	0.0197832292	0.73677386
y	9.9107930382	0.306459682	-5.24022871	0.73677386	6.7990811384

Analysis of Variance					
Source	DF	Sum of Squares	Mean Square	F Value	Pr > F
Model	3	2697.38157	899.12719	793.45	<.0001
Error	6	6.79908	1.13318		
Corrected Total	9	2704.18065			

Root MSE	1.06451	R-Square	0.9975
Dependent Mean	16.53500	Adj R-Sq	0.9962
Coeff Var	6.43792		

Parameter Estimates						
Variable	DF	Parameter Estimate	Standard Error	t Value	Pr > t 	99% Confidence Limits
Intercept	1	9.91079	1.28136	7.73	0.0002	5.16026 14.66133
x1	1	0.30646	0.12628	2.43	0.0514	-0.16171 0.77462
x2	1	-5.24023	0.88085	-5.95	0.0010	-8.50593 -1.97452
x2sq	1	0.73677	0.14973	4.92	0.0027	0.18167 1.29187

The REG Procedure
Model: MODEL1
Dependent Variable: y

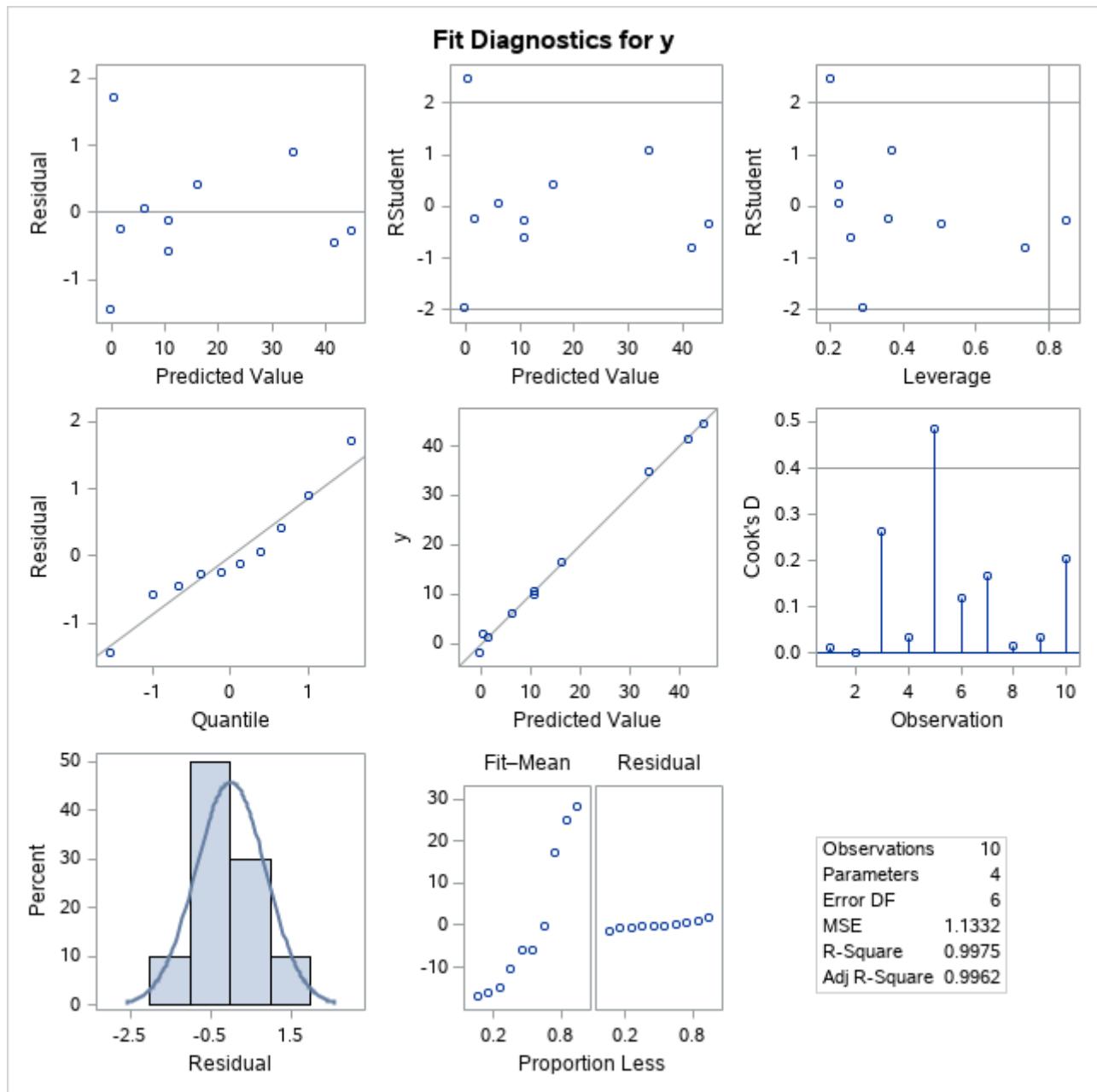
Durbin-Watson D	1.702
Pr < DW	0.2927
Pr > DW	0.7073
Number of Observations	10
1st Order Autocorrelation	-0.075

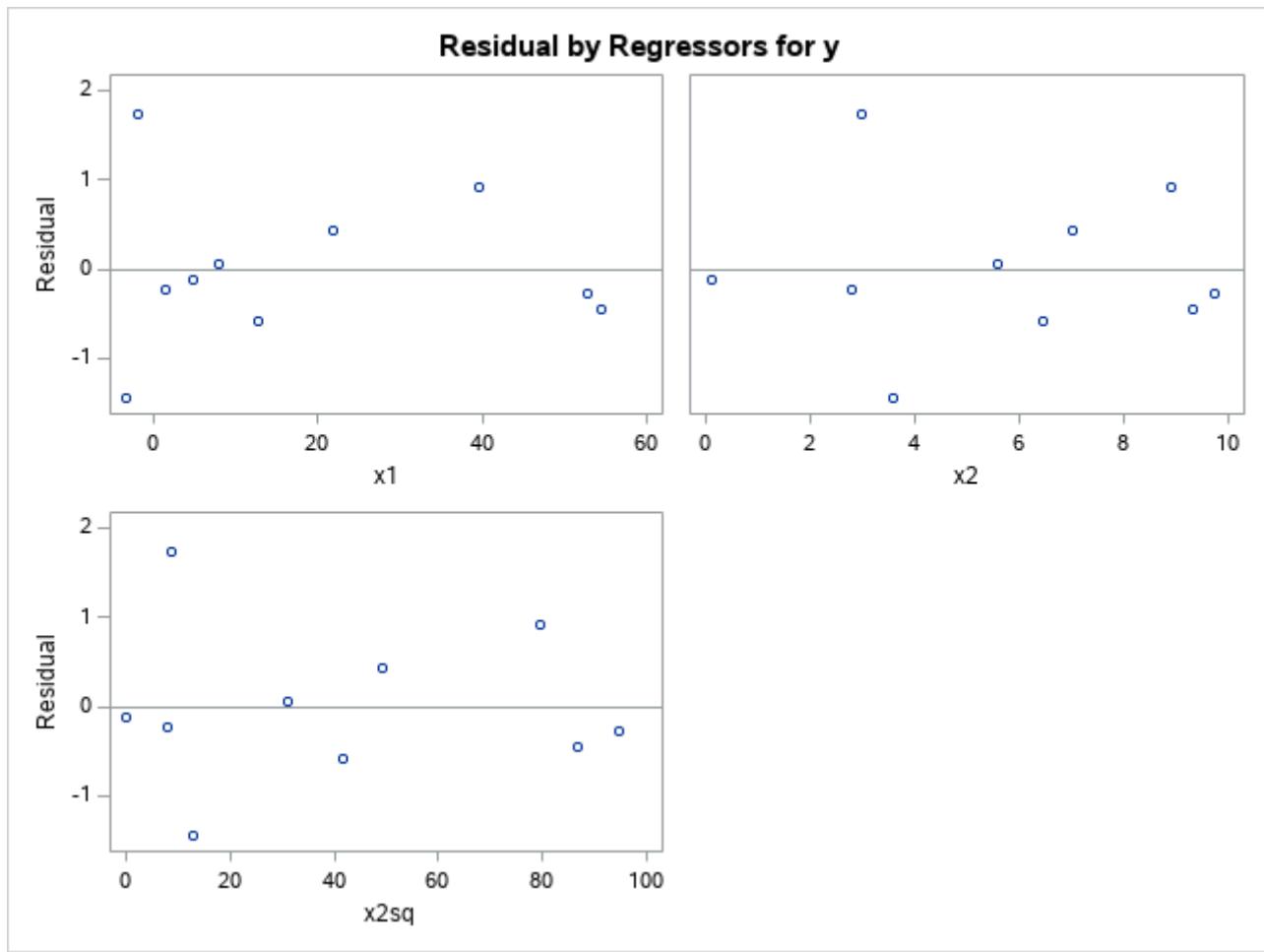
Note: Pr<DW is the p-value for testing positive autocorrelation, and Pr>DW is the p-value for testing negative autocorrelation.

The REG Procedure
Model: MODEL1
Dependent Variable: y

Obs	Dependent Variable	Predicted Value	Std Error Mean Predict	Output Statistics		99% CL Predict	Residual
				99% CL Mean			
1	1.23	1.4639	0.6383	-0.9027	3.8305	-3.1379	6.0657
2	6.12	6.0617	0.5012	4.2034	7.9201	1.6995	10.4239
3	-1.90	-0.4484	0.5709	-2.5649	1.6681	-4.9267	4.0299
4	44.61	44.8784	0.7555	42.0775	47.6793	40.0389	49.7179
5	41.28	41.7381	0.9124	38.3554	45.1209	36.5402	46.9361
6	10.56	10.6824	0.9792	7.0520	14.3128	5.3200	16.0448
7	34.78	33.8718	0.6465	31.4750	36.2685	29.2544	38.4891
8	16.57	16.1495	0.5038	14.2818	18.0172	11.7833	20.5157
9	10.09	10.6732	0.5371	8.6820	12.6644	6.2527	15.0937
10	2.01	0.2794	0.4751	-1.4822	2.0410	-4.0425	4.6013
							1.7306

Sum of Residuals	0
Sum of Squared Residuals	6.79908
Predicted Residual SS (PRESS)	15.84524





The REG Procedure
Model: MODEL1

Test mytest Results for Dependent Variable y				
Source	DF	Mean Square	F Value	Pr > F
Numerator	2	25.63895	22.63	0.0016
Denominator	6	1.13318		