

Fitted Value	Predictor Variable	Simple Linear Regression
Gauss-Markov Assumptions	Regressand	Model
Heteroskedasticity	Regression through the Origin	Slope Parameter
Homoskedasticity	Regressor	Standard Error of $\hat{\beta}_1$
Independent Variable	Residual	Standard Error of the
Intercept Parameter	Residual Sum of Squares	Regression (SER)
Mean Independent	(SSR)	Sum of Squared Residuals
OLS Regression Line	Response Variable	(SSR)
Ordinary Least Squares (OLS)	R-squared	Total Sum of Squares (SST)
Population Regression	Sample Regression Function	Zero Conditional Mean
Function (PRF)	(SRF)	Assumption
Predicted Variable	Semi-elasticity	

PROBLEMS

2.1 In the simple linear regression model $y = \beta_0 + \beta_1 x + u$, suppose that $E(u) \neq 0$. Letting $\alpha_0 = E(u)$, show that the model can always be rewritten with the same slope, but a new intercept and error, where the new error has a zero expected value.

2.2 The following table contains the *ACT* scores and the *GPA* (grade point average) for eight college students. Grade point average is based on a four-point scale and has been rounded to one digit after the decimal.

In class

HW

Student	GPA	ACT
1	2.8	21
2	3.4	24
3	3.0	26
4	3.5	27
5	3.6	29
6	3.0	25
7	2.7	25
8	3.7	30

- (i) Estimate the relationship between *GPA* and *ACT* using OLS; that is, obtain the intercept and slope estimates in the equation

$$\widehat{GPA} = \hat{\beta}_0 + \hat{\beta}_1 ACT.$$

Comment on the direction of the relationship. Does the intercept have a useful interpretation here? Explain. How much higher is the *GPA* predicted to be if the *ACT* score is increased by five points?

- (ii) Compute the fitted values and residuals for each observation, and verify that the residuals (approximately) sum to zero.