

II. Some Concepts

11. The next 16 lines is the shazam code for the HPRICE1 data from class (with variable descriptions given in the program as comment statements; CAREFULLY LOOK AT THE GENR STATEMENTS):

```
read(hprice1.raw) price assess bdrms lotsize sqrft colonial lprice lassess &
llotsize lsqrft
* 1. price          price, in dollars
* 3. bdrms         number of bedrooms
* 5. sqrft         size of house, square feet
stat price bdrms sqrft
*next generate deviations from the means
genr bdrm_dev = bdrms - 3.5682
genr sqrft_dev = sqrft - 2013.7
genr inter_dv = bdrm_dev*sqrft_dev
ols price bdrms sqrft inter_dv
test
  test bdrms=0
  test inter_dv=0
end
stop
```

THIS IS THE RELEVANT OUTPUT FROM THE PROGRAM ABOVE:

NAME	N	MEAN	ST. DEV	VARIANCE	MINIMUM	MAXIMUM
PRICE	88	0.29355E+06	0.10271E+06	0.10550E+11	0.11100E+06	0.72500E+06
BDRMS	88	3.5682	0.84139	0.70794	2.0000	7.0000
SQRFT	88	2013.7	577.19	0.33315E+06	1171.0	3880.0

VARIABLE	ESTIMATED	STANDARD	T-RATIO	PARTIAL STANDARDIZED ELASTICITY			
NAME	COEFFICIENT	ERROR	84 DF	P-VALUE	CORR.	COEFFICIENT	AT MEANS
BDRMS	11262.	9407.	1.197	0.235	0.130	0.0923	0.1369
SQRFT	116.29	14.48	8.033	0.000	0.659	0.6535	0.7977
INTER_DV	23.448	10.16	2.308	0.023	0.244	0.1732	0.0204
CONSTANT	13210.	0.3340E+05	0.3955	0.693	0.043	0.0000	0.0450

_test

_test bdrms=0

_test inter_dv=0

_end

F STATISTIC = 4.0141144 WITH 2 AND 84 D.F. P-VALUE= 0.02163

WALD CHI-SQUARE STATISTIC = 8.0282288 WITH 2 D.F. P-VALUE= 0.01806

UPPER BOUND ON P-VALUE BY CHEBYCHEV INEQUALITY = 0.24912

a) Interpret the BDRMS and SQRFT coefficients?

b) What does the INTER_DV estimate?

c) What does the F STATISTIC indicate (be specific about the null hypothesis being tested)?

12. Indicate whether the following statement is True, False or Uncertain and explain why. You are graded only on the explanation for your answer.

a. "If all my sample observations have exactly the same level of ability, then omitting ability from the empirical specification where I regress wages on education will lead to no omitted variable bias."

b. "The most important model assumption is that the population errors are normally distributed. Unless this assumption holds, we can do no testing in our regression models."

13. The estimators (i.e., the formulas from which we calculate the estimates) for the OLS slope and intercept coefficients in the simple regression model ($y_i = \beta_0 + \beta_1 x_i + \mu_i$) are derived from the following “normal equations”:

$$-2 \sum_{i=1}^N (y_i - \hat{\beta}_0 - \hat{\beta}_1 x_i) = 0$$

$$-2 \sum_{i=1}^N (y_i - \hat{\beta}_0 - \hat{\beta}_1 x_i) x_i = 0$$

(note that we can easily divide out the -2 s). But how do we get the normal equations? You show that we can get these normal equations either by

- a) Calculus (minimizing the sum of squared residuals), or
- b) from the orthogonality condition on the residuals.

Show it both ways. Be explicit in your explanations (on each step): again, you don't have to solve the normal equations for the OLS estimators, you have to derive the normal equations in two different ways.

14. For the simple regression, $y_i = \beta_0 + \beta_1 x_i + \mu_i$, we can regress the dependent variable on the independent variable and get $\hat{\beta}_1$, as we did in problem 13. But we can also divide through the dependent variable by its standard deviation (call it σ_y) to get “ $y_i / \sigma_y = y_i^*$,” and similarly divide through the independent variables by its standard deviation to get “ $x_i / \sigma_x = x_i^*$.” Then we have a model of normalized regressors: $y_i^* = \alpha_0 + \alpha_1 x_i^* + \varepsilon_i$. Prove that following relationship exists between the slope coefficient of the simple model and the slope coefficient of the normalized model: $\hat{\alpha}_1 = \hat{\beta}_1 \frac{\sigma_x}{\sigma_y}$.

15. Prove that under the usual model assumptions that the least squares estimator of the multiple regression model, $\hat{\beta}$, is (a) an unbiased estimator, and (b) a consistent estimator of the population coefficient vector β .