Sociology 63993
Exam 1 Answer Key
February 18, 2011

I. True-False. (20 points) Indicate whether the following statements are true or false. If false, briefly explain why.

1. A data set contains a few extreme outliers. It is usually best to use Stata’s rreg (Robust Regression) routine to deal with the problem.

False. Indeed, this may be one of the worst options. Check the coding first, consider adding new vars to the model, try running the analysis with and without the outlier, or try some other robust regression technique (e.g. qreg).

2. The independent variables in an analysis include X1, X2, and X1X2 (i.e. X1 * X2). X1 has missing data (and hence X1X2 does too). If multiple imputation is being used, you should first compute X1X2, and then impute the missing values for X1 and X1X2.

True. Passive imputation, where you impute X1 first and then compute X1X2, may seem more intuitive to some. But, as Allison and others note, it can bias correlations toward zero. [Note: I think I was more definitive about this in class than I was in the notes, so I will show a little leeway when grading if you show you understand the issues and concepts.]

3. Cronbach’s Alpha is used to test for serial correlation.

False. Cronbach’s Alpha assesses the reliability of a scale. The Durbin-Watson statistic can be used for serial correlation.

4. The less true variability there is in a population, the higher the reliability of measures will tend to be.

False. Reliability = True Variance/ Total Variance, so the higher the true variability, the higher the reliability tends to be.

5. The most extreme outliers on Y (i.e. the cases where Y is furthest from the mean) will always have the most influence on the regression line.

False. Influence = discrepancy * leverage. A highly discrepant case can still have little or no influence on the regression line if its X values are at or near the means of X.

II. Short answer. Discuss all three of the following problems. (15 points each. 45 points total.) In each case, the researcher has used Stata to test for a possible problem, concluded that there is a problem, and then adopted a strategy to address that problem. Explain (a) what problem the researcher was testing for, and why she concluded that there was a problem, (b) the rationale behind the solution she chose, i.e. how does it try to address the problem, and (c) one alternative solution she could have tried, and why. (NOTE: a few sentences on each point will probably suffice – you don’t have to repeat everything that was in the lecture notes.)
II-1.

. `sum` income white male age fathered

<table>
<thead>
<tr>
<th>Variable</th>
<th>Obs</th>
<th>Mean</th>
<th>Std. Dev.</th>
<th>Min</th>
<th>Max</th>
</tr>
</thead>
<tbody>
<tr>
<td>income</td>
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<td>16.9698</td>
<td>8.464258</td>
<td>.5</td>
<td>25</td>
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<tr>
<td>white</td>
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<td>.3424337</td>
<td>0</td>
<td>1</td>
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<tr>
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<td>.5001245</td>
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<td>1</td>
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<tr>
<td>age</td>
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<td>38.53695</td>
<td>11.92651</td>
<td>18</td>
<td>81</td>
</tr>
<tr>
<td>fathered</td>
<td>695</td>
<td>11.44173</td>
<td>3.838113</td>
<td>0</td>
<td>20</td>
</tr>
</tbody>
</table>

. `fre` fathered

fathered -- HIGHEST YEAR SCHOOL COMPLETED, FATHER

<table>
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<tr>
<th></th>
<th>Freq.</th>
<th>Percent</th>
<th>Valid</th>
<th>Cum.</th>
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<td>0.49</td>
<td>0.58</td>
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<td></td>
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<td>5</td>
<td>10</td>
<td>1.23</td>
<td>1.44</td>
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<td></td>
<td>6</td>
<td>38</td>
<td>4.68</td>
<td>5.47</td>
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<td></td>
<td>7</td>
<td>17</td>
<td>2.09</td>
<td>2.45</td>
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<td></td>
<td>8</td>
<td>84</td>
<td>10.34</td>
<td>12.09</td>
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<td></td>
<td>9</td>
<td>28</td>
<td>3.45</td>
<td>4.03</td>
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<td></td>
<td>10</td>
<td>30</td>
<td>3.69</td>
<td>4.32</td>
</tr>
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<td></td>
<td>11</td>
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<td>2.59</td>
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<td>27.59</td>
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<td>2.46</td>
<td>2.88</td>
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<td>14</td>
<td>64</td>
<td>7.88</td>
<td>9.21</td>
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<td>15</td>
<td>9</td>
<td>1.11</td>
<td>1.29</td>
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<td>16</td>
<td>71</td>
<td>8.74</td>
<td>10.22</td>
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<td>17</td>
<td>7</td>
<td>0.86</td>
<td>1.01</td>
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<td></td>
<td>18</td>
<td>15</td>
<td>1.85</td>
<td>2.16</td>
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<td>19</td>
<td>10</td>
<td>1.23</td>
<td>1.44</td>
</tr>
<tr>
<td></td>
<td>20</td>
<td>16</td>
<td>1.97</td>
<td>2.30</td>
</tr>
<tr>
<td>Total</td>
<td>695</td>
<td>85.59</td>
<td>100.00</td>
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</tr>
<tr>
<td>Missing</td>
<td>117</td>
<td>14.41</td>
<td></td>
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</tr>
<tr>
<td>Total</td>
<td>812</td>
<td>100.00</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

. `gen` one = 1
. `gen` mdfathered = missing(fathered)
. `impute` fathered one, gen(fathered2)

14.41% (117) observations imputed
. `fre` fathered2 mdfathered
The researcher observed that fathered had a lot of missing data. Further, the reason it was missing was because some respondents came from families where there was no father, i.e. it was missing because the value didn’t exist, not because the respondent failed to report it. [Note: In order to make the rationale clear, it is important to point out why the data was missing; if it were missing for other reasons this would be a bad approach.] The researcher therefore decided to use Cohen and Cohen’s dummy variable adjustment method, where you substitute the mean for the missing and then include a dummy variable that indicates that the data was missing. This is often a bad
method, but it is fine when the missing values simply don’t exist. Listwise deletion might have been the next best option.

II-2.

. reg warm ed age prst

<table>
<thead>
<tr>
<th>Source</th>
<th>SS</th>
<th>df</th>
<th>MS</th>
<th>Number of obs = 4586</th>
</tr>
</thead>
<tbody>
<tr>
<td>Model</td>
<td>249.541491</td>
<td>3</td>
<td>83.1804971</td>
<td>F( 3, 4582) = 103.01</td>
</tr>
<tr>
<td>Residual</td>
<td>3699.96047</td>
<td>4582</td>
<td>.807499012</td>
<td>Prob &gt; F = 0.0000</td>
</tr>
<tr>
<td>Total</td>
<td>3949.50196</td>
<td>4585</td>
<td>.861396284</td>
<td>R-squared = 0.0632</td>
</tr>
</tbody>
</table>

| Coef.    | Std. Err. | t     | P>|t| | [95% Conf. Interval] |
|----------|-----------|-------|-------|----------------------|
| ed       | .0374512  | .0054324 | 6.89 | 0.000  | .0268012 - .0481013 |
| age      | -.0094214 | .0008435 | -11.17 | 0.000  | -.0110751 - .0077677 |
| prst     | .0018836  | .0011332 | 1.66  | 0.097  | -.000338 - .0041052 |
| _cons    | 2.498711  | .0748558 | 33.38 | 0.000  | 2.351958 - 2.645465 |

. estat hettest

Breusch-Pagan / Cook-Weisberg test for heteroskedasticity
Ho: Constant variance Variables: fitted values of warm
chi2(1) = 7.00
Prob > chi2 = 0.0081

. reg warm ed age prst male

<table>
<thead>
<tr>
<th>Source</th>
<th>SS</th>
<th>df</th>
<th>MS</th>
<th>Number of obs = 4586</th>
</tr>
</thead>
<tbody>
<tr>
<td>Model</td>
<td>389.311386</td>
<td>4</td>
<td>97.3278466</td>
<td>F( 4, 4581) = 125.23</td>
</tr>
<tr>
<td>Residual</td>
<td>3560.19058</td>
<td>4581</td>
<td>.7771645</td>
<td>Prob &gt; F = 0.0000</td>
</tr>
<tr>
<td>Total</td>
<td>3949.50196</td>
<td>4585</td>
<td>.861396284</td>
<td>R-squared = 0.0986</td>
</tr>
</tbody>
</table>

| Coef.    | Std. Err. | t     | P>|t| | [95% Conf. Interval] |
|----------|-----------|-------|-------|----------------------|
| ed       | .0368867  | .0053295 | 6.92  | 0.000  | .0264383 - .0473351 |
| age      | -.0099226 | .0008284 | -11.98 | 0.000  | -.1115466 - .0882986 |
| prst     | .0025542  | .0011128 | 2.30  | 0.022  | .0003726 - .0047359 |
| male     | -.3508326 | .0261607 | -13.41 | 0.000  | -0.4021202 - .299545 |
| _cons    | 2.664683  | .0744719 | 35.78 | 0.000  | 2.518682 - 2.810683 |

. estat hettest

Breusch-Pagan / Cook-Weisberg test for heteroskedasticity
Ho: Constant variance Variables: fitted values of warm
chi2(1) = 0.03
Prob > chi2 = 0.8613

The researcher tested for heteroskedasticity and found that it was present. Apparently, however, she thought this might be an artifact of an improperly specified model, so she added the variable male to the analysis. This appears to have been a good choice; the effect of male is highly significant and heteroskedasticity (at least linear heteroskedasticity) is no longer a problem. She could have also used robust standard errors or weighted least squares, but it is best to make sure the model is correctly specified first.
II-3.

```
. reg price w1 w2 w3

Source |       SS       df       MS              Number of obs =      74
--------+--------------------------------------------------------------
Model |   196801072     3  65600357.4           Prob > F      =  0.0000
Residual |   438264324    70  6260918.91           R-squared    =  0.3099
--------+--------------------------------------------------------------
Total |   635065396    73  8699525.97           Root MSE      =  2502.2
        -------------------------------------------
price |      Coef.   Std. Err.      t    P>|t|     [95% Conf. Interval]
--------+--------------------------------------------------------------
    w1 |   1.998095   1.423422     1.40   0.165   -.8408306    4.83702
    w2 |   .9836392   .9768691     1.01   0.317   -.9646648    2.931943
    w3 |  -.9777821   .9785287    -1.00   0.321   -.2929396    .9738319
_cons |   114.4055   1177.767     0.10   0.923  -2234.576    2463.387

. corr price w1 w2 w3
(obs=74)

|     price       w1       w2       w3
<table>
<thead>
<tr>
<th></th>
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<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>price</td>
<td>1.0000</td>
<td>1.0000</td>
<td>1.0000</td>
</tr>
<tr>
<td>w1</td>
<td>0.5386</td>
<td>0.5389</td>
<td>0.9347</td>
</tr>
<tr>
<td>w2</td>
<td>0.5389</td>
<td>0.9347</td>
<td>0.8695</td>
</tr>
<tr>
<td>w3</td>
<td>0.4644</td>
<td>0.9299</td>
<td>0.8695</td>
</tr>
</tbody>
</table>

. sw, pe(.05): reg price w1 w2 w3
begin with empty model
p = 0.0000 < 0.0500 adding w2

Source |       SS       df   MS              Number of obs =      74
--------+--------------------------------------------------------------
Model |   184420235     1  184420235           Prob > F      =  0.0000
Residual |   450645161    72  6258960.58           R-squared    =  0.2904
--------+--------------------------------------------------------------
Total |   635065396    73  8699525.97           Root MSE      =  2501.8
        -------------------------------------------
price |      Coef.   Std. Err.      t    P>|t|     [95% Conf. Interval]
--------+--------------------------------------------------------------
    w2 |   1.884568   .3471831     5.43   0.000     1.192471    2.576664
  _cons |   474.8814   1087.899     0.44   0.664  -1693.806    2643.569

Multicollinearity seems to be a problem. The global F is significant but none of the individual T values are. The correlation matrix reveals that the three independent variables are highly correlated with each other. The researcher therefore decided to use forward stepwise selection to decide what variables to include, and only w2 met the selection criteria. This may be a bad choice of strategies though. Note that w1 and w2 have virtually identical correlations with price; a slightly different sample could lead to other variables being selected. The researcher could have just used theory to choose between the variables, or she could have tried creating a scale out of them.

III. Computation and interpretation. (35 points total) The Indiana State legislature is considering a measure that would make gay marriage unconstitutional. The Indianapolis Chamber of Commerce opposes the measure because it worries that the resolution will cast the state as intolerant and put off talented workers who might otherwise relocate to Indianapolis. The Chamber has therefore commissioned a study of 10,000 Hoosiers to see where residents of the state stand on the issue. The variables are
An analysis of the data yields the following results. [NOTE: You’ll need some parts of the following to answer the questions, but other parts are extraneous. You’ll have to figure out which is which.]

\[ \text{sum} \]

<table>
<thead>
<tr>
<th>Variable</th>
<th>Obs</th>
<th>Mean</th>
<th>Std. Dev.</th>
<th>Min</th>
<th>Max</th>
</tr>
</thead>
<tbody>
<tr>
<td>age</td>
<td>10337</td>
<td>47.5637</td>
<td>17.21678</td>
<td>20</td>
<td>74</td>
</tr>
<tr>
<td>black</td>
<td>10337</td>
<td>.10506</td>
<td>.3066449</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>evangel</td>
<td>10337</td>
<td>.2907033</td>
<td>.4541088</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>educ</td>
<td>10337</td>
<td>14.26352</td>
<td>5.043619</td>
<td>5</td>
<td>20</td>
</tr>
<tr>
<td>gaymarriage</td>
<td>10337</td>
<td>23.12387</td>
<td>50.68773</td>
<td>-188.7194</td>
<td>186.1061</td>
</tr>
</tbody>
</table>

\[ \text{reg gaymarriage evangel black educ age, beta} \]

| Source          | SS    | df  | MS          | Number of obs = 10337
|-----------------|-------|-----|-------------|---------------------------|
| Model           | 14993619.8 | 4   | 3748404.95  | 0.0000
| Residual        | 11562101.6 | 10332 | 1119.05746  | 0.0000
| Total           | 26555721.4 | 10336 | 25754885.94 | 0.0000

| gaymarriage | Coef.        | Std. Err. | t     | P>|t| | Beta |
|-------------|--------------|-----------|-------|------|------|
| evangel    | -42.53951    | .7288237  | [3]   | 0.000| -.3811094 |
| black       | -34.44778    | 1.078767  | -31.93| 0.000| -.2083983 |
| educ        | 6.174029     | .0652522  | 94.62 | 0.000| .6143391 |
| age         | -.2635312    | .0191403  | -13.77| 0.000| -.089512 |
| _cons       | 1.38087      |           |       | 0.000| .      |

\[ \text{estat hettest} \]

Breusch-Pagan / Cook-Weisberg test for heteroskedasticity
Ho: Constant variance
Variables: fitted values of gaymarriage
chi2(1) = 49.70
Prob > chi2 = 0.0000

\[ \text{pcorr gaymarriage evangel black educ age (obs=10337)} \]

Partial and semipartial correlations of gaymarriage with

<table>
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<tr>
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</thead>
<tbody>
<tr>
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<td>-0.3789</td>
<td>0.2480</td>
<td>0.1436</td>
<td>0.0000</td>
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<tr>
<td>black</td>
<td>-0.2997</td>
<td>-0.2073</td>
<td>0.0898</td>
<td>0.0430</td>
<td>0.0000</td>
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<tr>
<td>educ</td>
<td>0.6813</td>
<td>0.6142</td>
<td>0.4642</td>
<td>0.3773</td>
<td>0.0000</td>
</tr>
<tr>
<td>age</td>
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<td>-0.0894</td>
<td>0.0180</td>
<td>0.0080</td>
<td>0.0000</td>
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</table>
. predict rstandard, rstandard
. sum rstandard

<table>
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<tr>
<th>Variable</th>
<th>Obs</th>
<th>Mean</th>
<th>Std. Dev.</th>
<th>Min</th>
<th>Max</th>
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<tr>
<td>rstandard</td>
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<td>-8.04e-07</td>
<td>1.000047</td>
<td>-3.671386</td>
<td>3.441897</td>
</tr>
</tbody>
</table>

. test evangel black educ age

( 1)  evangel = 0
( 2)  black = 0
( 3)  educ = 0
( 4)  age = 0

F(  4, 10332) = [5]
Prob > F = 0.0000

. test evangel = black

( 1)  evangel - black = 0

F(  1, 10332) = 42.49
Prob > F = 0.0000

. reg gaymarriage evangel black educ age, beta robust

Linear regression
Number of obs = 10337
F( 4, 10332) = 3387.31
Prob > F = 0.0000
R-squared = 0.5646
Root MSE = 33.452

------------------------------------------------------------------
gaymarriage | Coef.  Std. Err.      t    P>|t|     Beta
-------------|--------|-------------|------|---------|--------|
evangel    | -42.53951  .723011  -58.84  0.000   -.3811094 |
black      | -34.44778  1.087479  -31.68  0.000   -.2083983 |
educ       |  6.174029  .0642269   96.13  0.000   .6143391 |
age        | -0.2635312  .0191713  -13.75  0.000   -.089512 |
_cons      | -36.41955  1.385137  -26.29  0.000        |
-------------|--------|-------------|------|---------|--------|

a) (10 pts) Fill in the missing quantities [1] – [5]. (A few other values have also been blanked out, but you don’t need to fill them in.)

Here are the key uncensored parts of the output:

. reg gaymarriage evangel black educ age, beta

Source | SS     | df | MS      | Number of obs = 10337
--------|--------|----|---------|------------------------
Model   | 14993619.8  4  3748404.95 | F( 4, 10332) = 3387.31 |
Residual| 11562101.6 10332  1119.05746 | Prob > F = 0.0000 |
Total   | 26555721.4 10336  2569.2455 | R-squared = 0.5646 |
                    | Root MSE = 33.452   |
                    |                     |
| gaymarriage | Coef.  Std. Err.      t    P>|t|  Beta |
-------------|--------|-------------|------|---------|--------|
evangel    | -42.53951  .7288237  -58.37  0.000   -.3811094 |
black      | -34.44778  1.087479  -31.93  0.000   -.2083983 |
educ       |  6.174029  .0642269   96.13  0.000   .6143391 |
age        | -0.2635312  .0191713  -13.75  0.000   -.089512 |
_cons      | -36.41955  1.38087    -26.29  0.000 |
-------------|--------|-------------|------|---------|--------|
. test evangel black educ age

( 1)  evangel = 0
( 2)  black = 0
( 3)  educ = 0
( 4)  age = 0

F(  4, 10332) = 3349.61
Prob > F =    0.0000

[1] = R^2 = SSR/SST = 14993619.8/26555721.4 = 0.5646
[2] = MST = V(Y) = SD(Y)^2 = 50.68773^2 = 2569.25.

Or, do SST/DFT = 26555721.4/ 10336 = 2569.25
[3] = T_{evangel} = B_{evangel}/SE_{evangel} = -42.53951/.7288237 = -58.37

Or, do SE_{Constant} * T_{Constant} = 1.38087 * -26.37 = -36.41
[5] = Global F = 3349.61 (i.e. this is the same F test as the regression command already did. You don’t need to calculate anything.)

b) (25 points) Answer the following questions about the analysis and the results, explaining how the printout supports your conclusions.

1. Summarize the key findings. What groups or types of individuals are most supportive of gay marriage and which are least supportive?

Evangelicals, blacks and older individuals all have lower levels of support for gay marriage. The better educated someone is, the higher their support tends to be.

2. There was a problem with the study that almost caused the variable age not to be measured. How would R^2 have declined if age was not included in the model?

As the squared semipartial shows, the R^2 would have gone down by .0080. To confirm,

. reg gaymarriage  black evangel educ

Source |       SS       df       MS
---------+----------------------
Model |  14781481.7     3  4927160.57           Prob > F      =  0.0000
Residual |  11774239.7 10333  1139.47931           R-squared     =  0.5566
---------+----------------------
Total |  26555721.4 10336   2569.2455           Root MSE      =  33.756

-------------------------------

gaymarriage |      Coef.   Std. Err.      t    P>|t|     [95% Conf. Interval]
-----------|----------------------|----------------------|----------------------|----------------------|
    black  |  -33.90988     1.087852   -31.17   0.000    -36.04228    -31.77748
   evangel |  -42.09633     .7347262    -57.30 0.000    -43.53653    -40.65612
    educ  |   6.178306     .0658442    93.83   0.000     6.049239    6.307374
  _cons  |  -49.20043     1.03159      -47.69 0.000    -51.22254    -47.17831

-------------------------------

3. Why did the researchers run the regression a second time? What, if anything, was different about the two regressions? Do the differences have any major effects on the conclusions?

The Breusch-Pagan test revealed that heteroskedasticity is a problem with the data. She therefore used robust standard errors, which relax the assumptions about iid errors, to address the problem. In practice, however it had virtually no effect. The coefficient estimates remained the same (as they should) and the standard errors and T
values changed only slightly. The analyses also suggested outliers may be an issue but robust standard errors do not address that.

4. Before she began the study, the researcher expected education to be the least important determinant of support for gay marriage. Indicate whether you think the results support or do not support her belief.

All the evidence seems to suggest just the opposite. Education has the largest T value, the largest standardized beta, and the largest squared semipartial correlation. [Note: There are multiple ways of assessing how important a variable is and a good answer should include more than just one of them.]

5. The statistician preparing the report is very annoyed with her assistant who did the computer runs. She specifically told him that she wanted an incremental F test of the hypothesis that neither evangel nor black affected support for gay marriage, NOT just separate t tests of each coefficient; but she says the output does not contain the information she needs. Explain why you either agree or disagree with her; if you disagree, give her the information she wants.

She is right to be annoyed; the incremental F statistic is not in the output. The assistant did include the command `test evangel = black`, but that tests whether the two effects are equal to each other, not whether either or both equals zero. The command `test evangel black` would have given the statistician what she wanted, e.g.

```
. quietly reg gaymarriage evangel black educ age
. test black evangel
```

```
( 1)  black = 0
( 2)  evangel = 0
F(  2, 10332) = 2050.47
    Prob > F =    0.0000.
```

She could have also run multiple models and computed the incremental F statistic. For example,

```
. nestreg, quietly: reg gaymarriage (educ age) (evangel black)
```

```
| Block |       F     df        df   Pr > F       R2    in R2 |
|-------|-------------------|-------------------|-------------------|-------------------|
| 1     | 3328.51          | 2                 | 10334             | 0.0000            | 0.3918            |
| 2     | 2050.47          | 2                 | 10332             | 0.0000            | 0.5646            |
```

As you would have expected from the T values, the effects of either or both variables significantly differ from 0.
Appendix: Stata Code

use "D:\SOC63993\Homework\missing.dta", clear
version 11.1
* II-1
* Set up data
recode race (1=1)(else=0), gen(white)
recode sex (1=1)(else=0), gen(male)
recode rincome (1=.5) (2=2) (3=3) (4=4.5) (5=5.5) (6=6.5) (7=7.5) (8=9) ///
(9=12.5) (10=17.5) (11=22.5) (12=25) (else=.), gen(income)
drop if missing(income)
cleonevar fathered = paeduc
drop if fathered > .a
label define fathered .a "R is from Fatherless Family"
label values fathered fathered
* Output for problem
sum income white male age fathered
fre fathered
gen one = 1
gen mdfathered = missing(fathered)
impute fathered one, gen(fathered2)
fre fathered2 mdfathered
reg income white male age fathered2 mdfathered
* II-2
* Set up data
use "http://www.indiana.edu/~jslsoc/stata/spex_data/ordwarm2.dta", clear
expand 2
* Output for problem
reg warm ed age prst
estat hettest
reg warm ed age prst male
estat hettest
* II-3
* Set up data
sysuse auto, cl
clonevar w1 = weight
corr2data e2 e3, sd(300 300)
gen w2 = w1 + e2
gen w3 = w1 + e3
* Output for problem
reg price w1 w2 w3
corr price w1 w2 w3
sw, pe(.05): reg price w1 w2 w3
* III
* Set up data
webuse nhanes2f, clear
corr2data e, sd(10)
gen evangel = smsa2
recode agegrp(6 = 1)(3=2)(5=6) (1=5) (2=3)(4=4)
gen educ = 3 * agegrp + 2
gen gaymarriage = (-39 - 48* evangel - 39 * black + 6.8 * educ -.3 * age + 3*e + e*educ/20) * .9
keep if !missing(gaymarriage)
keep gaymarriage evangel black educ age
* Output for problem
sum
reg gaymarriage evangel black educ age, beta
estat hettest
pccorr gaymarriage evangel black educ age
predict rstandard, rstandard
sum rstandard
test evangel black educ age
test evangel = black
collin evangel black educ age if e(sample)
reg gaymarriage evangel black educ age, beta robust
* Confirm the decline in R^2 from dropping age
reg gaymarriage black evangel educ
* Do joint tests of the significance of evangel and black quietly reg gaymarriage evangel black educ age
test black evangel
nestreg, quietly: reg gaymarriage (educ age) (evangel black)