Models for Group Comparisons – Summary

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Since we are estimating and comparing several models, it will be helpful to list several of them all in one place. This handout summarizes how to do group comparisons by running separate models for each group and by using interaction terms. We won't go through this handout separately, but it may help you to keep everything straight.

I. Pooled (Constrained) Model. Blacks and whites are combined into a single analysis, hence the coefficient estimates are constrained to be the same for both racial groups, i.e. the intercepts and the effect of education and job experience are the same for both groups.

```
. use "https://www3.nd.edu/~rwilliam/statafiles/blwh.dta"
. reg income  educ jobexp

      Source |       SS       df       MS              Number of obs =     500
-------------+------------------------------           F(  2,   497) = 1103.96
       Model |  32798.4018     2  16399.2009           Prob > F      =  0.0000
    Residual |  7382.84742   497  14.8548238           R-squared     =  0.8163
-------------+------------------------------           Adj R-squared =  0.8155
       Total |  40181.2493   499  80.5235456           Root MSE      =  3.8542

------------------------------------------------------------------------------
      income |      Coef.   Std. Err.      t    P>|t|     [95% Conf. Interval]
-------------+----------------------------------------------------------------
        educ |    1.94512   .0436998    44.51   0.000     1.859261     2.03098
      jobexp |   .7082212   .0343672    20.61   0.000     .6406983     .775744
       _cons |  -7.382935   .8027781    -9.20   0.000    -8.960192   -5.805678
------------------------------------------------------------------------------
```

II. Unconstrained Models: Separate Models for each group. Here, we estimate separate models, first for whites, then blacks. This makes it possible for the intercepts and slope coefficients to freely differ across populations. This is equivalent to Model IIIC.

Whites:

```
. reg income educ jobexp if black == 0

      Source |       SS       df       MS              Number of obs =     400
-------------+------------------------------           F(  2,   397) =  620.07
       Model |  18361.9894     2  9180.99472           Prob > F      =  0.0000
    Residual |  5878.16991   397  14.8064733           R-squared     =  0.7575
-------------+------------------------------           Adj R-squared =  0.7563
       Total |  24240.1594   399  60.7522791           Root MSE      =  3.8479

------------------------------------------------------------------------------
      income |      Coef.   Std. Err.      t    P>|t|     [95% Conf. Interval]
-------------+----------------------------------------------------------------
        educ |   1.893338   .0562591    33.65   0.000     1.782735    2.003941
      jobexp |    .722255   .0412236    17.52   0.000     .6412111    .8032988
       _cons |  -6.461189   1.089219    -5.93   0.000    -8.602547    -4.31983
------------------------------------------------------------------------------
```

Blacks:

```
. reg income educ jobexp if black == 1

      Source |       SS       df       MS              Number of obs =     100
-------------+------------------------------           F(  2,    97) =  267.80
       Model |  4924.27286     2  2462.13643           Prob > F      =  0.0000
    Residual |   891.81705    97  9.19399021           R-squared     =  0.8467
-------------+------------------------------           Adj R-squared =  0.8435
       Total |  5816.08991    99   58.748383           Root MSE      =  3.0322

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      income |      Coef.   Std. Err.      t    P>|t|     [95% Conf. Interval]
-------------+----------------------------------------------------------------
        educ |   1.677949   .0725479    23.13   0.000     1.533962    1.821936
      jobexp |    .421975   .0581021     7.26   0.000     .3066585    .5372915
       _cons |    -3.0512   1.154604    -2.64   0.010    -5.342771   -.7596302
------------------------------------------------------------------------------
```

Doing an incremental F Test (see earlier notes for details):

$$F_{K+1,N_1+N_2-2K-2} = \frac{(SSE_c - SSE_u)*(N_1+N_2-2K-2)}{SSE_u*(K+1)} = \frac{(7383-6770)*494}{6770*3} = 14.91$$

III. Unconstrained and Partially Unconstrained Models: Dummy Variables and Interaction Effects. In this approach, interaction effects and dummy variables are used to allow for racial differences in parameter effects. Interaction effects allow more flexibility in model specification.

IIIa. Intercepts vary across groups, all other parameters the same. To do this, you regress Y on the IVs and include a dummy variable for race.

```
. reg income educ jobexp i.black

      Source |       SS       df       MS              Number of obs =     500
-------------+------------------------------           F(  3,   496) =  787.14
       Model |  33206.4588     3  11068.8196           Prob > F      =  0.0000
    Residual |  6974.79047   496  14.0620776           R-squared     =  0.8264
-------------+------------------------------           Adj R-squared =  0.8254
       Total |  40181.2493   499  80.5235456           Root MSE      =  3.7499

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      income |      Coef.   Std. Err.      t    P>|t|     [95% Conf. Interval]
-------------+----------------------------------------------------------------
        educ |   1.840407   .0467507    39.37   0.000     1.748553    1.932261
      jobexp |   .6514259   .0350604    18.58   0.000     .5825406    .7203111
     1.black |   -2.55136   .4736266    -5.39   0.000    -3.481921   -1.620798
       _cons |   -4.72676   .9236842    -5.12   0.000    -6.541576   -2.911943
------------------------------------------------------------------------------

. * Wald test of differences in intercepts
. testparm i.black

 ( 1)  1.black = 0

       F(  1,   496) =   29.02
            Prob > F =    0.0000
```

IIIb. Intercepts and some slopes vary across groups, other slopes are the same.
Regress Y on the IVs, a dummy variable for race, and (in this example) one interaction term:

```
. reg income educ jobexp i.black i.black#c.jobexp

      Source |       SS       df       MS              Number of obs =     500
-------------+------------------------------           F(  4,   495) =  604.39
       Model |  33352.2559     4  8338.06397           Prob > F      =  0.0000
    Residual |  6828.99339   495  13.7959462           R-squared     =  0.8300
-------------+------------------------------           Adj R-squared =  0.8287
       Total |  40181.2493   499  80.5235456           Root MSE      =  3.7143

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      income |      Coef.   Std. Err.      t    P>|t|     [95% Conf. Interval]
-------------+----------------------------------------------------------------
        educ |   1.834776   .0463385    39.60   0.000     1.743732    1.925821
      jobexp |   .7128145   .0395293    18.03   0.000     .6351486    .7904805
     1.black |   .4686862   1.040728     0.45   0.653    -1.576103    2.513475
             |
black#c.jobexp |
           1 |  -.2556117   .0786289    -3.25   0.001    -.4100993   -.1011242
             |
       _cons |  -5.514076   .9464143    -5.83   0.000    -7.373561   -3.654592
------------------------------------------------------------------------------

. * Wald test of racial difference in effect of jobexp
. testparm i.black#c.jobexp

 ( 1)  1.black#c.jobexp = 0

       F(  1,   495) =   10.57
            Prob > F =    0.0012
```

IIIc. Intercepts and all slopes free to vary across groups (totally unconstrained). Equivalent to Model II, where we estimated separate models for each group.

```
. reg income educ jobexp i.black i.black#c.educ i.black#c.jobexp

      Source |       SS       df       MS              Number of obs =     500
-------------+------------------------------           F(  5,   494) =  487.60
       Model |  33411.2623     5  6682.25246           Prob > F      =  0.0000
    Residual |  6769.98696   494  13.7044271           R-squared     =  0.8315
-------------+------------------------------           Adj R-squared =  0.8298
       Total |  40181.2493   499  80.5235456           Root MSE      =  3.7019

------------------------------------------------------------------------------
      income |      Coef.   Std. Err.      t    P>|t|     [95% Conf. Interval]
-------------+----------------------------------------------------------------
        educ |   1.893338    .054125    34.98   0.000     1.786994    1.999681
      jobexp |    .722255   .0396598    18.21   0.000     .6443322    .8001777
     1.black |   3.409988   1.756477     1.94   0.053    -.0410984    6.861075
             |
black#c.educ |
          1  |  -.2153886   .1038015    -2.08   0.039    -.4193354   -.0114418
             |
black#c.jobexp |
          1  |  -.3002799   .0812705    -3.69   0.000    -.4599584   -.1406015
             |
       _cons |  -6.461189     1.0479    -6.17   0.000    -8.520079   -4.402298
------------------------------------------------------------------------------

. * Wald test of differences in slopes
. testparm i.black#c.educ i.black#c.jobexp

 ( 1)  1.black#c.educ = 0
 ( 2)  1.black#c.jobexp = 0

       F(  2,   494) =    7.47
            Prob > F =    0.0006

. * Wald test of any differences across groups, including intercepts
. testparm i.black i.black#c.educ i.black#c.jobexp

 ( 1)  1.black = 0
 ( 2)  1.black#c.educ = 0
 ( 3)  1.black#c.jobexp = 0

       F(  3,   494) =   14.91
            Prob > F =    0.0000
```

Note that $N_u = 500$, $SSE_u = 6770$, $DFE_u = 494$. These are the exact same numbers we got for the incremental F test using the Model II procedure of estimating separate models for each racial group. Further, the regression coefficients estimated under this approach can easily be converted to the betas estimated under the previous approach, and vice versa.