help fairlie

<u>Title</u>

fairlie — Nonlinear decomposition of binary outcome

Syntax

```
fairlie depvar indepvars [if] [in] [weight], by(groupvar) [ options ]
where the syntax for indepvars is
   term [term ...]
with term as a variable name or, alternatively, a set of variables specified as
  ([name:] varlist)
```

name is any valid Stata name and labels the set. If name is omitted, the name of the first variable is used to

| options | Description |
|------------------------------|--|
| by(groupvar) | specify the groups (required); <i>groupvar</i> must be 0/1 |
| <u>r</u> eps(#) | number of decomposition replications; default is 100 |
| nodots | suppress the replication dots |
| ro | randomize ordering of variables in the detailed decomposition |
| <u>ref</u> erence(#) | specify the reference model; # must be 0 (use group 0 model) or 1 (use group 1 model); defau |
| <pre>pooled[(varlist)]</pre> | use a pooled model as reference; <u>varlist</u> is added to the model if specified |
| probit | use a probit model; default is to use a logit model |
| <u>noe</u> st | suppress model estimation output |
| <u>save</u> est(name) | store model estimation results under <i>name</i> |
| <u>l</u> evel(#) | set confidence level; default is level(95) |
| <u>noleg</u> end | suppress legend |
| estopts | options passed through to the internal call of <u>logit</u> or <u>probit</u> |

fweights, pweights, and iweight are allowed; see $\underline{\text{weight}}.$

Description

fairlie computes the nonlinear decomposition of binary outcome differentials proposed by Fairlie (1999, 2003, 2003), 2003, 200

The implementation of the decomposition technique closely follows the suggestions provided by Fairlie (2003). In computation of the detailed decomposition (see below).

The decomposition technique involves one-to-one matching of cases between the two groups. If the groups have directly depend on the specific sample, the process is repeated and mean results are reported. Use reps() to specificable results; see help <u>set seed</u>.

The separate contributions from independent variables or groups of independent variables may be sensitive to the **ro** option described below to randomize the ordering of variables, thus approximating results over all possible to the contribution of the contribution

Alternative decomposition approaches for binary response variables are provided, e.g., by Gomulka and Stern (199

Algorithm for weighted data: The algorithm by Fairlie for the detailed decomposition is based on matching observandom sample (without replacement) from the larger group. The goal of the matching is to generate a hypothetical the first group and some from the second group. In the case of weighted data, the original algorithm cannot be variables in the hypothetical sample. However, an appropriate hypothetical sample can be constructed by matching proportional to the weights. In the present implementation the sizes of the two sub-samples are set to half the replacement. The choice of the sub-sample size is arbitrary but that does not matter much since the precision observations, which is a function of the sub-sample size and the number of decomposition replications as set by counterbalanced by an increase (a decrease) in the number of replications. The results from the original algorithm but they have the same expectation if the weights are uninformative (i.e. if the weights are equal for all observations).

Options

by(groupvar) defines the groups between which the decomposition is to be performed. groupvar must be 0/1.

reps(#) specifies the number of decomposition replications to be performed. The default is 100.

nodots suppresses the display of replication dots.

ro causes the ordering of variables to be randomized in the detailed decomposition. The default is to estimate a groups of independent variables) one after another in the specified order. Note that results are sensitive to the variables in each replication and, therefore, approximate average results over all possible orderings. The variables is also be useful to increase the number of replications when using this option.

reference(#) specifies the reference estimates to be used with the decomposition. reference(0), the default, inc reference(1) specifies that the coefficients from the groupvar==1 model are used.

pooled[(varlist)] specifies that the coefficients from the pooled model over all cases be used for the decompose restricted to the non-missing cases of groupvar. This is reasonable because it is sometimes desirable to in sample of the pooled model.) Optionally, varlist will be added as additional control variables to the pooled important that the reference group in the pooled model is the group for which groupvar==0.

probit specifies that the probit command is used for model estimation. The default is to use logit.

noest suppresses the display of the model estimates.

saveest(name) stores the model estimation results under name using estimates store.

level(#); see estimation options.

nolegend suppresses the legend for the variable sets.

estopts are options passed through to the internal call of <u>logit</u> or <u>probit</u>.

Examples

- . use http://fmwww.bc.edu/RePEc/bocode/h/homecomp.dta

- . generate black2 = black==1 if white==1|black==1

Saved Results

```
Scalars
                number of observations
 e(N)
 e(N_0)
                number of obs for which groupvar==0
                number of obs for which groupvar==1
 e(N_1)
 e(N_match)
                sample size used for one-to-one matching
                number of decomposition replications
 e(reps)
 e(pr_0)
                outcome probability for groupvar==0
 e(pr_1)
                outcome probability for groupvar==1
 e(diff)
                differential e(pr_0)-e(pr_1)
 e(expl)
                total contribution of group differences in regressors
```

Macros fairlie e(cmd) name of dependent variable e(depvar) e(by) name group variable e(_cmd) command used for model estimation (logit or probit) weight type e(wtype) weight expression e(wexp) e(reference) reference estimates (0, 1, or pooled) e(legend) definitions of regressor sets e(ro) ro, if the random order option was specified e(properties) b V Matrices e(b) detailed decomposition results e(V) variances for **e(b)** (covariances are set to zero) e(_b) reference coefficients e(_V) variance-covariance matrix of e(_b) Functions e(sample) marks estimation sample

References

Fairlie, Robert W. (1999). The Absence of the African-American Owned Business: An Analysis of the Dynamics of Se DOI:10.1086/209914

Fairlie, Robert W. (2003). An Extension of the Blinder-Oaxaca Decomposition Technique to Logit and Probit Models http://ideas.repec.org/p/egc/wpaper/873.html

Fairlie, Robert W. (2005). An extension of the Blinder-Oaxaca decomposition technique to logit and probit model: DOI:10.3233/JEM-2005-0259

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Also see

Online: help for logit, probit, set seed