

## help apc

## <u>Title</u>

apc — Estimating age, period and cohort effects

## <u>Syntax</u>

apc depvar [if] [in], age(varname) period(varname) cohort(varname) [options]

options	description
Model	
<pre>method(identmethod)</pre>	<i>identmethod</i> can be <b>dp</b> or <b>me</b> ; default is <b>dp</b>
<u>l</u> owerbound(#)	lower bound of dependent variable in the population; must be supplied if method is <b>me</b>
upperbound(#)	upper bound of dependent variable in the population; must be supplied if method is <b>me</b>
Estimation	
<pre>technique(opttech)</pre>	opttech can <b>nr</b> , <b>dfp</b> , <b>bfgs</b> or <b>nm</b> ; default is <b>nr</b>
ptol(#)	<pre>ptol controls the optimization routine; default is ptol(1e-10)</pre>
vtol(#)	<pre>vtol controls the optimization routine; default is vtol(1e-10)</pre>
<pre>maxiter(#)</pre>	<pre>maxiter controls the optimization routine; default is maxiter(2000000)</pre>
Reporting	
<pre>graph(graphopt)</pre>	controls whether to produce and save graphs; graphopt must be either <b>off</b> or <b>on, replace</b> ; default is <b>off</b>

#### Description

**apc** estimates the coffecients of the linear age-period-cohort (apc) model with dependent variable *depvar*.

The apc model aims to separate, for some outcome of interest, those influences associated with the process of aging, from those influences associated with the date at which subjects are observed, from those influences associated with a subject's date of birth. In the linear apc model, whose coefficients are estimated by **apc**, the dependent variable is considered to be a linear additive fuction of dummies representing each age, period and cohort value observed in the sample.

The linear apc model suffers from a well-known fundamental identification problem, namely that there is a perfect linear relationship between these effects: current year equals year of birth plus age (in years).

**apc** estimates the coefficients using two methods that have been proposed. The first method is that of Deaton & Paxson (1994) which identifies the coefficients by imposing the restriction that the period effect is orthogonal to a trend and sums to zero. The second method is that of Browning, Crawford & Knoef (2010) which is based on a maximum entropy method and can be applied when the dependent variable has a finite support in the population.

## **Options**

\_\_\_\_ Model [

 $d\mathbf{p}\colon$  the parameters are identified using the restriction proposed by Deaton & Paxson (1994). This is the default.

**me:** the parameters are chosen based on a maximum entropy method proposed by Browning, Crawford & Knoef (2010). This method is applicable only if the dependent variable has a finite support in the population.

lowerbound(#) The lower bound of the dependent variable in the population. Must be specified if the method used is me.

upperbound(#) The upper bound of the dependent variable in the population. Must be specified if the method used is me.

Estimation

technique(opttech) If the method of estimation is me then, then the parameters are selected in order to maximise the entropy function. For details see Browning, Crawford & Knoef (2010). This maximisation uses STATA's optimize() routine. The optimzation technique can be one of:

technique	description
nr	modified Newton-Raphson
dfp	Davidon-Fletcher-Powell
bfgs	Broyden-Fletcher-Goldfarb-Shanno
nm	Nelder-Mead

The default is **nr**. See <u>optimize init technique()</u> for more details

- ptol(#) Controls the convergence criterion used by <u>optimize()</u>. See <u>optimize init conv ptol()</u>. Default is ptol(1e-10).
- vtol(#) Controls the convergence criterion used by <u>optimize()</u>. See <u>optimize init conv vtol()</u>. Default is vtol(1e-10).
- maxiter(#) Sets the maximum number of iterations allowed by <u>optimize()</u>. See <u>optimize init conv maxiter()</u>. Default is maxiter(2000000).

Reporting

graph(graphopt)} controls whether to produce and save graphs. Allowed options
forgraphopt are:

off No graphs are displayed or saved. This is the default.

on, replace If this option is specified, four graphs are produced and saved in the current working directory. These are: a graph of the trajectory of the age effects, the period effects, the cohort effects and all three combined. 95% confidence intervals are included on the graphs. The first three of these are saved as *varname.gph* where *varname* is the name of the age, period or cohort variable in the data. The graph that shows all three tractories is saved as *apc.gph*. These saved graphs will overwrite graphs of the same name in the current working directory - so for graphs to be produced *replace* has to specified as part of *graphopt*.

#### **Examples**

These examples use the STATA example data set union.dta. The dependent variable is grade completed. This dependent variable has bounds in the population (between 0 and 18), so the maximum entropy method can be used.

Open data

. webuse union, clear

Generate a cohort variable as year less age . gen cohort = year - age Estimating coefficients using maximum entropy approach
 . apc grade, age(age) period(year) cohort(cohort) method(me) lower(0)
 upper(18)

Estimating coefficients using maximum entropy approach; displaying and saving graphs

. apc grade, age(age) period(year) cohort(cohort) method(me) lower(0)
 upper(18) graph(on, replace)

Estimating coefficients using Deaton and Paxson's method; displaying and saving graphs (note, no bounds required)

. apc grade, age(age) period(year) cohort(cohort) method(dp) graph(on, replace)

Estimating coefficients using Deaton and Paxson's method; restricting the sample
to those aged less than or equal to 30
 . apc grade if age<=30, age(age) period(year) cohort(cohort) method(dp)</pre>

### Saved results

apc saves the following in e():
Scalars
 e(N) number of observations
Macros
 e(cmd) apc
Matrices
 e(b) coefficient vector
 e(V) variance-covariance matrix of the estimators
Functions
 e(sample) marks estimation sample

## A note on the identification of the apc coefficients

The linear dependence that results in the identification problem comes from the fact that "age + year of birth = current year". However, some characteristics of the data, if present, can result in that equation not holding for everyone in the data. In such a case the coefficients are identified without further restriction. **apc** checks whether the parameters are identified without restriction and if so returns a message informing the user that this is the case. Estimation then proceeds using whichever method the user specified (i.e. **dp** or **me**). Whether one proceeds with a method such as Deaton and Paxson's or the maximum entropy approach or one switches to OLS (or similar) depends on whether the identification restrictions (which may be implicit in the data setup rather than explicitly imposed by the user) are credible.

Two common reasons for the linear apc coefficients being identified without further restrestrictions are:

1. One of the age, period or cohort variables is denominated using a different time unit to the others. For example, age and period could be represented by dummies representing single years while cohort could be a dummy representing a particular decade. The identifying assumption here is that the cohort effect is the same for all those born in a particular decade. If the user is satisfied that this is a credible assumption, then OLS (or similar) can be used.

2. Due to the time of sampling the equation "age + year of birth = current year" might not hold. For example, if an individual is born in December 1990 (so cohort = 1990), and they are surveyed in January 2012 (so period = 2012), their age in the data is 21. For this individual the equation above doesn't hold. The linear apc coefficients can be identified using OLS. Identification is coming from variation in ages between individuals born in the same year. It will, in many applications, be difficult to justify this as an identification assumption.

## Citing apc

The use of **apc** can be cited as follows:

O'Dea, Cormac. 2012. "APC: Stata module to estimate age, period and cohort effects". Available at: <u>www.ifs.org.uk/publications/5998</u>.

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### **References**

- Browning, M., Crawford, I. and Knoef, M. 2010. "The Age-Period-Cohort Problem; Set identification and point identification". Working Paper. http://www.economics.ox.ac.uk/members/ian.crawford/papers/apc.pdf
- Deaton, Angus, and Christina Paxson. 1994. "Saving, Growth, and Aging in Taiwan," in D. Wise (ed) Studies in the Economics of Aging, pp. 331'362 National Bureau of Economic Research.