

Sequential characteristics of Wooldridge's estimators of a dynamical random effects probit model

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In a balanced panel of observations of  $N$  individuals through  $T$  equispaced times  $t_1 < t_2 < \dots < t_T$  with respect to a binary response variable  $Y$  and a set of  $p$  exogenous variables  $X_1, X_2, \dots, X_p$ , a conditional dynamic probit model with random effects  $e_{it} = a_i + u_{it}$  may be expressed for individual  $i$  at time  $t$  as  $P(Y_{it}=1) = \Phi(X_{it}\beta + \gamma Y_{i(t-1)} + a_i)$  for  $i=1, \dots, N$ ;  $t=t_1, t_2, \dots, t_T$ , with  $\Phi(u)$  denoting the standard Normal distribution function at  $u$ . In this context, with the additional hypotheses of i) independence of the sequence  $u_{i1}, \dots, u_{it}$ , given  $a_i$ , ii) a Normal distribution for  $a_i$  and iii) a specific assumption relative to the initial state  $Y_{i0}$ , a marginal likelihood may be written and maximum marginal likelihood estimates can be computed, by numerical maximization. The complexity of this issue, leads to different approaches to its solution, as one from Stewart (2006), proposing the use of maximum simulated likelihood, and that from Wooldridge (2005) who points out a simple solution to this task. In a prospective study, it may be useful to plan interim analysis of the observations at  $K$  specified time points  $(t_{mj})_{j=1..K}$  with  $t_{m1} < t_{m2} < \dots < t_{mK} = T$ . At each of these time points, parameters estimates of the model may be obtained using standard statistical software for a static random effects probit model, after an appropriate manipulation of the data. In this paper, after having developed a code in R software for fitting a static random effects probit model, and making use of it for fitting a dynamic probit model following Wooldridge's approach, we examine the  $K \times K$  empirical covariance matrix of the estimators of one specific regression coefficient, obtained in the  $K$  interim analysis. This study is pursued through bootstrap methods, following the same steps as Bole, Rebeck (2004). The objective is to test if the respective theoretical covariance matrix has a structure of independent increments, in order to be possible to apply the group sequential methodology as in Spiessens et al (2002).

Bole, V.A., Rebec, P. (2004). Bootstrapping the Hausman Test in Panel Data Models. Available at SSRN: <http://ssrn.com/abstract=628321>

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