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Accounting for Latent Cropping Management Practices Choices in Crop Production Models: a Random Parameter Hidden Markov Model Approach

Esther DEVILLIERS, Philippe Obafèmi KOUTCHADÉ, Alain CARPENTIER



General Introduction

The Econometric Framework

Estimation Results

Conclusion



From the economist point of view on pesticide reduction...

- Pesticides are still abundant in the European agriculture despite a long-known pesticide noxiousness (e.g., Pimentel and Greiner, 1997; Pimentel et al., 1992; Wilson, 1999);
- European Directive 2009/128/EC: EU members have to implement National Action Plans (NAP) to reduce their pesticide use ;
- Pesticide taxation scheme:
 - Standard economists' answer to negative externality (e.g., Aubertot et al., 2005; Lichtenberg, 2004);
 - Limited impact in Norway, Denmark, Sweden and France (Böcker and Finger, 2016);
 - Low pesticide demand elasticity (*e.g.*, Böcker and Finger, 2017; Skevas et al., 2013).



...to the agronomist point of view

Cropping management practices (CMP) = ordered sequence of yield production decisions made by the farmer along the crop season to achieve a targeted yield.

- Farmers cropping decisions are a coherent set ;
- In a given set of CMP, only small adjustments of input use are possible ;
- Pesticides are the keystone of conventional practices (Aubertot et al., 2007).

Pesticide reduction requires a change in CMP practices towards low-input production practices \rightarrow we need to account for CMPs in economists' production functions.



An application to the case of French winter wheat producers

- Cost accounting data on 1351 farmers from 1998 to 2014 situated in la Marne area ;
- Data on observed yields and wheat prices, on pesticide and fertilizer expenditures ;
- No information on cropping management practices ⇒ latent CMPs.

A double objective:

- Uncover CMPs, their characteristics and adoption ;
- Investigate into policies to encourage the adoption of low-input practices.



A brief literature review

An abundant literature on technology adoption:

- Heterogeneous economic returns (e.g., Griliches, 1957; Michler et al., 2018; Suri, 2011);
- Adoption as a social process: information exchanges (e.g., Foster and Rosenzweig, 2010; Ryan and Gross, 1943), imitation (e.g., Rogers, 1962);
- Other factors as technical efficiency (*e.g.*, Kumbhakar et al., 2009), labour-constraints (*e.g.*, Fernandez-Cornejo et al., 2005), risk & uncertainty (*e.g.*, Chavas and Nauges, 2020), social concerns (*e.g.*, Mzoughi, 2011).
- A literature on latent class stochastic frontier models (Alvarez and Corral, 2010; Dakpo et al., 2021; Greene, 2005; Martinez Cillero et al., 2018; Orea and Kumbhakar, 2004).





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A parametric approach to uncover farmers' CMP choice

♦ Model for latent wheat yield (y) and input uses (x_k, k ∈ {1,...,K}) levels:

For
$$c \in \{1, ..., C\}$$
,
$$\begin{cases} y_{it}^c = b_{y,i}^c + d_{y,t,0} + \delta'_{y,0} \mathbf{z}_{it} + \varepsilon_{y,it}^c \\ \mathbf{x}_{it}^c = \mathbf{b}_{x,i}^c + \mathbf{d}_{x,t,0} + \mathbf{\Delta}_{x,0} \mathbf{z}_{it} + \varepsilon_{x,it}^c \end{cases}$$
;

i =farmer index, t =time index, c =CMP index

(b^c_{y,i}, b^c_{x,i}) are random terms accounting for CMP effects ;
A decreasing intensity parametrization:

For
$$c \in \{2, ..., C\}$$
, $b_{y,i}^c = b_{y,i}^1 \prod_{d=2}^c a_{y,i}^d$,

with $b_{y,i}^1 \ge 0$ the targeted yield of the most intensive CMP, and $\forall c > 1, a_{y,i}^c \in [0, 1]$.



A Markov model to describe the dynamics of CMP choice

- No sunk costs as for an investment choice as irrigation technology → easy to switch ;
- ✤ Intangible learning and opportunity costs → farmers tend to keep the same CMP across year ;
- Probability for farmer i at time t to choose CMP d given that previous CMP was c:

$$\frac{\exp\left(\rho_i(\pi_{it}^d - \mu_i^{d|c})\right)}{\sum_{k \in C} \exp\left(\rho_i(\pi_{it}^k - \mu_i^{k|c})\right)},$$

with $\pi_{it}^{c} = w_{y,t-1}b_{y,i}^{c} - w'_{x,t}b_{x,i}^{c} + \eta_{it}$;

 π_{it}^{d} represents the expected return of CMP *d*, $\mu_{i}^{d|c}$ represents farm-specific switching costs.





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Agronomic characteristics of the three uncovered CMPs



Figure: Mean input uses and yield differences in percentage across CMPs, north eastern France, 1999-2014



Economic characteristics of the three uncovered CMPs



Figure: Estimated annual mean expected return per CMP type, north eastern France, 1999–2014



Expected return variations are mainly attributed to wheat price variations



— Wheat price (€/t)

Figure: Wheat annual prices, France, 1999-2014



Adoption of low-input CMP driven by non-economic factors



Figure: Estimated annual share of farmers per CMP type, north eastern France, 1999-2014



Impact of a 100% chemical inputs tax on CMP adoption



Figure: Mean annual change in the CMP adoption share after simulating a 100% tax on chemical inputs, north eastern France, 1999-2014





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Main results and potential extensions

- Uncovered CMPs have similar characteristics to agronomists' CMPs (Loyce et al., 2012; Rolland et al., 2003);
- Compared to the revenue loss associated to yield losses, the input savings are too small so that a taxation scheme encourage low-input adoption ;
- Investigate into the non-economic motives that are impacting low-input adoption to design more efficient public policies.



Supplementary material

Sketch of the estimation procedure

- A <u>fully parametric</u> <u>random parameter hidden Markov model</u> Maximum Likelihood Stochastic Approximation of Expectation-Maximization
 - \Rightarrow **SAEM** algorithm (Delyon et al., 1999);
- Hidden Markov model consists in solving "three basic problems" (Lavielle, 2018; Rabiner, 1989):
 - 1. Training problem: estimate the vector of model parameters to maximize the likelihood ;
 - 2. Evaluation problem: evaluate the likelihood of the data given the model \Rightarrow Forward-Backward algorithm ;
 - Decoding problem: estimate the most probable sequence of hidden state corresponding to the observed data ⇒ Viterbi algorithm.

Annual mean yields for the three CMP categories



Figure: Annual mean yield per CMP type, north eastern France, 1999-2014

Annual mean nitrogen use for the three CMP categories



Figure: Annual mean nitrogen use per CMP type, north eastern France, 1999-2014

Annual mean pesticide use for the three CMP categories



Figure: Annual mean pesticide use per CMP type, north eastern France, 1999-2014

Simulation results: a 10% price premium



Figure: Mean annual change in the CMP adoption share after simulating a 10% premium on wheat prices, north eastern France, 1999-2014

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