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# **Farmer Participation, Entry and Exit decisions in the Italian Crop Insurance Program**

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30 January 2016

Online at <https://mpra.ub.uni-muenchen.de/69508/>  
MPRA Paper No. 69508, posted 12 Feb 2016 23:00 UTC

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12 **Abstract**

13 The factors affecting the demand for agricultural insurance in the US have been extensively studied over the  
14 last two decades. However, the determinants of a farm's entry and exit decisions in the insurance market have  
15 received relatively little attention. Turnover in the insurance book of business is an important issue in most private  
16 and public crop insurance plans. Moreover, insurance markets in the EU are still largely under-investigated. We  
17 investigate empirically the determinants of crop insurance participation in Italy. We show that the participation rate is  
18 high for large firms and that it is negatively correlated with crop diversification, which is itself a form of insurance.  
19 High premiums tend to inhibit both entry and exit from the insurance market. Larger and wealthier farms are more  
20 likely to adopt insurance and renew coverage over time. We discuss implications of our results for public intervention  
21 and the private industry. In particular, we demonstrate that the decision to drop coverage by an insured grower may  
22 differ significantly from the corresponding decision to enroll in an insurance program by an uninsured farmer. To the  
23 extent that policymakers want to encourage participation in subsidized crop insurance programs, education and  
24 outreach efforts toward uninsured farmers may differ substantially from those directed toward keeping insured  
25 farmers enrolled in the program. We investigate these differences.

26  
27 **Keywords:** Crop insurance, Entry, Exit, Participation, Turnover, Italy

28 **JEL:** G22, Q12, Q18  
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# Farmer Participation, Entry and Exit Decisions

## in the Italian Crop Insurance Program

### Introduction

Over the last several decades, risk management policies in agriculture have been significantly modified. In Italy, the *Fondo di Solidarietà Nazionale* (FSN) was developed in the 1970s and was intended to compensate farmers who had been affected by natural disasters. This policy, which has played a prominent role in Italian agriculture, is now mainly regulated by Legislative Decree No. 102/2004 which subsidizes insurance contracts (Cafiero *et al.*, 2007). The market for insurance in Italy is evolving rapidly and there is considerable interest in understanding the operation of the program and in monitoring farmers' participation over time. In fact, although the budget for the FSN has never been limited, the type of available contracts and the set of subsidized policies have increased over time. At the same time, participation has been stable over time.

Policymakers often act to encourage participation in crop insurance programs, most often through the use of large subsidies. However, such promotion requires an understanding of participation as well as entry and exit decisions. We investigate the demand for crop insurance using individual models of participation, entry and exit decisions. We seek to inform policymakers by providing an understanding of the determinants of turnover in insurance markets that may affect participation in crop insurance programs.

Contracts that cover losses from multiple risks have also increased in prominence around the world. Between 2003 and 2009, the share of single-peril insurance contracts, which mainly compensate losses from hail, has declined in Italy by fifty percent while the share of multiple risk contracts has increased substantially. Under the current Italian insurance program, farmers receive a premium subsidy of up to 80% to insure a farm's production against losses larger than thirty percent of the historical average level of production. In the EU, empirical evidence on the effects of subsidies on participation rates in insurance programs is not clear (Garrido and Zilberman, 2008; Finger and Lehmann, 2012), and their effects

1 are often debated (see, for example, Bakhshi and Gray, 2012; and Di Falco et al., 2014). The Italian case is of  
2 particular interest for a number of reasons. Participation is low despite the Italian government's subsidy  
3 being one of the highest in world (*cfr.* Mahul and Stutley, 2010). In Italy the vast majority of contracts are  
4 purchased by farms located in Northern Italy rather than in other parts of the country (European  
5 Commission, 2009; Enjolras *et al.*, 2012). This is a consequence of the structure of insurance premium rates  
6 in the North, where the typical loss ratio (the ratio of indemnity payments to premiums) is closer to unity.  
7 In contrast, the southern part of Italy has a loss ratio of about one half. While greater insurance returns to  
8 farmers may well explain greater participation in the north than in other regions, geographically-distinct  
9 farmers also face different sources of risk. Moreover, insurance contracts are far from being widely  
10 adopted as a stable tool of risk management in Italy. We observe that few farms carry insurance for more  
11 than two consecutive years. Understanding the factors underlying this high turnover rate has important  
12 implications for the operation of the programs since, despite large subsidies, participation in crop insurance  
13 is both limited and volatile. In order to increase participation it is important not only to stimulate entry but  
14 also to encourage insurance renewal and thus inhibit exits from the program. The determinants of these  
15 decisions have not been yet fully explored.

16 The demand for crop insurance in U.S. has received significant empirical attention in a large  
17 number of empirical studies (*e.g.* Goodwin, 1993, Goodwin and Smith, 2013; Skees and Reed, 1986; Smith  
18 and Goodwin, 1996; Sherrick *et al.* 2004). Goodwin (1993) shows that land size, land value, and a corporate  
19 farm structure have positive effects on insurance demand. Coble et al. (1997) conclude that the higher the  
20 expected return to insurance, the higher the adoption rate. Education, farm experience, debt and disaster  
21 payments have also been shown to be associated with the adoption of crop insurance (Goodwin and  
22 Kastens, 1993; Smith Baquet, 1996)<sup>2</sup>.

23 The empirical literature on crop insurance in EU Countries is also rich, though turnover has not  
24 been explicitly investigated. An important analysis by Garrido and Zilberman (2008) shows that premium  
25 subsidies are the leading factor that increases the probability of using insurance in Spain. In contrast, Finger  
26 and Lehmann (2012) show that support to farmers' incomes tends to decrease insurance adoption rates in

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<sup>2</sup> An extensive survey on the determinants of crop insurance adoption is provided by Knight and Coble (1997).

1 Switzerland. Cabas *et al.* (2008) model the entry and exit decisions using panel data consisting of the total  
2 number of insured and uninsured farmers at the county level. They find that insured farmers are more  
3 sensitive than uninsured farmers to changes in the preceding year's yield. Moreover, participation is  
4 positively related to yield variability, entry and exit decisions are, respectively, positively and negatively  
5 affected. Their analysis of entry and exit decisions at an aggregate level provides an interesting benchmark.  
6 In an empirical study of insurance participation in France, Enjoras and Sentis (2011) show that the highest  
7 risk farms are more likely to purchase insurance. They also note that the existing empirical literature has  
8 largely focused on studies of aggregated data and highlight the potential importance of farm-level analyses.

9 A limited number of studies have analyzed the demand for crop insurance in Italy. Exploring the  
10 demand for insurance in Italy provides useful insights into policy interventions in Europe as a whole. In fact,  
11 lacking a common framework, European member states have autonomously adopted national policies for  
12 assisting farmers in dealing with production risks and natural disasters. These policy interventions, typically  
13 in the form of subsidies on crop insurance or agricultural solidarity funds, have been primarily adopted in  
14 the Southern EU countries (France, Greece, Italy and Spain). In contrast, public intervention in the United  
15 States and Canada aims at supporting farmers' management activities in a very broad sense by supporting  
16 farmers' revenue through hedge funds, revenue insurance programs, mutual funds, and weather indexes<sup>3</sup>.  
17 More important is the fact that the determinants of turnover (adopting and dropping coverage) have been  
18 given scant attention in the literature. The analysis conducted by Cabas *et al.* (2008) aims at filling this gap.  
19 However, the authors analyzed the phenomenon at an aggregate level, whereas we explicitly model farm-  
20 level decisions of whether to adopt, enter or exit the insurance market.

21 A better understanding of the factors driving participation, entry and exit decisions remains a  
22 pressing issue in order to enhance crop insurance coverage. Numerous questions arise. First, what are the  
23 frictions that limit participation in insurance contracts? Second, why are farmers reluctant to maintain

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<sup>3</sup> Detailed summaries of these plans are provided by Knight and Coble (1997), Coble and Dismukes (2007), and Capitanio (2010).

1 continuous coverage in the program? Third, what factors drive the adoption and dropping of coverage in  
2 the Italian crop insurance program, where turnover is an especially striking feature?<sup>4</sup>

3 We have two objectives. First, we investigate the factors and farm characteristics that are  
4 associated with participation in the Italian crop insurance program. Second, we evaluate the dynamics of  
5 participation patterns over the recent past and investigate the factors that are associated with the different  
6 participation rates observed in Northern and Southern Italy.

7

### 8 **Public Intervention in the Italian Crop Insurance Market**

9 Public intervention in agricultural risk management in Italy dates back to 1974, when the “Fondo di  
10 Solidarietà Nazionale in Agricoltura” (FSN) was instituted. The system has been reformed over time and  
11 currently conforms to the European Community guidelines for state aid in the agricultural sector  
12 concerning compensation for damages and insurance premium subsidies. Legislative Decree 102 in 2004  
13 defined new operational rules for the FSN and determined regulations on financial tools for risk  
14 management and capitalization incentives that favor agricultural firms.

15 Under the current FSN, two services are supplied: subsidies on insurance policies and ex-post  
16 payments. The two interventions are mutually exclusive in that crops and damages that are deemed  
17 insurable are not entitled to ex-post disaster compensation financed by the FSN. The latter regulates *ad hoc*  
18 compensation to farmers affected by damages. A key aspect of this policy intervention is that the  
19 occurrence of an exceptional event needs to be officially recognized by the central government prior to any  
20 compensation being made. Compensation is then calculated according to several criteria and usually  
21 reflects the availability of funds rather than the extent of damages<sup>5</sup>. During the last decade, actual losses  
22 and compensation paid to farmers have been poorly correlated. A further drawback of ex-post payments

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<sup>4</sup> Our analysis does not explicitly model turnover, but it provides insights on how farmers’ entry and exit decisions are influenced by several factors. For a broader discussion of turnover in the agricultural, financial and other sectors, the interested reader may refer to Bottazziet *al.*(2011), Cefis and Marsili (2012), and Hirsch and Gschwandtner (2013).

<sup>5</sup> Recent findings suggest that governments may use agricultural disaster relief payments as a political tool to favor their core supporters (Chang and Zilberman, 2014). Strengthening participation in insurance programs maybe economically more efficient.

1 relates to the time lags between the occurrence of the damaging events and compensation. These  
2 weaknesses have pushed policymakers to shift the bulk of the FSN to subsidies on crop insurance.

3 Currently, public intervention for crop insurance is also regulated by Legislative Decree No.  
4 102/2004<sup>6</sup>. Insurance policies (for crops and damages<sup>7</sup>) covered by the Annual Insurance Plan have, on  
5 average, received subsidies of about forty percent of total premiums in recent years. State subsidies apply  
6 to single-peril, combined/named perils, and multi-peril policies. The annual insurance plan defines the level  
7 of state intervention on the basis of public budget availability and the demand for crop insurance. Since  
8 2005, farmers have been required to take crop insurance for the whole area devoted to the insured that  
9 falls within the borders of their township. This regulation has stimulated the demand for crop insurance  
10 and in particular the subscription of collective policies through cooperatives and their operating  
11 consortiums, which operate as catalysts for demand.

12 Finally, under the current legislation farmers are allowed to create mutual funds in favor of specific  
13 crops and structures that are not included in the annual insurance plan. The payments from these funds are  
14 made only in the event of losses greater than thirty percent of total production.

15 Post-reform data have shown a limited increase in crop insurance participation rates. More  
16 specifically, the growth in total area insured has not been matched by a proportional diffusion of insurance  
17 contracts across new producers. Rather, expansion has been mainly motivated by the obligation to insure  
18 the entire cropped area for a given product. During the last decade, the state contribution has been growing  
19 in nominal terms, mainly due to a sharp growth in combined perils policies for which premiums are  
20 subsidized by up to eighty percent<sup>8</sup>. On the other hand, the share of contracts providing coverage only  
21 against hail damages (single-peril insurance) decreased from 92.0% in 2004 to 50.2% in 2010 (Table 1).

22 TABLE 1 ABOUT HERE

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<sup>6</sup>The Legislative Decree 102 has been published in the Official Journal 95 on April the 23<sup>rd</sup> of 2004 and is available at [www.camera.it](http://www.camera.it).

<sup>7</sup>Starting in 2006, insurance policies on losses arising from cattle diseases are subsidized.

<sup>8</sup>In particular, the subsidy is up to 80% of the cost of premium for policies against damages (reaching at least 30% of assured production) caused by adverse weather conditions and other natural disasters, and it is up to 50% of the cost of the premium if the insurance contract also covers other losses caused by adverse weather conditions that are not considered to be widespread natural disasters, or losses caused by animal or plant diseases.





1 focus on the entry and exit decisions of individual farmers. The data are collected to be representative of  
2 the entire population of Italian farms.

3 Assuming that farmers are price-takers and markets are perfectly competitive, a household chooses  
4 to adopt crop insurance based on expected utility:  $E[U(Insured)] > E[U(Uninsured)]$ . An uninsured  
5 (insured) farmer will choose to enter (exit) in the crop insurance market if the expected utility from  
6 entering(exiting) is greater than the expected utility of not entering (not exiting).

7 Our empirical investigation is conducted through a variety of probit models. Our first specification  
8 assesses participation in crop insurance programs, which is modeled as a time-varying binary variable  
9 representing the discrete insurance participation decision. The remaining models consider entry and exit in  
10 the insurance program. In particular, the entry and exit decisions are modeled using two dichotomous  
11 variables. The variable “entry” is equal to one if the farmer was not insured in time  $t-1$ , but purchased an  
12 insurance contract in time  $t$ . The model is estimated only for those observations for which the variable  
13 “insurance” had value equal to zero at time  $t-1$ . This model considers all farmers that were not insured and  
14 thus explains why some uninsured farmers purchased insurance in time  $t$ , while others did not. The variable  
15 “exit” is equal to one if the farmer purchased insurance in time  $t-1$  and did not purchase insurance in time  
16  $t$ . This model is estimated only for those farmers that were insured in time  $t-1$ . These models allow us to  
17 focus on the entry and exit decisions individually and thus permit farm and operator characteristics to have  
18 different effects on the entry and exit decisions<sup>10</sup>.

19 In order to take into account for the panel nature of our dataset, we condition on unobserved  
20 effects in estimation using the methods outlined by Wooldridge (2002):

$$21 \quad (1) \Pr(y_{it} = 1 | \mathbf{x}_i, \alpha_i) = \Pr(y_{it} = 1 | \mathbf{x}_{it}, \alpha_i) = \Phi(\alpha_i + \mathbf{x}'_{it}\beta) \quad \text{with } t = 1, \dots, T,$$

22 where the first equality states that the explanatory variables are exogenous, conditional on unobserved  
23 effects ( $\alpha_i$ ) so that the unobserved effects can be excluded from the RHS. The assumption allows us to omit  
24 lagged variables. The second equality is the standard assumption of probit models. We adopt a random

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<sup>10</sup> Note that, due to the relatively low participation rate in the insurance programs, the data set used to model the exit decision is smaller than that used to model the entry decision.

1 effects (RE) probit estimators and a fixed effects (FE) for each  $j$  model ( $j =$  participation, entry, exit). The  
 2 RE model assumes that the individual effects are normally distributed, that is  $\alpha_i | \mathbf{x}_i \sim N(0, \sigma_\alpha^2)$ :

$$3 \quad (2) P_j(Y_{ij,t} = 1 | X_{ij}, Z_{ij,t}, \alpha_{ij}) = \Phi_j(X_{ij}'\beta_j + Z_{ij,t}'\gamma_j + \alpha_{ij})$$

4 where  $Y_{i,t}$  is a binary dependent variable,  $X_{ij}$  represents a set of  $i^{th}$  firm-specific, time-invariant variables  
 5 and  $Z_{ij,t}$  reflects firm-specific time-varying variables, and  $\Phi(\cdot)$  is the standard normal *cdf*. The FE model,  
 6 which does not require distributional assumptions, has been estimated following the approach proposed by  
 7 Mundlak (1978): we added as additional explanatory variables the within-group means of the time-varying  
 8 covariates ( $Z_{i,t}$ ) to capture the correlation between the unobserved heterogeneity ( $c_i$ ) and the covariates.

9 The resulting specification is estimated as a random effects model:

$$10 \quad (3) P_j(Y_{ij,t} = 1 | X_{ij}, Z_{ij,t}, c_{ij}) = \Phi_j(X_{ij}'\beta_j + Z_{ij,t}'\gamma_j + \bar{Z}_i'\delta)$$

11 where  $E[c_i | X_i] = \bar{Z}_i'\delta$ . The three probit models consider the insurance participation decision, the decision  
 12 of an uninsured farmer in  $t-1$  to enroll in the program (entry) in time  $t$ , and the decision of an insured  
 13 farmer in time  $t-1$  to drop coverage in time  $t$  (exit). This approach implies that the entry and exit models are  
 14 estimated on subsets of the entire sample.<sup>11</sup> We suspect that exogenous factors may have different  
 15 influences on entry and exit decisions and thus our specification allows for such differences.

16 Several control variables that are conceptually relevant to the insurance decisions are considered.  
 17 We include the entrepreneur's main characteristics (I), such as age, sex and level of education, and  
 18 structural variables (II) related to the farms' location, organization and farming systems<sup>12</sup>. We also consider  
 19 financial factors (III) reflected in a farm's capital, financial leverage, and other relevant financial variables.  
 20 Finally, we include two variables related to parameters of the insurance programs (IV) and two alternative  
 21 risk management strategies<sup>13</sup> (V).

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<sup>11</sup> As suggested by a referee, given our 4 year panel data and because entry/exit equations are independent it seems more appropriate to stress once more that we are modeling entry and exit decisions, whereas complete turnover (i.e. entry-exit-entry) is not directly modeled in our framework.

<sup>12</sup> An anonymous referee has noted that insurance decisions may be affected by crop rotation choices. We essentially assume that crop choice decisions are pre-determined relative to the insurance decision in that modeling the endogeneity of crop choice decisions would require a different econometric strategy and is beyond the scope of the present analysis. The interested reader may refer to Lacroix and Thomas (2011) and Carpentier and Letort (2012) for recent applications considering these issues.

<sup>13</sup> Understanding how farmers cope with risks by adopting alternative strategies such as irrigation and crop diversification is an important issue. Recent studies suggest that farmers, on average, are risk averse and adopt

1 TABLE 3 ABOUT HERE

2 "Farms' capital" is the sum of farm assets net of current liabilities (net worth). "Financial leverage"  
3 is the farm's debt-to-equity ratio which is defined as the ratio of total farm liabilities over equity (i.e. owned  
4 capital). "Expected premia" is computed by averaging within regions and farming systems the (crop-  
5 specific) total premia. The variable "Expected loss ratio" is the ratio of total indemnities per hectare over  
6 premia per hectare: the expected loss ratio is the average of the farm-specific loss ratio across region and  
7 farming system. Due to data limitations, the loss ratio considers all indemnities, regardless of the type of  
8 insurance contract. Aggregated data provide a better representation of expected returns per dollar paid in  
9 premium since indemnity payments are highly variable in any single year at the farm level. An aggregate  
10 premium provides a valid representation of expected premia for all farms of a given type and in a specific  
11 region. As noted, farms are quite heterogeneous. For example, large variation is observed in farms' capital,  
12 cultivated areas and numbers of crops across the sample (Table 3).

13 In order to investigate decisions of exit or entry with respect to changes in selected strategies  
14 (cultivated area, irrigated area, and crop diversification) we have included variables' in first differenced  
15 form. We include positive and negative changes in key variables in order to identify asymmetric effects on  
16 entry and exit decisions. In particular we have introduced the variables "Increase in cultivated area",  
17 "Increase in irrigated area", and "Increase in crop diversification", as well as the correspondent "Decrease"  
18 variables. The underlying assumption is that farmers that are experiencing land expansion (or contraction),  
19 increases (or decreases) in irrigated land, or changes in the number of cultivated crops face different  
20 situations that may influence their decisions on crop insurance<sup>14</sup>. In particular, changes in the structure of a  
21 farm operation may lead to changes in risk management strategies, which are represented by, among other  
22 things, entry or exit from the crop insurance program. We distinguish increases from decreases in key  
23 variable to allow for asymmetric responses to changes of opposite signs. Note that a symmetric response

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strategies to manage risk. Additional details on these risk management strategies are provided by Di Falco and Perrings (2005), Foudi and Erdlenbruch (2012), and Finger (2013).

<sup>14</sup> Finally we have introduced variables to capture substantial changes (in either direction) of cultivated area, irrigated area and crop diversification. Large changes are likely to influence entry and exit decisions. The results confirms previous findings and are provided in the appendix.

1 would be implied if the coefficients are of the same magnitude but of opposite sign. To the best of our  
2 knowledge, this approach is original in modeling exit and entry decisions in crop insurance markets.

3 TABLE 4 ABOUT HERE

4

### 5 **Empirical results**

6 We computed likelihood ratio tests to select which estimator (random effects or Mundlak's fixed  
7 effects) and set of variables would best fit our sample. For the participation model a FE specification is  
8 preferred, while the entry and exit models are best estimated using a RE estimator. Results for the model of  
9 insurance participation are shown in the first column of table 5, while the entry and exit model results are  
10 shown in the 2<sup>nd</sup> and 3<sup>rd</sup> columns of the table. Our models seem to fit well, with McFadden (1974) pseudo  
11 R-square values ranging from 0.36 to 0.54. The percentages of correct predictions are also satisfactory at  
12 about 65% for the entry and exit models and as high as 94% for the participation model.

13 TABLE 5 ABOUT HERE

14 Part of the heterogeneity in participation, entry and exit decisions is captured by geographical  
15 location. At least one of the dummy variables defined as "North West", "Centre", and "South" (with North  
16 East as the default omitted category) is statistically significant in all three models. Three factors may help  
17 explain important geographic differences. In the Northern regions there is a strong presence of producer  
18 organizations and cooperatives that have aggregated the demand for crop insurance. Such a phenomenon  
19 is largely absent in the South. Second, crop insurance in Italy has been established to help farmers cope  
20 with damages from hail, a hazard that is much more relevant in the North, where grapes and fruits are  
21 cultivated, than in the South. Third, the defense consortia, which aggregate the vast majority of the  
22 demand for insurance and process reimbursements to farmers when losses occur, are much more effective  
23 in the North than in the South.

24 Previous studies have found mixed results in terms of the effect of education on the adoption of  
25 risk management tools (Van de Ven and Van Praag, 1981; Mishra and El-Osta, 2002; Enjolras and Sentis,  
26 2011). On theoretical grounds, Shapiro and Brorsen (1988) suggest that farmers may become less risk  
27 averse as they gain education, with more educated farmers being less likely to adopt risk management

1 strategies such as crop insurance contracts, consistent with the human capital theories developed by Welch  
2 (1970) and Schultz (1972). We find that farmer education does not significantly affect participation (other  
3 than negatively at the highest level) or entry and exit decisions.

4 As for firm characteristics, “High altitude” and “Cultivated area” are significant in explaining  
5 participation (controlling for regions). Greater participation of farms located at higher altitudes reflects  
6 important risk differences that correspond to altitude, such as risks of hail, low temperatures, frosts, and  
7 excessive wind, which tend to increase with altitude (Mahoney et al., 2012). This result agrees with  
8 previous studies (Enjolras et al., 2012) that have also found a positive correlation between altitude and  
9 adoption of crop insurance.

10 Larger farms, either in terms of economic size or by total area under cultivation, are more likely to  
11 participate in crop insurance. These findings are consistent with the results of previous studies (Goodwin,  
12 1993; Smith and Goodwin, 1996; Enjolras and Sentis, 2011; Singerman *et al.*, 2012; Di Falco et al., 2014),  
13 suggesting that farmers' endowments are a key driver for crop insurance decisions (Harrington and  
14 Niehaus, 1999). The fixed costs associated with enrollment in insurance schemes may inhibit operators of  
15 small farms as well as insurance agents and companies that service these small farms which thereby can be  
16 expected to limit participation.

17 The “Expected Loss Ratio” is statistically significant in the participation and exit models. However,  
18 contrary to prior expectations, higher expected loss ratios correspond to a lower likelihood of participation  
19 and to a higher likelihood of exit. This may reflect the aggregated nature of loss-ratios and premiums and  
20 the large heterogeneity across regions, and thus may reflect other unobserved, aggregate factors<sup>15</sup>. To  
21 investigate this further, we included the interaction terms of “Expected Loss Ratio” and regional dummies.  
22 We find a negative correlation with the entry decision for Northeast and Northwest, and a positive  
23 correlation with the exit decision only for Northwest. In the North the “Expected Loss Ratio” is close to  
24 one. In other respects, the apparent inconsistency is limited only to the North and may be partially  
25 explained by the time lag (usually 1 year) occurring between the compensation and the assessment of the

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<sup>15</sup> Analyzing these issues is beyond the scope of the present analysis and data and is left as an important item for future research.

1 damage (Enjolras et al., 2012). In all other cases either the expected sign is either confirmed or the  
2 variables are not statistically significant. A limitation of our variable is that it is constructed only from  
3 insured farmers, for which we observe data on indemnities and premiums. As shown by Just et al. (1999),  
4 loss ratios may be significantly higher for insured than for non-insured farmers due to adverse selection.

5 We interacted the variable “Expected Premia” with dummies for geographical location (Northwest,  
6 Northeast, Center and South) in order to control for heterogeneity at the regional level: therefore the table  
7 shows four variables ( $E[\text{premia}] * \text{Northwest}$ ;  $E[\text{premia}] * \text{Northeast}$ ;  $E[\text{premia}] * \text{Center}$ ;  $E[\text{premia}] * \text{South}$ ). It  
8 is likely that risk is more homogeneous within macro-regions than between macro-regions. The approach is  
9 similar to that followed by Goodwin (1993). We find that the higher the expected premium, the lower the  
10 participation in Northwest and Center. Conversely, in “Northeast” the higher is the expected premium, the  
11 greater is participation in insurance programs. The results are not surprising considering that “Expected  
12 Premia” is lower in the Northeast, while participation is very significant. The combination of higher  
13 premiums and lower loss ratios in the “Northeast” suggest that exposure to systemic risk (Miranda and  
14 Glauber, 1997) may be an issue for this particular region, where apple and grape production is prevalent. In  
15 fact, the indemnities paid in the “Northeast” are three to eight times as large as in the rest of Italy (Table 2).  
16 As a result, the higher is the “Expected Premia” (which reflects a higher level of underlying risk), the higher  
17 is the participation in crop insurance program.

18 Coefficients of variables related to alternative strategies for risk management show that farmers  
19 who are more diversified or have irrigation are less likely to purchase insurance (although only at the 5%  
20 significance level). These results suggest that both diversification and irrigation can be substitute for  
21 insurance—a result that is consistent with Smith and Goodwin, 1996; Blank and McDonald, 1996; Di Falco  
22 and Chavas, 2009; Enjolras and Sentis, 2011; and Di Falco et al., 2014. The negative signs for “Crop  
23 diversification” may also reflect a form of moral hazard (Smith and Goodwin, 1996), where insured farmers  
24 do not use alternative risk-coping strategies, such as crop diversification. In addition, farmers that decrease  
25 diversification tend to enter the insurance market, possibly to manage the risks associated with the new

1 activities<sup>16</sup>.

2 The results on the determinants of farmers' insurance market entry and exit decisions (2<sup>nd</sup> and 3<sup>rd</sup>  
3 columns, table 5) merit additional discussion. First, the results are not always symmetric, with the  
4 determinants of the entry decision often differing substantially from those of the exit decision. As  
5 expected, the entry decision model corresponds rather closely with the participation model, showing the  
6 same regional pattern as the participation model. The larger the decrease in cultivated area or irrigated  
7 area, the lower is the probability of adopting insurance for farms that are uninsured.

8 "Irrigation" is statistically significant in the participation model, and is not statistically significant for  
9 entry and exit decisions. Participation varies in a positive manner with irrigation, suggesting that increases  
10 in irrigation tend to be associated with a higher probability of participation in insurance programs. Foudi  
11 and Erdlenbruch (2012) found that irrigation technology serves as self-insurance in that buying "insurance  
12 decreases the probability of adopting irrigation", p.454. The coefficient for "Decrease in irrigated area" is  
13 statistically significant, and negative, in the model of the entry decision, indicating that uninsured farms (at  
14 t-1) that reduce their irrigation systems (in time t-1) are less likely to sign up for insurance (in time t).

15 Crop diversification is significantly inversely correlated with participation, as expected, and not  
16 significant for entry. However, "Decrease in crop diversification" is statistically significant and positive for  
17 entry. This suggests that uninsured farms that reduce the number of cultivated crops are more likely to  
18 sign up for insurance. Following Cabas et al., 2008 and Bezabih and Sarr, 2012, we may assume that risk  
19 aversion and crop diversification are correlated, such that risk-averse farmers, as well as non-specialized  
20 farms, are less likely to be insured. However, farmers that decrease the number of cultivated crops tend to  
21 enter into a crop insurance contract.

22 High values of the expected loss ratio appear to favor entry decisions. Specifically, higher values of  
23 the "Expected Loss Ratio" are likely to correspond to farms producing riskier crops in riskier regions,  
24 favoring insurance decisions. It is also apparent that higher loss ratios correspond to higher returns to  
25 insurance, a factor that also tends to favor entry into the insurance program.

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<sup>16</sup> Farmers that increase diversification are more likely to exit. This result is statistically significant in one specification at the 10% level. Farmers that decrease the irrigated area are less likely to enter in the insurance market.

1 It is important to evaluate the role of insurance premia on dynamics in insurance markets. In  
2 “Northwest”, “Center” and “South” high insurance premia tend to lower the probability of entry by  
3 uninsured farmers, and to increase the probability of exit by insured farmers<sup>17</sup>. Though not always  
4 statistically significant, the results tend to suggest that the demand for insurance is downward sloping with  
5 respect to premia, a result consistent with adversely selected participants in crop insurance programs in  
6 that the larger the premium the lower the attractiveness of the contract.

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## 8 **Concluding Remarks**

9 We consider three important aspects of the decision to insure crops in Italy. These include the  
10 question of participation, and also the decisions to enter and to exit from an insurance scheme. The  
11 decisions are related but individual models of entry and exit provide additional information about factors  
12 affecting participation in crop insurance. Beyond understanding participation patterns, policymakers have a  
13 keen interest in understanding the dynamics of insurance participation. This interest is substantiated by the  
14 significant investment of public funds to support such schemes and the oft-repeated goal of increasing  
15 participation. We investigate these dynamics by focusing on the entry decision of uninsured farmers and  
16 the exit decision of insured farmers. We use a dynamic specification that considers how changes in  
17 cultivated area, irrigation and crop diversification are related to entry and exit decisions. We allow these  
18 changes to be asymmetric in that increases do not necessarily correspond to an opposite adjustment for  
19 decreases in the same variable. We find that farm and market characteristics have different impacts on  
20 these individual aspects of insurance demand, and that increasing or decreasing the cultivated and irrigated  
21 area, and the crop diversification tend to have different effects on insurance uptake.

22 Although subsidized crop insurance programs continue to proliferate around the world,  
23 participation remains sporadic and not well understood in many cases. If policymakers intend to use  
24 subsidized crop insurance as an important mechanism for agricultural risk management, they are likely to  
25 be concerned with the factors that lead a farmer to adopt insurance and to remain insured. To the extent

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<sup>17</sup>We discuss the signs of the coefficients, but it is worth noting that in many cases the coefficients are statistically not significant: for example the variables “E[premia]\*Northeast”, “E[premia]\*Center”, and “E[premia]\*South” are statistically not significant in the entry decision equation (column 2, table 5).



1 that farm and operator characteristics differ across those farmers that enter and exit crop insurance  
2 schemes, policies intended to support participation may take different approaches for farmers that are  
3 already insured than for farmers that do not currently insure. Targeted technical support is usually  
4 recommended to enhance the participation in agricultural insurance programs (Mahul and Stutley, 2010).  
5 For example, education and outreach programs may adopt different approaches toward encouraging  
6 insured farmers to maintain coverage than what might be optimal in encouraging uninsured farmers to  
7 enroll in insurance programs.

8 We find that education and farm size are determinants of participation in insurance markets,  
9 confirming the earlier findings of Enjolras and Sentis (2011), Finger and Lehmann (2012), and Singerman *et*  
10 *al.*, 2012, among others. Our analysis also explicitly models the entry and exit decisions at the farm level.  
11 We find that entry and exit decision are driven by different factors and that adjustments to changes may  
12 reflect asymmetric patterns of adjustment, with increases in key variables implying different adjustments  
13 that would be the case for corresponding decreases. Our results are consistent with previous studies in this  
14 regard (see, for example, Smith and Goodwin, 1996; and Singerman *et al.*, 2012). The negative correlation  
15 that we found for crop diversification (and irrigated area) in the participation equation suggests that  
16 farmers tend to adopt crop diversification (and irrigation) and insurance contracts as alternate risk  
17 management strategies. These factors are certainly alternative mechanisms for managing risk and thus  
18 would be expected to serve as substitutes for insurance participation.

19 A few caveats are relevant to this study. First, our data were collected over a four year period. This  
20 reflects that fact that our focus on entry and exit decisions required observing individual farms over  
21 multiple periods. We thus included only those farms continuously observed during the period. Although  
22 we rely on a large set of data made up of more than three-thousand farms, our results do not capture more  
23 recent developments in the continually-changing insurance program and markets. Another drawback of our  
24 study is that detailed, farm-level data on crop insurance in Italy (such as characteristics of individual  
25 insurance contracts, realized losses, etc.) are largely unavailable. Even if such data were collected by  
26 surveys, we may not be able to observe the parameters associated with insurance offerings to uninsured  
27 farmers. To the extent that promotion of participation in insurance programs is a key objective of the

1 European Commission Agenda, empirical work on the dynamics and turnover in insurance markets  
2 represents a promising and fruitful area for additional future research.

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1 **Tables**2 **Table 1 - Crop insurance market in Italy (2004-2012)**

			2004	2005	2006	2007	2008	2009	2010	2011	2012
(a)	Certificates	.000	212	212	211	237	265	226	208	208	214
(b)	Insured land	.000 ha	982	1074	1125	1051	1450	1355	1153	1164	na
(c)	Insured value	M €	3.710	3.810	3.789	4.380	5.436	5.131	5.313	6.145	6.826
(d)	Total premia	M €	177	269	265	293	338	317	285	287	321
(e)	Indemnities	M €	152	159	149	184	272	234	169	171	231
	Public contribution *	%	56.8	65.9	66.6	66.8	66.3	67.0	66.4	66.1	na
(c/a)	Average certificate value	.000 €	17.5	18.0	18.0	18.5	20.5	22.7	25.5	29.5	31.9
(e/d)	Loss ratio		0.66	0.59	0.55	0.64	0.81	0.75	0.60	0.58	0.72
	Monorisk policies (%) <sup>†</sup>		92.0	na	77.4	na	53.7	na	50.2	na	na
	Pluririsks policies (%) <sup>†</sup>		7.7	na	19.6	na	40.0	na	46.6	na	na
	Multirisks policies (%) <sup>†</sup>		0.3	na	2.9	na	6.3	na	3.3	na	na

3 (\*):premiums/insured value. *na* indicates not available.

4 (†) The statistic refers to the percentage of the total insured value.

5 Source: Our elaboration on data from the Istituto di Servizi per il mercato agricolo alimentare (Ismea)

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1 **Table 2 - Geographical distribution of crop insurance contracts in Italy (2011)**

	Units	North East	North West	Middle	South	Italy
(a) Certificates	k	108	68	15	33	2824
(b) Insured Value	M €	2.396	1.486	419	754	1656
(c) Total premia	M €	178	69	20	40	471
(d) Indemnities	M€	142	33	17	38	865
(e) Percent of agriculture gross value	%	34.0	29.5	19.5	17.0	100.0
(c/a) Average certificate value	k €	22.19	21.85	27.93	22.85	22.56
(d/c) Loss ratio		0.80	0.48	0.85	0.95	0.75

2 K and M indicate thousand and million.

3 Source: Ismea data

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10 **Table 3 – Definition of variables and descriptive statistics**

Variable	Description	Mean	Std. Dev.
<b>(I) Entrepreneur's characteristics</b>			
Age	Entrepreneurs' age	54.6	13.6
Sex	One for male entrepreneurs	93.7%	
Education level [1]	One for middle school, zero otherwise	29.3%	
Education level [2]	One for high school degree, zero otherwise	52.6%	
Education level [3]	One for bachelor degree, zero otherwise	11.9%	
Education level [4]	One for post-graduate, zero otherwise	1.8%	
<b>(II) Structural variables</b>			
Organic farms	One if organic firm, zero otherwise	2.8%	
Crop and livestock farms	One if producing both livestock and crops	43.8%	
High altitude	One if located 600 meters above sea level	20.7%	
Less favored areas	One if located in disadvantage areas	0.39%	
Corporations	One for corporations	0.47%	
Land size	Cultivated hectares (ha)	32.4	66.3
<b>(III) Financial determinants</b>			
Farms' capital	Millions of Euros in real terms	0.13	0.40
Financial leverage	Liabilities / equity	0.60	3.90
Return on equity	Net income / equity	0.08	0.14
Crop revenue	Thousands of Euros in real terms	42.1	186.4
Crop revenue per hectare	Thousands of Euros per hectare in real terms	4.47	58.7
<b>(IV) Insurance markets</b>			
E[Premia]	Average premia (k €/ha), by regions and crops	0.11	0.09
E[LossRatio]	Average loss ratios, by regions and crops	0.76	1.59
<b>(V) Alternative risk management strategies</b>			
Irrigation	Irrigated hectares (ha)	2.6	8.1
Crop diversification	Number of cultivated crops in one year	2.92	1.75

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**Table 4 – Summary statistics by category of participant**

	Units	Participant	Not participant	Entry	Exit
Observations	Num.	3000	22358	14773	2152
Age	Years	53.1	54.8	55.0	53.4
Sex - male	%	94.7	93.6	93.6	94.7
Education level [1]	%	29.4	29.3	29.3	29.1
Education level [2]	%	53.0	52.5	52.5	53.5
Education level [3]	%	11.4	11.9	11.9	11.4
Education level [4]	%	0.9	1.9	1.9	0.8
Organic farms	%	2.8	2.8	2.8	2.6
Crop and livestock farms	%	44.1	43.8	43.8	44.2
High altitude	%	28.0	19.8	19.6	28.6
Less favored areas	%	47.1	38.3	38.3	47.3
Corporations	%	51.5	46.5	46.5	51.6
Farms' capital	M €	0.33	0.10	0.10	0.34
Financial leverage		0.57	0.60	0.60	0.63
Return on equity		0.07	0.08	0.08	0.07
Crop revenue	k €	116.3	29.5	88.1	91.8
Crop revenue per hectare	k €	2.4	4.0	2.6	2.4
Land size	ha	5.76	2.90	2.89	6.06
Irrigation	%	0.34	0.25	0.25	0.35
Crop diversification	Num. crops	2.81	2.94	2.94	2.85
E[Premia]	k €/ha	0.12	0.11	0.11	0.12
E[LossRatio]		0.46	0.80	0.81	0.46
North West	%	37.5	34.0	34.1	37.0
North East	%	34.4	13.2	13.0	35.2
Centre	%	9.2	15.5	15.4	9.7
South	%	18.9	37.4	37.5	18.0

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1 **Table 5 – Adoption, entry and exit decisions in crop insurance market**

	Participation (FE omitted)	Entry (RE omitted)	Exit (RE omitted)
Age	-0.003 (4.06)**	0.006 (0.30)	0.041 (1.17)
Age <sup>2</sup>		-0.000 (0.16)	-0.000 (1.03)
Sex	0.127 (2.53)*	-0.070 (0.51)	0.301 (0.91)
Education level [1]	-0.092 (1.60)	0.101 (0.52)	-0.092 (0.34)
Education level [2]	-0.088 (1.59)	0.056 (0.29)	-0.190 (0.72)
Education level [3]	-0.062 (0.98)	-0.068 (0.31)	-0.173 (0.57)
Education level [4]	-0.388 (3.39)**	-0.217 (0.54)	0.222 (0.31)
Organic firms	0.082 (1.20)	0.012 (0.05)	-0.351 (0.85)
Crop and livestock farms	0.053 (1.99)*	-0.043 (0.51)	-0.011 (0.07)
High altitude	0.093 (2.69)**	0.159 (1.51)	0.069 (0.39)
Less favored areas	0.076 (2.74)**	-0.033 (0.38)	-0.156 (1.04)
Corporations	0.081 (0.53)	0.091 (1.25)	-0.023 (0.19)
Cultivated area(10 ha)	0.017 (10.23)**	0.001 (0.09)	0.012 (1.77)+
Increase in cultivated area		-0.029 (1.36)	-0.047 (0.77)
Decrease in cultivated area		-0.045 (3.84)**	0.117 (0.99)
Farms' capital	0.260 (10.31)**	0.038 (0.35)	0.001 (0.01)
Financial leverage	0.001 (0.02)	-0.008 (0.82)	0.012 (1.10)
Return on equity	0.165 (2.05)*	-0.286 (0.73)	0.513 (1.20)
Crop revenue	0.197 (3.51)**	0.029 (0.14)	0.075 (0.48)
Crop revenue per hectare	-4.987 (3.60)**	-0.234 (0.13)	0.597 (0.05)
Irrigated area	-0.028 (2.16)*	0.004 (0.09)	0.048 (0.72)
Crop diversification	-0.016 (2.22)*	0.010 (0.46)	0.024 (0.68)
Increase in irrigated area		0.230 (0.67)	1.307 (1.06)
Decrease in irrigated area		-0.853 (3.31)**	
Increase in crop diversification		0.016 (0.17)	0.310 (1.66)+
Decrease in crop diversification		0.354 (1.80)+	0.105 (0.40)
E[LossRatio]	-0.034 (2.89)**	-0.035 (1.09)	0.335 (2.33)*
E[premia]*NorthWest	-5.210 (10.99)**	-2.952 (2.49)*	0.383 (0.10)
E[premia]*NorthEast	4.650 (15.54)**	0.781 (0.76)	-1.865 (1.28)
E[premia]*Centre	-5.999	-0.585	3.190

E[premia]*South	(7.85)** -0.244 (0.90)	(0.24) -1.443 (1.35)	(0.62) 3.277 (1.33)
NorthWest	0.878 (12.03)**	0.415 (1.87)+	-0.022 (0.05)
South	0.132 (1.93)+	-0.144 (0.66)	-1.142 (2.54)*
Centre	0.802 (8.47)**	-0.210 (0.62)	-0.281 (0.59)
Time dummy: 2004	0.036 (1.12)		
Time dummy: 2005	-0.032 (0.98)		
Time dummy: 2006	0.018 (0.56)		
Constant	-1.576 (13.96)**	-2.653 (4.42)**	-1.596 (1.82)+
Correct predictions	94.4%	66.82%	64.38%
McFadden Pseudo R <sup>2</sup>	0.47	0.37	0.55
Observations	22,415	14116	1,517

+  $p < 0.1$ ; \*  $p < 0.05$ ; \*\*  $p < 0.01$

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1 **Online Appendix – Not intended for publication**

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3 We have compared two estimators: RE and the FE (as per Mundlak (1978), cfr. Greene, 2008, p.210). We  
4 added, as additional explanatory variables, the within-group means of the time-varying covariates ( $\bar{x}'_{it}$ ) to  
5 capture the correlation between the unobserved heterogeneity and the covariates. This specification makes  
6 the random effects model consistent:  
7

$$\begin{aligned} y_{it} &= x'_{it} \beta + c_i + \varepsilon_{it} \\ &= x'_{it} \beta + \bar{x}'_{it} \gamma + (c_i - E[c_i|X_i]) + \varepsilon_{it} \\ &= x'_{it} \beta + \bar{x}'_{it} \gamma + \varepsilon_{it} \end{aligned}$$

8 where  $E[c_i|X_i] = \bar{x}'_{it} \gamma$

9 The above specification is then estimated as a random effects probit model. The likelihood ratio tests for  
10 the FE model and the (nested) RE model are, respectively, 9910.5, 0.0412 and 17.8 for the adoption, entry,  
11 and exit models. The results support a FE estimator for the adoption model, and a RE for the entry and exit  
12 models.  
13

14 For the participation model, we report several estimates to show how we have proceeded to select the  
15 final model and the appropriate estimator. We included fixed time effects, eliminated the delta variables,  
16 and dropped the “Vegetable revenue” variable (to assess if the potential collinearity with “Cultivated area”  
17 is an issue. We report the selected model in the last column of table A.  
18

19 For the entry and exit models, the RE estimator is preferred. The estimates are not significantly different  
20 from the FE model and the likelihood ratio tests do not favor the FE estimator. In addition the FE estimates  
21 are not difficult to implement and interpret for the entry and exit models that are estimated on subsamples  
22 of the entire dataset. Specifically, we cannot include the time dummies for four years because of the  
23 peculiar nature of the entry and exit subsamples. The entry subset excludes observations for which the  
24 dependent variable (Insurance) is 1 at time t and at time t-1 and observations for which the dependent  
25 variable is 1 at time t and 0 at time t-1. The exit subset excludes observations for which the dependent  
26 variable (Insurance) is 0 at time t and at time t-1 and observations for which the dependent variable is 0 at  
27 time t and 1 at time t-1. In this light, we have maintained the previous specification of the entry and exit  
28 models.  
29

30 **References**

- 31 Greene, W. H. (2008). *Econometric analysis*. Pearson Education India.  
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Table A – Participation model

	RE	RE	RE	RE	FE a là Mundlak
Age	-0.007 (1.98)*	-0.006 (1.71)+	-0.005 (1.99)*	-0.005 (2.02)*	-0.003 (4.06)**
Sex	0.224 (0.99)	0.234 (1.02)	0.182 (1.01)	0.183 (1.02)	0.127 (2.53)*
Educ1	-0.338 (1.25)	-0.337 (1.24)	-0.304 (1.46)	-0.300 (1.44)	-0.092 (1.60)
Educ2	-0.239 (0.91)	-0.240 (0.92)	-0.318 (1.57)	-0.314 (1.55)	-0.088 (1.59)
Educ3	-0.202 (0.69)	-0.203 (0.69)	-0.329 (1.43)	-0.328 (1.43)	-0.062 (0.98)
Educ4	-1.033 (1.95)+	-1.031 (1.95)+	-0.919 (2.22)*	-0.921 (2.23)*	-0.388 (3.39)**
Organic:0/1	0.135 (0.43)	0.131 (0.41)	0.250 (1.00)	0.256 (1.02)	0.082 (1.20)
Veg=0/Veg+Anim=1	0.122 (1.00)	0.127 (1.03)	0.173 (1.82)+	0.187 (1.96)*	0.053 (1.99)*
Altitude:High=1	0.311 (1.98)*	0.301 (1.90)+	0.246 (1.90)+	0.244 (1.89)+	0.093 (2.69)**
Less fav. areas:0/1	0.169 (1.32)	0.175 (1.36)	0.194 (1.87)+	0.195 (1.89)+	0.076 (2.74)**
Organization:Group=1	0.163 (1.51)	0.164 (1.52)	0.221 (2.59)**	0.221 (2.59)**	0.081 (3.50)**
Capital Employed(mln€)	1.331 (5.83)**	1.352 (7.58)**	0.608 (6.69)**	0.628 (6.95)**	0.260 (10.31)**
Vegetal revenue(mln€)	0.113 (0.62)	0.133 (0.81)	0.231 (1.98)*		0.197 (3.51)**
Vegetal revenue(mln€)/ha	-11.650 (2.05)*	-11.743 (2.05)*	-2.731 (0.97)	-2.056 (0.79)	-4.987 (3.60)**
Financial leverage(%)	0.005 (0.44)	0.005 (0.47)	0.001 (0.08)	0.000 (0.02)	0.000 (0.02)
ROE	0.462 (1.88)+	0.487 (1.98)*	0.182 (1.03)	0.187 (1.05)	0.165 (2.05)*
Cultivated area(10 ha)	0.060 (5.98)**	0.064 (4.41)**	0.039 (6.67)**	0.040 (6.89)**	0.017 (10.23)**
Irrigated area(10 ha)	-0.093 (1.44)	-0.100 (1.53)	0.009 (0.19)	0.009 (0.20)	-0.028 (2.16)*
Different crops	-0.068 (1.96)*	-0.069 (2.03)*	-0.046 (1.84)+	-0.046 (1.83)+	-0.016 (2.22)*
E_primepaNorthWest	-21.452 (9.72)**	-21.192 (9.51)**	-16.379 (9.49)**	-16.460 (9.56)**	-5.210 (10.99)**
E_primepaNorthEast	18.238 (12.14)**	18.665 (12.47)**	12.936 (9.71)**	12.905 (9.68)**	4.650 (15.54)**
E_primepaCentre	-11.316 (3.10)**	-10.814 (2.86)**	-12.699 (4.05)**	-12.731 (4.08)**	-5.999 (7.85)**
E_primepaSouth	-0.055 (0.05)	-0.043 (0.04)	-0.272 (0.28)	-0.248 (0.26)	-0.244 (0.90)
E[LossRatio]	-0.054 (0.98)	-0.034 (0.61)	-0.067 (1.12)	-0.066 (1.16)	-0.034 (2.89)**
NorthWest	3.561 (10.14)**	3.607 (10.31)**	2.799 (9.83)**	2.797 (9.83)**	0.878 (12.03)**
South	1.025 (3.23)**	1.074 (3.39)**	0.432 (1.63)	0.416 (1.57)	0.132 (1.93)+
Centre	2.482 (5.32)**	2.486 (5.27)**	1.962 (5.21)**	1.952 (5.19)**	0.802 (8.47)**
Delta-Different crops	0.051 (0.79)	0.051 (0.80)			
d2005		0.183 (2.67)**	0.118 (2.10)*	0.125 (2.23)*	0.036 (1.12)
d2006		-0.072 (1.05)	-0.052 (0.92)	-0.044 (0.77)	-0.032 (0.98)
d2004			0.083 (1.50)	0.089 (1.61)	0.018 (0.56)
Constant	2.942 (77.09)**	2.902 (74.92)**	1.900 (42.47)**	1.902 (42.78)**	-1.576 (13.96)**
Observations	22,415	22,415	22,415	22,415	22,415

Table B – Entry and exit models

RE vs FE specification	Entry model		Exit model	
	RE	FE a là Mundlak	RE	FE a là Mundlak
Age	-0.001 (0.53)	-0.001 (0.53)	0.002 (0.20)	-0.001 (0.19)
Sex	-0.150 (1.53)	-0.149 (1.52)	0.354 (0.58)	0.260 (0.64)
Educ1	-0.045 (0.34)	-0.045 (0.34)	0.176 (0.33)	0.188 (0.54)
Educ2	-0.076 (0.60)	-0.076 (0.59)	-0.035 (0.07)	0.051 (0.15)
Educ3	-0.106 (0.72)	-0.106 (0.72)	0.333 (0.57)	0.340 (0.91)
Educ4	-0.306 (1.12)	-0.306 (1.12)	-5.556 (0.00)	
Organic:0/1	0.060 (0.36)	0.060 (0.36)	-8.461 (0.00)	
Veg=0/Veg+Anim=1	0.041 (0.65)	0.043 (0.69)	-0.033 (0.11)	-0.057 (0.30)
Altitude:High=1	-0.032 (0.36)	-0.032 (0.36)	-0.091 (0.26)	-0.138 (0.63)
Less fav. areas:0/1	0.004 (0.05)	0.004 (0.06)	0.104 (0.38)	0.048 (0.28)
Organization:Group=1	0.114 (2.10)*	0.114 (2.10)*	-0.071 (0.32)	-0.054 (0.38)
Capital Employed(mln€)	-0.055 (0.44)	-0.050 (0.41)	0.243 (1.76)+	0.158 (1.93)+
Vegetal revenue(mln€)	0.042 (0.21)		0.152 (0.52)	0.104 (0.51)
Vegetal revenue(mln€)/ha	0.586 (0.56)	0.607 (0.58)	-30.112 (0.78)	-13.554 (0.59)
Financial leverage(%)	-0.008 (1.43)	-0.008 (1.44)	0.020 (0.51)	0.016 (0.62)
ROE	-0.412 (1.36)	-0.406 (1.35)	0.350 (0.50)	0.512 (1.09)
Cultivated area(10 ha)	0.001 (0.12)	0.001 (0.17)	0.008 (0.59)	0.008 (0.95)
Irrigated area(10 ha)	-0.001 (0.03)	-0.001 (0.04)	0.082 (0.72)	0.053 (0.81)
Different crops	0.034 (2.28)*	0.034 (2.28)*	-0.023 (0.34)	-0.039 (0.88)
E_primepaNorthWest	-0.806 (1.42)	-0.810 (1.42)	3.926 (0.58)	3.228 (0.81)
E_primepaNorthEast	0.250 (0.29)	0.244 (0.28)	-1.915 (0.69)	-1.472 (0.81)
E_primepaCentre	-3.870 (2.08)*	-3.869 (2.08)*	6.933 (0.90)	4.743 (0.98)
E_primepaSouth	-0.930 (1.25)	-0.927 (1.25)	0.862 (0.18)	0.624 (0.21)
E[LossRatio]	-0.001 (0.07)	-0.001 (0.07)	-0.196 (0.53)	-0.097 (0.41)
NorthWest	0.108 (0.64)	0.107 (0.63)	-0.409 (0.50)	-0.276 (0.55)
South	-0.260 (1.53)	-0.262 (1.55)	-0.406 (0.47)	-0.368 (0.64)
Centre	0.148 (0.62)	0.145 (0.61)	-0.937 (0.99)	-0.687 (1.11)
Constant	-2.035 (7.89)**	-2.035 (7.89)**	-2.794 (2.28)*	-1.976 (2.56)*
Observation	14,773	14,773	1,516	1,516

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Table C – Estimates from alternative specifications for the entry and exit models

	Entry model		Exit model	
	1	2	1	2
Increase in cultivated area	-0.030 (1.41)		-0.050 (0.84)	
Decrease in cultivated area	-0.048 (4.12)**		0.104 (0.89)	
Increase in irrigated area	0.290 (0.87)		0.938 (0.87)	
Decrease in irrigated area	-0.911 (3.51)**			
Increase in crop diversification	0.028 (0.29)		0.287 (1.56)	
Decrease in crop diversification	0.333 (1.73)+		0.141 (0.54)	
E[Premia]	-1.167 (1.93)+	-1.277 (2.09)*	-1.343 (1.05)	-1.350 (1.09)
E[Loss Ratio]*NorthWest	-0.061 (1.65)+	-0.043 (1.35)	0.877 (3.93)**	0.903 (4.29)**
E[Loss Ratio]*NorthEast	-0.701 (3.05)**	-0.696 (3.04)**	0.203 (0.63)	0.196 (0.63)
E[Loss Ratio]*Centre	-0.017 (0.04)	-0.066 (0.15)	0.090 (0.13)	0.110 (0.17)
E[Loss Ratio]*South	0.148 (1.79)+	0.141 (1.71)+	-0.157 (0.89)	-0.162 (0.95)
Large change in cultivated area		-0.042 (3.64)**		0.001 (0.02)
Large change in irrigated area		0.264 (0.79)		0.245 (0.22)
Large change in crop diversification		0.103 (1.24)		0.486 (2.24)*

1 We control for all factors included in previous specifications (age, education, organic farm, ROE, size, geographic location, etc.).

2