Adoption of Soiless Cropping Systems in Mediterranean Greenhouses: An Application of Duration Analysis

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Abstract. The Mediterranean basin holds the greatest area of horticultural crops grown under plastic in the world. Spain houses almost 50,000 ha of greenhouses and is one of the largest suppliers of European Union markets. Soiless cropping systems constitute an efficient production practice used in greenhouses and lessen the limitations arising from the ban of methyl bromide. Soiless cultivation is associated with the use of expensive technology, which makes farmers consider such adoption carefully. The aim of this work is to statistically identify the factors that significantly influence the timing of adopting soiless cultivation by using duration analysis. The study focuses on Mediterranean greenhouses, specifically horticultural producers in the southeast of Spain. The results suggest that certain individual characteristics as well as aspects of a social nature exert a favorable influence by reducing the delay in adopting soiless cultivation. Furthermore, technical complexity, commercial aspects related to price, and the repercussion of policies regulating methyl bromide use are also discussed in this work.

Recent decades have witnessed important changes in international production of horticultural crops. Furthermore, food safety and quality demands must be added to the progressive increase in consumption. Year-round demand means production must take place outside the natural season, which is the reason why the surface area of greenhouse crops has increased. The Mediterranean region houses 135,000 ha of horticultural species grown under plastic. The surface of greenhouse-covered areas is ≈27,700 ha in Italy, 14,000 ha in Turkey, 10,000 ha in Morocco, and almost 50,000 ha in Spain (FAO, 2002). Spain is the main producer in Europe of fresh horticultural produce for export. Greenhouse cultivation began to spread in Spain in the 1970s, coinciding with the development of plastics for agricultural use. These greenhouses are mainly found in southeastern Spain in regions with a warm climate and high solar radiation, which first favored the generalized use of the “parral” type shelter (a kind of low-cost plastic greenhouse widely used in the Mediterranean basin with a flat covering and a wooden structure). In recent years, such installations have undergone changes in their structure and have notably increased their technological levels (García-Martínez et al., 2008).

The most important horticultural crops grown in Spanish greenhouses are tomatoes and peppers for fresh consumption. These crops tend to be grown as monocultures and have achieved high levels of specialization in some coastal regions of the Spanish Mediterranean as is the case of the tomato in the Mazarrón region (Murcia) or the pepper in Campo de Cartagena (Murcia and Alicante), whereas in Almería, there is greater crop diversification. In fact, Almería is considered to be the region with the largest concentration of greenhouse crops in the world, the surface area of the greenhouses has increased gradually, reaching 38,780 ha (MAPA, 2006).

There are a number of advantages to greenhouse production, an outstanding one being that it enables harvesting outside the natural season corresponding to the climate of each region with the profit related to price seasonality together with a notable increase in yields and final production quality. This intensive farming system increases the efficacy of the main disbursements.

Despite all these advantages, the greenhouses also have associated drawbacks, among which the most outstanding are the high investment needed to install the different components and training needs for proper management of the installations.

Demands for quality, health, and environment are also issues guaranteed by the European normative in environmental matters with this normative becoming increasingly far reaching and strict. Crop intensification and repetition lead to soil degradation and increased resistance of pathogens like Verticillium, Phytophthora, or nematodes, making soil disinfection essential. In recent decades, the most common method to use soil disinfectants for soils vegetable growing has been the use of methyl bromide (MB). Nevertheless, the risks it poses to the environment have generated great deliberation and, after long debate, the Montreal Protocol was signed in 1987 with the decision to ban general use after 2005. An alternative to MB is soiless cropping systems, which appear to be a safer and more decisive alternative for the producer (Engindeniz, 2004).

Currently, soil cultivation continues to dominate; however, soiless cultivation can be considered one of the most important technological breakthroughs to have taken place in greenhouse cultivation, facilitating control of almost the whole growing cycle. Soiless cultivation systems already occupy over 20% of the total surface area of greenhouses.

There are different soiless cropping models, all of which require more or less complex technology. This system has a series of advantages and drawbacks when compared with traditional production. Among the drawbacks, the most outstanding advantages are higher efficiency in the use of production factors, above all irrigation water, fertilizers, and other chemical and plant protection products (Van Os, 1999). In closed systems, the filtered waste can be collected and disinfected, avoiding environmental pollution (Garibaldi et al., 2003). Furthermore, both quality and production of yields are increased if considered in terms of the commercial parameters for caliber, shape, uniform size, and shine. Moreover, sales planning is improved given increased regularity of production and an extended offer period. Among the drawbacks of soiless cultivation, the most important is the strong investment required given that greenhouses must be more automated, especially in fertilization. Regarding climatic control, if one wishes to protect the crop from the sharp drops in temperature and extend the production calendar, it may be necessary to install some type of heating system. Such equipment also...
adoption depends from a process of learning and experience, in which the individual concerned acquire, integrate, and evaluate the information that enables them to make the best decisions about technology. Thus, knowledge acquired during this process is unique to each individual and is based on a mixture of scientific information, personal experience, and cultural influences. Thus, the decision to adopt will be influenced by (Pannell et al., 2006):

A. Farmers’ individual characteristics, which include social, personal, and cultural aspects as well as environmental, demographic, and economic factors; and

B. The characteristics inherent to the technology. The potential users should perceive an added advantage over the traditional technology they currently use.

Considering the different variables contained within the groups of motives that can encourage innovation to be adopted, the decision to adopt soilless technology in greenhouses will be more likely along the time when the subjective use that the farmer derives ($U_j$) is greater than the use derived from cultivation in soil, which is the traditional production practice ($U_i$). The change will take place when $U_j > U_i$. For an individual farmer, the probability of changing technology is:

$$P_j = f(A, B)$$

where $P_j$ is the probability of adopting soilless practices; $A$ is the vector of cross-section and time-dependent variables describing the social, cultural, and personal influences on the adoption decision; and $B$ is the vector of cross-section and time-dependent variables describing the attributes of technology that affect the adoption innovations.

Duration analysis. DA is a statistical method that considers the expected time an individual spends in one state before transition to another, studying the difference in the time, $T$, between the two alternative and exclusive states (Lancaster, 1990). In the context of technological adoption, this transition lasts from the moment at which the technology is known about until its adoption is made effective.

It is constructed as a behavioral model, in which the individual choices are modeled using cross-section data and incorporates dynamic elements to the innovation adoption. Early works applying DA to the social sciences focused on factors affecting employment periods (Lancaster, 1972). Recently, this method has been applied to study other fields such as agriculture (Burton et al., 2003) and fishing (Smith, 2004).

In the present study into adoption of soilless technology in greenhouse production, the starting date has been set as the year in which the technology appeared on the Spanish market (1986) or the year in which the farmer began this agricultural activity in the event this was after the aforementioned year. The exit date, or end of the spell, will be the year the farmer adopted this innovation. In specific cases, there are farmers that had not adopted the technology at the date of analysis, and the end of the time spell is thus unknown. Although adoption could take place in the future, for these cases, the statistical procedure is right-censored establishing the year when data were collected (2006) as the end date of the observation period.

In each case, the objective is to analyze the sign and magnitude of the effects exerted by the explanatory variables on the length of the spell considering a homogeneous population in terms of the systematic factors affecting the random variable, $T$ (Kiefer, 1988; Lancaster, 1990). Probability theory plays a fundamental role in DA, because one can consider the probability of the end of the range instead of its length. The probability that a farmer will adopt a particular technology in period $t + dt$, given that he or she had not adopted until $t$, is defined by the hazard function.

Let $f(t)$ be the density function of the continuous probability of the random variable, $T$, where $t$ is the end of the range. Its corresponding cumulative function will be defined by:

$$F(t) = \int_0^t f(s)ds = Pr(T \leq t)$$

Likewise, the distribution of $T$ can be expressed by the survival function, $S(t)$, the reverse of the cumulative distribution function of $T$. This defines the probability of not adopting in time $t$, in other words, the probability that $T$ is greater or equal to $t$.

$$S(t) = 1 - F(t) = Pr(T > t)$$

The hazard function, $h(t)$, defines the instantaneous rate of completion of a spell at $T = t$, conditional on survival up to time $t$. Let us define as:

$$h(t) = \lim_{\Delta t \to 0} \frac{Pr(t \leq T < t + \Delta t | T \geq t)}{\Delta t}$$

$$= \lim_{\Delta t \to 0} \frac{F(t + \Delta t) - F(t)}{\Delta t (1 - F(t))} = f(t) / S(t)$$

The hazard function is a sequence of conditional probabilities continuous over a time spell (in this case, conditional probability of adoption) with $F(t)$, $S(t)$ and $h(t)$ being the different ways of expressing the distribution of $T$.

The proportional hazard model shifts the instantaneous hazard function of survival to $t$. The hazard function reflects the effects of the explanatory variables ($X$) and the baseline hazard function ($h_0$), which can be semi-parametric or follow a specific functional form according to $h_0$ for all individuals (Kiefer, 1988).

On the other hand, the two most commonly used parametric specifications in the duration models are the Weibull and the
exponential distributions. The Weibull distribution is characterized by the hazard function \( h(t) = \lambda t^{\gamma - 1} e^{-\lambda t^\gamma} \) with \( \lambda > 0 \) and \( \gamma > 0 \). If \( \gamma > 1 \), the hazard function increases over time; by contrast, if \( \gamma = 1 \), this function will decrease over time, whereas if \( \gamma = 0 \), the function will collapse, becoming an exponential function.

The proportional hazard model uses an exponential specification that ensures that the hazard function is nonnegative without imposing restrictions on the impact of the covariates on the hazard. The parameters are estimated through maximum procedures of maximum likelihood considering the censored observations and assuming the duration for each individual is independent of the others.

In empirical studies, it is often the case that adoption time occurs at discrete time intervals. The annual agricultural production cycle makes that particularly relevant for agricultural innovations, because they tend to be adopted for a specific growing season. Therefore, the model estimated in the current work follows discrete time steps in accordance with the annual nature of the decision-making process in cultivation systems (Burton et al., 2003). In these cases, the discrete hazard function can be expressed as follows (Meyer, 1990):

\[
h_a(t) = 1 - \exp\left(-\exp\left(\gamma(t) + \beta X(t)\right)\right)
\]

where \( \gamma(t) = \ln \left(\int_0^{t+1} h(t) \, dt\right) \) and \( X(t) \) is the set of explanatory variables and includes cross-sectional data and time-dependent variables for each case.

A reason for using this specification is that changes in the values of the exogenous time-dependent variables are the same for all farmers, thus imparting a component of duration dependence that is equivalent for all individual adoption probabilities. Moreover, time-dependent variables are likely to capture most of the underlying time-dependent process of innovation diffusion.

Study area and data. This study focuses on the three main regions housing greenhouses in Spain, all of which are situated in the Mediterranean area. The areas analyzed are: LOCATION-1, found between El Ejido (Almeria) and Campo de Cartagena (Murcia); LOCATION-2, mainly based in the municipalities of Lorca, Mazarrón, and Águilas (Murcia); and LOCATION-3, situated in El Ejido (Almeria). The data used here were collected by 241 questionnaires given to greenhouse agricultural producers in the second semester of 2006.

Random sampling was stratified by growing areas. The population size comprised 6916 greenhouse owners with the strata corresponding to LOCATION 1 with 1888 owners and LOCATION 2 and 3 with 1314 and 3714 owners, respectively. The numbers of surveys are distributed by strata as follows: LOCATION 1 represents 27% of the sample, LOCATION 2 represents 19% of the total, and LOCATION 3 represents 54%. If compared, the percentages of farmers comprised in each acreage range in the general farmer population are really similar to those in the sample frame in every area. From the total greenhouse growers and with a confidence level of 95%, the sample generates an error of less than 6%.

The data collection method was accomplished by personal and individual interviews of the greenhouse owners on the number indicated by the sample size chosen at random from the list of owners provided by agricultural extension services. After a pretest, these interviews were conducted by interviewers who are specialized technicians in the production carried out in these areas.

Most farmers, although this does not necessarily guarantee generational takeover. Manual labor is under great demand most of the year, which can normally be satisfied by casual labor. There has been dynamic incorporation of the different technological improvements involving large capital investment. The high levels of greenhouse specialization, together with the auxiliary industries that have developed alongside, result in large coverage of the European demand by the Spanish fresh-market vegetables. The different ways of pooling the offer at origin have also played an important role, like the agricultural cooperatives and other social economy companies, which have become the main centers responsible for technological transfer in addition to their role in marketing production.

The questionnaire was divided into four blocks of questions, which collected data concerning: 1) farmers’ personal and family data and the general characteristics of the farm, including the maximum level of greenhouse technology; 2) data concerning soilless cultivation: surface area, year of adoption, management difficulty, and so on; 3) other factors affecting the technological innovation process (farmers’ information sources, training, marketing means, and so on); and 4) producer’s environmental awareness.

Additional information was taken from official agricultural statistics and the environmental regulations concerning the use of MB during the period under analysis.

The initial hypothesis is that there are a number of factors affecting the speed at which innovation is adopted. Variables collected for the specific case of soilless cultivation are shown in Table 1. Following the outline by Pannell et al. (2006), the variables included in the study can be grouped according to the individual traits of the farmers questioned, which likewise depend on their personal characteristics, social factors, and crops. Other factors are added, which depend intrinsically on technology.

Figure 1 shows the number of farmers in the sample that adopted soilless cultivation from the moment the innovation is available to the year the data were taken. Adoption took place at the highest rate between 1998 and 2004 with very few early adoptions. Meanwhile, one can observe that in the last years, the number of farmers adopting this technology has leveled off or dropped.

Table 1. Definitions of explanatory variables.

| Variable | Meaning |
|----------|---------|
| AGE | Farmer’s age |
| ACREAGE | Greenhouse size (hectares) |
| SON-RELIEVE | 1 = if the farmer’s children will take over; 0 = otherwise |
| INFO-SOURCE | 1 = if the farmer knew about the technology through information from specialized personnel in agriculture (technology suppliers or other inputs, agricultural extension services, cooperative technicians or research centers); 0 = the farmer knew about the existence of the technology by seeing it on other farms |
| EXT-LABOR | Number of workers, not relatives, hired a year |
| ENVIRONMEN-C | 1 = the farmer considers the environmental issue in the decision-making process; 0 = otherwise |
| COOP-TRADE | 1 = the farmer belongs to a cooperative; 0 = otherwise |
| LOCATION-1 | 1 = if farm is located in Pilar de la Horadada or Campo de Cartagena (Alicante and Murcia); 0 = otherwise |
| LOCATION-2 | 1 = if farm is located in Lorca, Mazarrón or Águilas (Murcia); 0 = otherwise |
| LOCATION-3 | 1 = if farm is located in El Ejido (Almeria); 0 = otherwise |
| EXPORT1 | Tons of pepper exported in Spain the year before of study period (t-1) |
| PEPPER-PRICE1 | Pepper price in the study period (€/kg) |
| PEPPER | 1 = if farmer grows pepper; 0 = otherwise |
| TOMATO | 1 = if farmer grows tomato; 0 = otherwise |
| OTHER CROPS | 1 = if farmer do not grow pepper and tomato; 0 = otherwise |
| MANAGEMENT | Level of difficulty that the farmer has in managing soilless cultivation (0 = easy… 5 = difficult) |
| MBROMIDE-REST1 | Percent of methyl bromide restriction according to the Montreal Protocol in the study period |

\( t \) = time-varying variables.
Farmers in the survey possessed an average greenhouse surface area of 2.65 ha. Generally, they had achieved high crop specialization; 51.4% corresponded to peppers and 20.3% to tomatoes. The main characteristics of the adopters and nonadopters of soilless cultivation can be seen in Table 2. The age of the farmers adopting soilless cultivation was lower than those that did not. Likewise, the adopters had greater surface areas of greenhouses, a higher percentage of generational takeover, and more hired labor. The largest greenhouses using soilless cropping were found in Almería.

The time-dependent variables analyzed are given in Figure 2. The restrictions imposed on the use of MB began on 1998 and increased gradually until total prohibition in 2005. One might expect that the higher the levels of MB restriction, the higher the levels of soilless cultivation adoption, because it is a direct alternative to deal with soil pathology. It is also possible that as export increases, for example, in pepper-growing, farmers seek to obtain larger-higher-quality production, which soilless cultivation affords. Moreover, a relationship is expected to exist between the speed of soilless cultivation adoption and market prices of the analyzed productions.

Results and Discussion

The results derived from fitting Eq. 6 are estimated for three different specifications. Table 3 reports the maximum likelihood estimation of the parameters determining the adoption of soilless cropping systems of three models. Each of the three models presents a set of explanatory variables and, to select them, it is necessary to examine the significance of the variables and the robustness of the estimated parameters. Model 1 contains all the analyzed variables and Model 2 uses a specification that excludes those variables that were not statistically significant on the basis of t-values in Model 1, in which Model 2 is nested. In Model 3, the time-dependent variables are omitted to establish their degree of significance with the inclusion of these variables being one of the main advantages of DA.

Comparing the most general model, Model 1 with Model 2, it is possible to test the hypothesis that the coefficients of the variables EXT-LABOR ENVIRONMEN-C; COOP-TRADE, LOCATION1, PEPPER, and TOMATO are jointly zero given the likelihood ratio (LR) test statistically fails to reject the joint hypothesis (LR = 6.47; \( \chi^2_{0.05,6} = 12.59 \)). Moreover, one can see that the estimated coefficients for the variables of Model 2 are very similar to those estimated in Model 1. Likewise, on comparing Model 2 with Model 3, the LR test shows that the coefficients of the time-dependent variables are equal to 0, but the hypothesis is rejected (LR = 10.78; \( \chi^2_{0.05,4} = 9.49 \)). Therefore, Model 2 specification is preferable, and the final results will be interpreted taking this as a base.

The proportional hazard model is parameterized by a Weibull distribution with positive time dependence. A LR test for the exponential form gives a \( \chi^2_{0.05,1} \) statistic of 32.30, rejecting the hypothesis of nonduration dependence baseline hazard function. The estimated coefficients give \( \beta \) values that are interpreted as the proportional change of the hazard function achieved by a marginal change in the explanatory variable and its sign shows the direction of the impact on the hazard function. These coefficients should be interpreted as exp(\( \beta \)), where values equal to 1 imply no impact on the hazard function. However, values higher (lower) than 1 indicate a positive impact (negative) on the hazard, thus the negative relationship between the variable and the adoption time.

Model 2 includes a total of nine significant variables, which encompass the social,
Table 3. Estimates of alternative specification of Weibull hazard functions for soilless technology adoption.

|                      | Coefficient | SE  | Coefficient | SE  | Coefficient | SE  |
|----------------------|-------------|-----|-------------|-----|-------------|-----|
|                      | Model 1     |     | Model 2     |     | Model 3     |     |
| AGE                  | -0.16       | 0.03*** | -0.16       | 0.02*** | -0.16       | 0.03*** |
| ACREAGE              | 0.26        | 0.06*** | 0.26        | 0.05*** | 0.25        | 0.05*** |
| SON-RELIEVE          | 0.80        | 0.35*  | 0.81        | 0.32**  | 0.88        | 0.32**  |
| INFO-SOURCE          | 1.29        | 0.31*** | 1.33        | 0.30*** | 1.46        | 0.26*** |
| EXT-LABOR            | 0.01        | 0.01   | 0.01        | 0.01   | 0.01        | 0.01   |
| ENVIRONMEN-C         | 0.12        | 0.32   | 0.1          | 0.25   | 0.03        | 0.33   |
| COOP-TRADE           | 0.43        | 0.33   | 0.43        | 0.33   | 0.43        | 0.33   |
| LOCATION-1           | -0.05       | 0.42   | -0.05       | 0.42   | -0.05       | 0.42   |
| LOCATION-2           | -2.47       | 1.09*  | -1.46       | 0.64*  | -1.55       | 0.66*  |
| EXPORT               | 3.01e-6     | 1.49e-6* | 2.74e-6     | 1.39*  | 2.74e-6     | 1.39*  |
| PEPPER-PRICE         | -0.02       | 0.01*  | -0.02       | 0.01*  | -0.02       | 0.01*  |
| PEPPER               | -0.62       | 0.37   | -0.62       | 0.37   | -0.62       | 0.37   |
| TOMATO               | 0.10        | 0.68   | 0.1          | 0.65   | 0.1         | 0.65   |
| MANAGEMENT           | -0.50       | 0.16*  | -0.44       | 0.15** | -0.41       | 0.27*** |
| MBROMIDE-REST        | 0.02        | 0.01** | 0.02        | 0.01** | 0.02        | 0.01** |
| Cons                 | -6.48       | 2.01*** | -6.77       | 1.98*** | -5.01       | 1.16*** |
| (p-1)                | 2.81        | 0.70*** | 2.90        | 0.71*** | 2.99        | 0.46**  |
| Log likelihood       | -198.89     |       | -202.136    |       | -207.529    |       |
| AIC                  | 0.107       | 0.106  | 0.1          | 0.106  | 0.1         | 0.106  |
| Number of observations| 241         |       | 241         |       | 241         |       |

Significance levels at *P < 0.05, **P < 0.01, and ***P < 0.001.

* = time-varying variables.

Model 1 contains all the analyzed variables.

Model 2 uses a more parsimonious specification that excludes those variables that were not statistically significant in Model 1.

Model 3 omits time-dependent variables from Model 2.

cultural, and personal influences of the potential adopters and other attributes of the technology itself. Of these variables, six are statistically significant at 99% or higher to explain the conditional probabilities of adopting soilless cultivation.

Inclusion of time-varying variables representing product export data series (EXPORT) and prices (PEPPER-PRICE) as well as legal restriction on MB use (MBROMIDE-REST) has proven to be significant (Model 2 versus Model 3). This demonstrates the importance of including these variables and the existence of an underlying time-dependent process of innovation diffusion by which farmers learn about the technology, which enables them to reduce inherent uncertainty and assist decision-making, at least in cases in which technology adoption can contribute to achieving farmers’ own objectives, as reported by Marra et al. (2003).

The personal characteristics influencing adoption decisions include the farmer’s age (AGE). Although a mixed response can be found in the literature (Rogers, 2003), in our study, age was found to exert a negative effect on the likelihood of adopting soilless cultivation, confirming that the older farmers have less incentive to adopt this package.

When there is a family successor on the farm, who will benefit from investment in this technology and training required by its adoption, measured as the variable SON-RELIEVE, the conditional probability of farmers adopting it is more than doubles that of farmers who have no generational take-over perspectives.

The positive coefficient for the variable ACREAGE indicates that the farmers possessing a greater greenhouse surface area tend to increase the global benefit of adoption and, with it, the conditional probability of adoption. Generally, large farm owners have been more innovative as a result of the fact they can bear the fixed costs of implementing the technology more easily as well as having the capital necessary for its acquisition (Feder and Umali, 1993). Normally, the large greenhouses are more business-like in nature and hire labor from outside the family; however, in our study, the variable EXT-LABOR was not statistically significant.

The variable INFO-SOURCE confirms that when the farmers obtain technical information from people they consider to be experts, adoption speed is greater than those who learned about the technology by seeing it on other farms but did not seek professional advice.

Among the social influences affecting adoption speed, LOCATION-2 was the production area that proved to be significant. Moreover, this area is highly specialized in tomato-growing, and this crop does not require high-demanding management given the type of crop was not significant. Belonging to an agricultural association also exerted an influence on technology adoption according to Kington and Pannell (2003), as did the farmers’ environmental awareness with favorable effects on the adoption of different soil-conservation practices (Pannell et al., 2006). Nevertheless, the variables PEPPER, TOMATO, ENVIRONMEN-C, and COOP-TRADE did not have a significant impact on soilless cultivation adoption speed in Spanish Mediterranean greenhouses.

It is possible that there are unidentified random influences, which have not been specified by the hazard function in the duration models. To identify this possible flaw and validate the final quality of our model, considering that the specification may be incomplete, an unobserved heterogeneity component has been incorporated following the gamma distribution in accordance with the work by Meyer (1997). After its estimation, the adoption decisions made for the following campaign.

There is a response to pepper prices (PEPPER-PRICE), and those decrease in the conditional probability rate. This shows that the farmer has adopted soilless cultivation to benefit from the relative advantage it gives and thus aspires to better market prices. Moreover, this technique favors innovations in a relatively short time, so this also favors adoption speed, as reported by Marsh et al. (2000). Furthermore, Garcia-Martinez et al. (2008) demonstrated that pepper prices in the Spanish greenhouses increase considerably when there is production outside the traditional harvesting epochs, focusing on the spring months in particular, and soilless cultivation is necessary to achieve this.

The farmer’s personal perception of the set of relative advantages afforded by soilless cultivation technology, and to what extent it can help achieve personal goals, can be partially set back if the farmers perceive the technology as difficult to manage. Thus, the value showing the marginal effect exerted by the variable MANAGEMENT indicates that adoption takes longer when farmers perceive technology management to be more difficult. Therefore, in these cases, farmers need to be better informed beforehand about the consequences of adoption (Rogers, 2003).

In the same way, the policies regulating the usage of MB (MBROMIDE-REST) given the environmental problems it poses, also affect the hazard function. As restrictions of MB use have increased, in line with the Montreal Protocol, the rates in the adoption probability changes have also increased. This increase in adoption speed shows that the use of soilless cultivation not only affords productive advantages, but is also a clear alternative to solving soil-associated problems and can be quickly implemented by farmers faced with the impossibility of using MB.

Other variables analyzed in this study, which were initially expected to be significant, have turned out not to be so in our proportional hazard model. The adoption of different types of MB was not significant. The variable of crop was not significant. Belonging to an agricultural association also exerted an influence on technology adoption according to Kington and Pannell (2003), as did the farmers’ environmental awareness with favorable effects on the adoption of different soil-conservation practices (Pannell et al., 2006). Nevertheless, the variables PEPPER, TOMATO, ENVIRONMEN-C, and COOP-TRADE did not have a significant impact on soilless cultivation adoption speed in Spanish Mediterranean greenhouses.

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the procedure put forward by Jenkins (1995), unobserved heterogeneity did not pose a serious problem in this case. Therefore, one can assume that the functional form chosen in the present work is suitable for this scenario.

Conclusion

The present study has analyzed the dynamics of the timing of soilless technology adoption in Spanish Mediterranean region using DA. DA estimates farmers’ timing in adopting soilless cultivation associated with their personal characteristics, the surroundings, and aspects inherent to the technology itself. This methodology affords an improvement over the technology and encouraging better decision models. The results suggest that the conditional probability of adopting soilless cultivation technology positively depends on timing and that farmers’ personal characteristics, like being younger, owning larger greenhouse surface areas, the family’s intention to continue working the farm, and the fact that farmers receive information from professionals and do not just imitate neighboring farms, are all factors exerting a positive influence on adoption speed.

Furthermore, the conditional probability of adopting soilless cultivation is increased by commercial perspectives with the latter being influenced by the price, which in turn depends on quality. Thus, higher quality products or decreasing prices stimulate farmers to adopt, so that they can obtain better quality products and achieve higher prices.

The present study confirms the importance of the timing of knowledge acquisition, thereby diminishing uncertainties about the technology and encouraging better decision-making. In this respect, although soilless cultivation affords farmers numerous advantages and is a technique that can lessen certain negative external factors derived from intensive soil cultivation, it has not spread with sufficient speed in Spanish Mediterranean greenhouses. This demonstrates that currently the economic and commercial aspects of vegetable production still exert a stronger influence on the final decision to invest in soilless cultivation technology and are, therefore, the variables that speed up (or not) adoption to a greater extent.

These results contribute to a better understanding of the adoption process as a whole; thus, farmers can be helped to achieve their goals when seeking more knowledge about technology and its performance. Also, better informed policymakers and extension services are desirable if, given the current environmental and economic concerns, more widespread use of this technology would help develop rural areas and improve agricultural revitalization.

The delay in adopting profitable technologies, described by this approach, could be complemented by other predictive approaches such as real options, which allow us to estimate the optimal adoption time, taking into account the irreversibility and uncertainty of the investment. There is a direct relationship between these two approaches, because the real options method provides the trigger value of the output price or yield that indicates when the investment should be made considering the option value of delaying adoption within an economic framework.

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