Why Isn’t More U.S. Farmland Organic?

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Abstract

We develop a theoretical model to assess the dollar compensation required to induce conventional growers to convert to organic. The model incorporates the uncertainty in producers’ expectations about future returns and about the impact of policy changes on these expectations in particular. We demonstrate that a new policy which favours organic can have opposing effects on the rate of conversion. An increase in relative returns to organic today will increase conversion rates. However, if the future of the policy program is uncertain, its introduction can increase the value of waiting to switch, which will decrease conversion rates. We then develop an empirical switching regression model that enables direct estimation of the value associated with being able to postpone the conversion decision until some of the uncertainty is resolved. The model is applied to data on organic and conventional soybeans before and after major changes in U.S. farm policy toward organic growers. The results suggest that sunk costs associated with conversion to organic coupled with uncertainty about future returns can help to explain why there is so little organic farmland in the United States.

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1. Introduction

Organic farming is one of the fastest growing segments of U.S. agriculture. The high prices commanded by organic products and often unmet demand have enticed growers to enter the market. Certified organic farmland doubled between 1995 and 2000, and then doubled again between 2000 and 2005. In total, there were 4 million acres of cropland and pasture dedicated to organic production in 2005 (USDA, Economic Research Service). Yet despite this fast-paced growth, organic agriculture is limited to a very small fraction of U.S. farmland—approximately one half of one percent.

There are social benefits to increasing organic production. Price premiums for organic products can increase farm profitability. Likewise, consumers can benefit from greater choice at the supermarket and from lower prices due to increased competition among suppliers (Dhar and Foltz, 2006). Converting a conventional farm to organic can also increase its provision of ecosystem services. For example, in a review of the literature on the biological differences between conventional and organic systems, Hole et al. (2005) conclude that organic farms have greater species abundance and biodiversity than conventional farms. These potential benefits provide some rationale for the increasing government support of organic agriculture.

Over the past decade, U.S. agricultural policy has evolved in ways that have favoured organic farmers. This process began with the 2000 Agricultural Risk Protection Act which first made organic farms eligible for federal crop insurance. Shortly thereafter, the 2002 farm bill introduced a subsidy program for those converting to organic and a new system of payments for conservation practices. The 2008 farm bill continues these programs (with some
modification) while simultaneously expanding organic farmers’ eligibility to receive cost-share payments. Perhaps most importantly, the 2008 farm bill adds a new title for organic agriculture, signaling that organic farms will continue to benefit from future farm programs. With all of these developments, why is there still so little organic farmland?

There is virtually no theoretical or empirical evidence on the dollar compensation needed to induce farmers to adopt organic practices, despite the body of literature on organic farming and the adoption of environmentally friendlier practices in general. Most of the relevant literature has focused on the use of discrete choice methods to analyze farmers’ choice between conventional and organic technologies (Burton, Rigby, and Young, 1999; Lohr and Salomonsson, 2000; Pietola and Oude Lansink, 2001). These methods identify the factors which prompt farmers to adopt the alternative farming practices, assess their relative importance in that decision, and yield probabilities of adoption. The resulting estimates cannot be readily converted into dollar compensation levels.

A related body of previous research examines organic crop returns using data from long-term experimental trials (for an overview see McBride and Greene, 2008) or investigates the impact of organic conversion on financial performance through simulation (e.g., Archer and Kludze, 2006; Acs et al., 2009)). Importantly, these studies leave out the actual farm setting and they assume decisions are based on the expected net present value of returns to alternative land uses. While the net present value rule is simple and direct, it fails to address the role of sunk costs and uncertainty in the farmer’s decision process. These factors can be the main determinants of actual land conversion decisions (Tegene et al., 1999; Schatzki 2003; Musshoff and Hirschhauer, 2008). With federal conservation
programs that rely on land use conversion to achieve environmental policy goals we would argue that improved understanding of strategic land use decision making has become increasingly important.

This article investigates how uncertainty and sunk costs affect the farmer’s decisions about if and when to convert to organic and how this is affected by policy in particular. The basic intuition underlying our model is straightforward. There may be a value associated with the option to postpone the conversion decision until some of the uncertainty about the stream of future returns is resolved (Dixit and Pindyck, 1994). The value of this option to switch at a later date raises the dollar compensation needed to induce the farmer to convert to organic today, as first highlighted by Odening et al. (1994).

Our analysis provides three new insights into the impacts of farm policy on organic agriculture that have not been addressed in the literature. Our theoretical model illustrates how the uncertainty of the three-year transition period required to obtain organic certification provides an incentive for potential organic farmers to postpone conversion. We demonstrate that a new policy which favours organic can have opposing effects on this conversion decision. An increase in relative returns to organic today will decrease the value of waiting to switch and increase conversion rates. However, if the future of the policy is uncertain, its introduction can increase the value of waiting to convert, which will decrease conversion rates. Secondly, if there is heterogeneity in the value of the option to switch, one can separately identify the time-constant and time-variant components using a switching regime estimator. Our estimator builds on Wossink and Gardebroek (2006) who apply a switching regime to quantify the impact of policy uncertainty on the decision to invest in
tradable quota. Third, by applying this estimator to data on conventional and organic soybeans before and after shocks to farm policy that favoured organic growers, we provide the first empirical evidence for the U.S. that values for the option to switch to organic farming exist and that these values are economically significant.

We proceed as follows. Section 2 begins by developing a formal model of the decision to convert to organic in the presence of policy uncertainty. Section 3 proposes the switching regime estimator that permits identification of the latent sunk cost of transition as well as time-constant and time-variant components of the value of the option to convert. Section 4 applies the estimator to data on conventional and organic soybeans before and after the Agricultural Risk Protection Act and introduction of the Conservation Security Program. Our empirical results help to explain why organic acres changed little following the introduction of these programs. After interpreting the results, we offer concluding comments and identify next steps in this line of research.

2. Theoretical Model

Our theoretical model begins from the observation that non-organic farmers typically perceive organic farming as being more risky than conventional farming. There are many reasons why this is true. First, the inability to intervene with chemical fertilizers and pesticides can increase the costs of pest management and increase the risks of crop loss. Second, organic farming systems often have greater agronomic complexity than conventional systems, and there is less information on how to manage organic farming systems efficiently (Padel and Lampkin, 1994). Third, organic farmers are exposed to higher market risks. Organic crop prices tend to be more volatile than conventional prices,
partly due to the comparatively small scale of organic markets (Klonsky and Smith, 2002). Finally, government support of organic agriculture in the United States has been more volatile in the past than support for conventional agriculture. Prior to the 2008 farm bill, it would have been natural to expect organic growers to be less certain of their inclusion in future support programs. Even if a conventional farmer is willing to take on these various sources of production and market risk, they must still undergo a rigorous process of converting their farm from conventional to organic before they can be legally certified as an organic grower.

Survey research has demonstrated that farmers perceive the conversion process as a major obstacle to making the transition from conventional to organic (Padel, 2001). A farm in transition must keep detailed records for three years before they can receive organic certification. The physical transition cost may be incurred for several years more through penalties in yield or cost incurred due to agro-ecosystem adjustments and management inefficiency as new practices are learnt. During the conversion process, a grower also faces important financial constraints including a lack of access to premium prices until conversion is complete, conversion-related investments and disinvestments, and information gathering costs for production and marketing (Lohr, 2001). Gardebroek (2006) identifies additional non-pecuniary psychological and sociological costs of switching to organic farming.

In the presence of these sunk costs and uncertainty about future development of earnings, a risk neutral farmer will not be indifferent between organic and conventional if their current expected returns are identical. To begin to investigate the role of sunk costs and uncertainty in the farmer’s decision making process, we develop a model of conversion
decisions using real options theory (Merton, 1976; McDonald and Siegel, 1986; Dixit and Pindyck, 1994).

2.1. A Real Options Approach to Modeling Conversion to Organic

An important pre-condition for widespread adoption of organic farming, or any new technology, is its profitability. A truism in the organic community is that the more recent converters are motivated by profitability, not ideology (Lohr, 2000; Padel, 2001). Therefore, our theoretical model begins with the assumption of profit maximization.

In the current period, \( t = 0 \), a representative risk-neutral farmer values the expected lifetime returns from conventional farming as \( C_0 \). In the meantime he scans the expected returns from organic farming. Let \( O_0 \) denote the present value of the expected lifetime earnings of organic farming. To obtain \( O_0 \) the farmer must incur fixed costs that become sunk upon transition. These sunk costs include the record keeping, physical transition, management, financial and information costs discussed above. Let \( R_t \) denote the differential between the discounted stream of expected cash flows for a farmer who gives up conventional farming for organic farming at time \( t \), and let \( T \) denote the fixed transition costs. Under a conventional net present value (NPV) calculation, the farmer will make the switch from conventional to organic if and only if \( R_t - T \geq 0 \).

While the NPV approach is simple and direct, it fails to capture the role of uncertainty in the decision making process. In contrast, the real options approach focuses on the opportunity of making decisions after some uncertainties have been resolved. One would expect the value of \( R_t \) to become increasingly uncertain the further \( t \) lies in the future.
because of production and market uncertainty. One way to depict this uncertainty is to model \( R \) as following a geometric Brownian motion with drift:

\[
dR = \mu R dt + \sigma R dz.
\]  

(1)

In the equation, \( \mu \) is the expected growth rate of the stream of value differences between the discounted expected cash flows, \( \sigma \) is the standard deviation of those differences, and \( dz \) is the random increment of a standard Wiener process such that \( E[dz] = 0 \) and \( [Edz]^2 = dt \). With geometric Brownian motion the present value of switching from conventional to organic farming may be different if the transition is postponed.

Equation (1) implies that future values of the investment are log-normally distributed with expected value \( E_0[R_t] = R_0 \exp(\mu t) \), where \( E_0 \) denotes the expectation at time \( t = 0 \). For an infinitely lived investment, McDonald and Siegel (1986) demonstrate that the expected value of the investment opportunity, \( F[R] \), can be expressed as:

\[
F[R] = \max \left[ E(R_t - T)e^{-rt} \bigg| \Omega_0 \right],
\]  

(2)

where the time of investment, \( t \), is the control variable, \( r \) is the discount rate, and \( \Omega_0 \) denotes the set of information available to the farmer at \( t = 0 \). In this case, the return to investing at the beginning of the period is \( rF \) and the value of postponing the investment decision is equal to the increase in \( F \) that is expected to occur during the period.

The first order condition to the investment problem is \( rF dt = E[dF|\Omega_t] \). Using Ito’s lemma to obtain the total differential for \( dF \), the first order condition can be rewritten as

\[
\frac{\gamma}{2} \sigma^2 R^2 F''(R) + \mu RF'(R) - rF = 0.
\]  

Equation (3) provides the analytical solution.
\[ R^* = \alpha T, \quad \text{where} \quad \alpha = \beta / (\beta - 1) \quad \text{and} \quad (3) \]

\[ \beta = \frac{1}{2} - \mu / \sigma^2 + \sqrt{\left(\mu / \sigma^2 - \frac{1}{2}\right)^2 + 2r / \sigma^2}. \]

To interpret this solution, first notice that if \( r \leq \mu \), the value of the investment opportunity will be infinite and the farmer will never exercise the option to invest. We acknowledge this trivial solution and move on to investigate the more interesting case where \( r > \mu \). Let \( \delta \) denote the difference between \( r \) and \( \mu \).

If the interest rate exceeds the growth rate of the investment opportunity, such that \( r > \mu \) (i.e. \( \delta > 0 \)), equation (3) implies that \( \beta > 1 \), and therefore \( \alpha > 1 \) and \( R^* > T \). Thus, in the presence of irreversible sunk costs associated with the three-year transition period and uncertainty about future returns, the NPV principle that equates \( R \) with \( T \) no longer applies. Intuitively, the presence of uncertainty introduces a wedge between the conversion trigger \( R^* \) and the traditional NPV hurdle \( R \). This wedge, defined by \( (\alpha - 1)T \), is the value of the option to postpone a decision about whether to convert to organic because of the production and market risks reflected in \( \sigma \). The wedge can be substantial even for small levels of uncertainty about future returns.

2.2. Sensitivity of the Conversion Trigger to Policy Uncertainty

Since 2000, U.S. farm policy has evolved in ways that have differentially affected the returns to organic farmers. For example, the 2002 farm bill introduced a new program to subsidize the process of converting from conventional to organic and it also introduced a new system of direct payments for conservation practices through the Conservation Security Program (CSP).
Since many of these practices are standard for organic growers, one would expect the program to increase the relative profitability of organic farming. The 2008 farm bill continued these new programs while simultaneously expanding organic farmers’ eligibility to receive cost-share payments through the Environmental Quality Incentives Program (EQIP). By increasing the return to organic production, this set of programs decreases the $R^*$ trigger required to induce growers to convert from conventional to organic. Specifically, the net present value of a series of CSP or EQUIP payments can be subtracted from the cost of transitioning to organic farming. It can be seen from equation (3) that a decrease in $T$ decreases $R^*$. Likewise, conversion subsidies decrease the cost of transition directly, also decreasing $R^*$.

Another way in which recent farm policy is likely to have decreased the conversion trigger is by reducing the uncertainty associated with returns to future organic production. This process began with the 2000 Agricultural Risk Protection Act which first made organic growers eligible for partially subsidized crop insurance. The 2008 farm bill expanded their eligibility and, perhaps most importantly, introduced a new title that explicitly targets organic producers (Title X: Horticulture and Organic Agriculture). This sends a strong signal that organic agriculture will continue to play a role in future farm bills. These policies reduce $R^*$ by reducing the uncertainty, $\sigma$, associated with the difference in the future stream of returns from organic and conventional farming. To see this, first observe that the positive root of $\beta$ in equation (3) must satisfy the quadratic equation: 

$$G = 0.5\sigma^2 \beta (\beta - 1) + \mu \beta - r = 0.$$ 

Differentiating with respect to a change in $\sigma$ gives 

$$\frac{\partial G}{\partial \beta} \frac{\partial \beta}{\partial \sigma} + \frac{\partial G}{\partial \sigma} = 0.$$ 

It is clear that $\frac{\partial \beta}{\partial \sigma} < 0$, because $\frac{\partial G}{\partial \sigma} = \sigma \beta (\beta - 1) > 0$ and at the positive root of $\beta$, $\frac{\partial G}{\partial \beta} > 0$. **
While the direct effect of conversion subsidies, payments, and crop insurance is to reduce the value of being able to postpone the conversion decision, these policies also introduce a new dimension of uncertainty both before their actual introduction and if farmers are unsure about the life of the programs and other changes that might occur in the future. This uncertainty may, in turn, affect the value of the option to switch. The effect may increase or decrease the conversion trigger depending on the nature of the policy.

The standard case in the literature is that of policy uncertainty surrounding a potential one time subsidy. For example, Hassett and Metcalf (1999) and Isik (2004) discuss policy uncertainty for the case of one-time cost share payments. An increase in the probability of a future policy that would make these payments would delay conversion (by increasing the value of the option to wait). Alternatively, if the cost share subsidy were currently in effect an increase in the probability of withdrawal of the program would encourage adoption now (by decreasing the value of the option to wait).

To extend our option model to include policy uncertainty, we generalize equation (1) to allow a discrete jump process that provides a second source of randomness. While there are many possible sources of policy uncertainty (see Gardner (2001) for a useful review), we focus on a single fairly general case. Suppose that farmers believe that at some unknown date in the future the difference in returns from organic and conventional farming may take a discrete jump down because of a policy change that reduces the profitability of organic farming. For example, a recession of the national economy could shrink the federal budget for the environmental programs that disproportionately benefit organic farmers (e.g. CSP, EQIP). Because these programs are relatively new, their funding levels may be more sensitive
to macroeconomic shocks than the long-standing income support programs for major commodities. With this in mind, equation (1) can be rewritten as a mixed Wiener-Poisson process:

\[ dR = \mu R dt + \sigma R dz - R dq , \]

where the Poisson (policy) event is uncorrelated with \( R \) and defined as:

\[ dq = \begin{cases} 
\theta \text{ with probability } \lambda dt \\
\theta \text{ with probability } (1 - \lambda) dt 
\end{cases} . \]

Equations (4)-(5) imply that at each point in time there is a probability \( \lambda \) that returns will drop by \( \theta \) percent. In this case, the expected trend in differential returns reflects the degree of policy uncertainty: \( E[dR]/R = \mu - \lambda \theta . \)

We assume that the jump component represents non-systematic risk (Merton (1976, p. 133) in the sense that policy events related to organic farming are industry specific rather than affecting the entire agriculture sector. In this case, the drift variable \( \mu \) can be modeled as including a “compensation” term equal to \( \lambda \theta \) in addition to any deviation, \( \delta \), from the risk-free interest rate \( r \) such that:

\[ \mu = r - \delta + \lambda \theta . \]

This analytical result follows from Merton (1976) and Martzoukos and Trigeorgis (2002).

In the presence of policy uncertainty, the first order condition for the value of the opportunity to convert to organic at a future date, \( rF dt = E[dF]|_{\omega_t} \), can be rewritten as:

\[ \frac{1}{2} \sigma^2 R^2 F'(R) + \mu RF'(R) - (r + \lambda)F(R) + \lambda F[(1 + \theta)R] = 0 . \] The solution to this problem has a structure that parallels the simpler case without policy uncertainty:
\( \hat{R}^* = \hat{\alpha}T, \) where \( \hat{\alpha} = \hat{\beta} / (\hat{\beta} - 1) \) and

\[
\hat{\beta} = \{\beta\} : \frac{1}{2} \sigma^2 \beta(\beta - 1) + \mu \beta - (r + \lambda) + \lambda (1 + \theta)^\beta = 0. 
\]

Here \( \hat{\beta} \) is implicitly defined as the root to a nonlinear equation. Combining (6) and (7) illustrates how policy uncertainty can delay conversion. It can be seen from (6) that conditional on \( \theta \), an increase in \( \lambda \) will have the same effect as an increase in \( r \). It can be demonstrated from (7) that an increase in either term will decrease \( \hat{\beta} \), which will increase \( \hat{\alpha} \) and increase the conversion trigger. Intuitively, an increase in the chance that a new farm policy (which is favourable to organic) might be terminated or cut in the future increases the value of the option to wait and see whether the program will be continued before investing in conversion.

2.3. Discussion

The main implication of the real option framework is that farmers will make choices today based partly on their beliefs about future production and market uncertainty. In order to characterize the role of uncertainty in the decision making process, we have made simplifying assumptions about the nature of the stochastic process used to characterize uncertainty, about farmers’ attitudes toward risk, and about heterogeneity in the distribution of the value of the real option within the farming community. These assumptions deserve some additional discussion.

First, it is important to emphasize that our main results in (3) and (7) depend only on the properties of \( dz \). Geometric Brownian motion serves as a convenient example because it has an analytical solution. However, geometric Brownian motion and the Weiner process can,
in principle, be generalized to a broad class of continuous-time stochastic processes (Merton, 1990). For example, in a recent application to hog farming, Odening, Musshoff, and Balmann (2005) obtain similar theoretical results by assuming that \( R \) follows a mean reverting process. One could use numerical methods to develop intuition for the theoretical properties of our model under a variety of alternative parametric or nonparametric descriptions for \( dz \).

Second, our theoretical model treats farmers as if they are risk neutral. Gardebroek (2006) and Serra, Zilberman, and Gil (2008) provide preliminary evidence that both organic and conventional growers display decreasing absolute risk aversion (DARA). Incorporating risk aversion into the theoretical model would not necessarily affect the overall pattern of results or the approach we take to characterizing optional behavior. This follows from Dixit (1989). Using a general real options model, he demonstrates that when the discount rate \( r \) is replaced by a risk-adjusted rate, the solutions for the trigger value can be determined in exactly the same way. Extending our model to explicitly incorporate DARA preferences (or other non-neutral attitudes toward risk) would be an interesting topic for future research.

Third, because the theoretical model describes a representative farmer, we do not provide a statistical distribution to describe heterogeneity in the real option value within the farming community. Of course farmers differ. This makes it important to recognize that our option value model can be applied equally well to different “types” of farmers. Differently skilled farmers will differ in the relative returns they would earn from organic and conventional systems. Likewise, they may differ in their expectations about the evolution of the stream of future relative returns to organic and in their assessment of policy uncertainty. Generalizing the theoretical model to depict these sources of heterogeneity would simply
require adding farmer-specific subscripts to $R, T, \mu, \sigma$, and $\lambda$. Different farmer “types” would have different real option values, but the comparative statics from our representative agent model would still apply to each type.

As we move from theory to practice, the key variable of interest is the “reduced form” multiplication factor $\alpha$, which reflects all of the structural parameters ($\mu, \sigma, \lambda, \theta$). The multiplication factor is what distinguishes the NPV trigger, $R$, from its real-options counterpart $R^*$ and has the potential to influence the rates at which farmers convert to organic. Studying the evolution of $\alpha$ over time offers the potential to reveal how changes in farm policy influence the value of postponing the conversion decision.2

Our model illustrates how a change in agricultural policy that seemingly favours organic farmers can have opposing effects on the conversion trigger. Holding the transition cost fixed, an increase in returns to investment through direct payments, for example, would decrease the trigger. At the same time, if there is uncertainty about the continuity of the policy or about the actual level of payments, producers will have an incentive to postpone their investment decision until the uncertainty is resolved.3 For example, the 2000 Agricultural Risk Protection Act and the 2002 Farm Bill both introduced polices that, once enacted, would increase relative returns to organic. Yet if growers were uncertain about the continuation of these policies in future farm bills, this added source of uncertainty could actually increase the value of postponing the conversion decision. Our next task is to develop an empirical model capable of identifying the net effect of these opposing forces.

3. Empirical Model

We envision that conventional growers differ in the values they assign to having the option
to convert. In addition to the sources of grower heterogeneity described above, real option values may differ systematically across space according to differences in market access and differences in experience. One would expect conventional growers to look to their organic neighbours to learn about transition costs, production costs, and the organic supply chain, for example. Where there is a more established population of organic growers the remaining conventional growers may be more certain about access to processors and about production costs. Systematic differences in real option values across space would support a strategy to identify the distribution of values in the farmer population. We implemented our model at the regional/county level. This avoids the need to have a complete set of individualized data; farm and farmer specific conditions both for conventional and organic are averaged out.

We develop a simple switching regime model that is capable of identifying real option values when there are two types of regions: those with positive values and those with zero values. In the latter case, the various sources of uncertainty have been resolved in the minds of conventional growers so that the difference in returns between conventional and organic simply equals the transition cost. In the formative regions, considerable uncertainty remains, driving a wedge between the conventional net present value trigger and the real option value trigger. Under the maintained assumption that farmers/regions can be classified according to these two categories, the observed difference in returns can be used to identify the (constant) transition cost and a parameter that represents the real option value “wedge”.

The differential between the discounted expected future stream of cash flows from organic and conventional farming can be attributed to unobservable transition costs and a real
option value, as in (3). Assuming this value is nonnegative, the differential can be depicted as follows for an individual observation $i$:

$$R_i = e^{aT_i} \quad \text{if positive real option value} \quad (8)$$

$$R_i = T_i \quad \text{if no real option value} \quad (9)$$

Following Spiller and Huang (1986) we specify transition cost $T_i$ as a random variable with constant mean:

$$T_i = Te^{\nu_i}, \quad (10)$$

where $T$ is the mean and $\nu_i$ is a normally distributed random variable with zero mean and constant variance $\sigma_{\nu}^2$. Substituting (10) into (8) and (9) and taking natural logs gives:

$$\ln R_i = a + \ln T + \nu_i, \quad \text{if positive real option value} \quad (11)$$

$$\ln R_i = \ln T + \nu_i, \quad \text{if no real option value.} \quad (12)$$

Equations (11) and (12) may be expressed as a switching regression system with a probability $\ell$ of observing an option value and a probability $1 - \ell$ of observing no option value. When returns follow a random walk, the rational expectations forecast for future returns is simply today’s returns. The definition of $R_i$ in the empirical model is based on this principle:

$$\ln \left( \frac{O_{it} - C_{it}}{r_t} \right) = a + \ln T + \nu_{it}, \quad \text{with probability } \ell. \quad (13)$$

$$\ln \left( \frac{O_{it} - C_{it}}{r_t} \right) = \ln T + \nu_{it}, \quad \text{with probability } 1 - \ell. \quad (14)$$

Here, $O_{it}$ is net earnings in organic farming for region $i$ at time $t$, $C_{it}$ is observed net earnings in conventional farming, $r_t$ is the interest rate, $\nu_{it}$ is a normally distributed iid
random variable, and $a$ and $T$ are parameters to be estimated. If equation (14) prevails, production and market risk are of no significance to organic growers. In the case of equation (13), there is a positive real option value if $a > 0$. Using the estimates of $T$ and $a$, real option values can be calculated as $(e^a - 1)T$.

Equations (13) and (14) imply that real option values are time-constant. To allow for temporal shocks to the value we can replace equation (8) with $R_i = e^{a + D_i \gamma} T_i$, where $D_i = 1$ if the observation relates to the specific year. In this case, equation (13) is replaced with (15).

$$\ln\left(\frac{O_{it} - C_{it}}{r_i}\right) = a + D_{it} \gamma + \ln T + \nu_{it}. \tag{15}$$

All of the parameters of the switching regime model defined in (14)-(15) can be estimated by maximizing the log of the following likelihood function:

$$L = \prod_{i=1}^{N} [l_1 f_{i}^1 + (1-l_1) f_{i}^2]. \tag{16}$$

Here, $N$ represents the total number of observations in both regimes, and $f_{i}^1$ and $f_{i}^2$ denote the density functions:

$$f_{i}^1 = \frac{1}{\sigma_v} \phi\left(\frac{Z_1}{\sigma_v}\right), \quad \text{where} \quad Z_1 = \ln\left(\frac{(O_{it} - C_{it})}{r_i}\right) - a - D_{it} \gamma - \ln T \tag{17}$$

$$f_{i}^2 = \frac{1}{\sigma_v} \phi\left(\frac{Z_2}{\sigma_v}\right), \quad \text{where} \quad Z_2 = \ln\left(\frac{(O_{it} - C_{it})}{r_i}\right) - \ln T, \tag{18}$$

where $\phi$ is the density of a standard normal variable. With access to data on the differential returns to organic and conventional production, estimation of (16) can identify the constant transition cost ($T$), the real option value parameter ($a$), and discrete shifts in real option
value that occur over time ($\gamma$) as uncertainty begins to be resolved and formative markets begin to mature. We estimate this model using data on returns to conventional and organic soybeans before and after farm policy shocks.

4. Estimating the Impact of Farm Policy Changes on the Option to Switch to Organic

The summer and fall of 2000 marked the beginning of a dramatic change in U.S. farm policy toward organic growers. In June, the Agricultural Risk Protection Act was passed. Prior to the Act, organic growers were eligible for less financial support than their conventional counterparts. While organic growers received compensation from standard commodity support programs, federal crop insurance policies would generally not compensate them for losses because organic farming prevented the use of some of the (chemical) techniques expected under the official definition of “good farming practices”. This changed under the 2000 Agricultural Risk Protection Act (ARPA) and crop insurance for many organic crops first became available by May 2001.

Then, in October, 2000, the senate began to debate a proposal introduced by Senator Harkin to amend the farm bill to establish the Conservation Security Program (CSP). This program, which eventually became part of the 2002 Farm Bill, introduced a new system of direct payments to growers for ongoing environmental stewardship on agricultural land. This was the first time that a farm bill contained an explicit provision for “green” payments, and the proposed system for distributing payments favoured organic growers. The duration and amount of payments depended on the extent of conservation work performed on the farm and on the number of acres enrolled. Payment rates differed across three “tiers” of conservation activity, with a payment ceiling of $45,000 per farm.
The bill introduced to the Senate in October, 2000 envisioned the CSP as an entitlement program. However, because of funding limits, it was initially operated on the basis of individual watersheds, and rotated around the country with signups for different areas each year. Program officials speculated that by 2010, every farmer in the U.S. would have had an opportunity to enroll (USDA, 2004). This did not happen. Funding limits required the program to be scaled back, and it was eventually restructured (and renamed) in the 2008 Farm Bill. However, organic farmers are unlikely to have foreseen this in 2001, the last year of our study period.

4.1. Data on Conventional and Organic Soybeans

We obtained data on individual organic soybean farms from the Organic Farm Research Foundation, which originally collected the data through national surveys in 1997 and 2001. The survey was discontinued after 2001 and the data we were able to obtain represent the most comprehensive source of micro data on organic farmers that is currently available. In recent years, the USDA has also collected data on organic production practices for a restricted number of crops in certain States through a sub-sample in their Agricultural Resource Management Survey. While the OFRF survey is not designed to be as comprehensive as the Census of Agriculture it represents a sizable portion of certified organic farms (approximately 23%). Moreover, soybeans account for a sizeable share of organic farmland in the United States. For example, in 2001, soybeans accounted for 16% of all certified organic cropland in the U.S. other than hay and pasture (Greene and Kremen, 2003).

Each organic farm-year observation in the OFRF data consists of a set of economic
variables and a spatial identifier. The spatial identifier includes the state and zip code. The economic variables include acreage, production, price, and yield. Together, these variables allow us to construct county-level organic prices and yields. As a proxy for conventional returns, we use county-level data on price and yield reported by the USDA National Agricultural Statistics Service. We use gross returns as a proxy for profitability under the maintained assumption that there is little difference between operating costs for conventional and organic soybeans. This is consistent with McBride and Greene (2008).9

After eliminating OFRF observations that lacked information on either price or yield, we were left with 247 farm level observations that stretch across 142 counties in 16 states. For 1997 these data represent 16% of all certified organic soybean acres in the nation. While the number of soybean acres certified as organic expanded in 2001, the size of the sample did not and the percentage dropped to 6%. Nevertheless, this sample is still quite large relative to the standard sampling frequencies for Census micro data.

Table 1 provides summary statistics for our data. They conform to the conventional wisdom about the differences between conventional and organic farming. Yields for conventional varieties are higher, but their average prices are lower. The higher price premiums more than compensate for lower average yields so that organic soybeans have higher average returns. At the same time, organic returns are more variable.

After omitting the counties with only one observation for organic production, we were left with data on 45 counties, each of which is treated as a unique observation for the purposes of estimation. These data were used to estimate the switching regime model with and without an indicator variable to allow for changes in policy uncertainty between 1997
and 2001. The coefficient on the indicator variable provides a test of whether the change in farm policy toward organic growers conveyed by passage of the Agricultural Risk Protection Act and introduction of the Conservation Security Program in the Senate altered the value associated with postponing the conversion decision.

4.2. Results

Before interpreting the results, it is important to highlight a caveat to the identification of the model. Without data on the interest rate \( r \), one cannot separately identify \( r \) and \( T \). This can be seen from (13)-(14). Therefore, we maximize the likelihood function for a range of interest rates between 0.05 and 0.15. The lower bound on this range is defined by the annual average return on 6-month T-bills, which was approximately 5% between 1997 and 2001. The upper bound is set to 15% to recognize that the relevant interest rate must also incorporate non-diversifiable risk associated with organic farming. We expect this risk to be substantial because price premiums in organic markets may depend on supply and demand factors that are largely uncorrelated with conventional markets. This helps to explain why organic farmers were ineligible to purchase government crop insurance before 2001.

Finally, we use a baseline interest rate of 10% as the focal point for model comparison.

Table 2 summarizes the results from estimating the likelihood function for the switching regime model under the maintained assumption of a time-constant real option value. The estimate for the regime probability \( \ell \) is consistent with the presence of real option values in a subset of markets. Its point estimate, 0.91, suggests that the regime with an option value was dominant. The estimate for the conversion cost, \( T \), ranges from $267 to $801 per acre. The option value parameter, \( a \), is positive and statistically significant. The
last row of table 2 reports the real option value implied by our parameter estimates for the parameter \( a \) and the conversion cost, \( T \), for each of five interest rates. The real option value ranges from $2,588 to $3,882 per acre.

Table 3 reports the results from estimating the likelihood function for the more general version of the switching regime model, where the option value is allowed to vary over time in response to changes in policy uncertainty. Estimates for the option model parameters \( (a, T, \ell, \sigma) \) are not statistically different from the more restrictive version of the model. However, we find a significant 30% reduction in the option value between 1997 and 2001. For example, conditional on a discount rate of 0.1, the value of being able to postpone the decision to convert to organic decreased from $2,124 in 1997 to $1,451 in 2001.

There are at least three potential explanations for this decrease in our estimate for the option value. One possibility is that the passage of the Agricultural Risk Protection Act and the introduction of the Conservation Security Program in 2000 reduced growers’ uncertainty about future organic returns. The reduction in uncertainty would decrease their incentive to further postpone the conversion decision. An alternative explanation is that the increase in the quantity of organic soybeans grown between 1997 and 2001 led growers to revise their expectations for the future organic price premium. A lower expectation for the future price premium would also decrease the value of waiting to convert. A third explanation would be that of a “growing knowledge stock”. In time, the direct costs of transition may reduce with an increase in the stock of public knowledge on how to grow organically and how to go through the transition process.

Finally, we use the estimates in Table 3 to compare actual CSP payment levels with
the predicted level of “green” payments that would be required to induce a risk neutral conventional farmer to switch to organic. Assuming an interest rate of 10 percent, a conventional soybean farmer would need to receive a one-time payment of $462 per acre as a compensation for the conversion cost and an additional $1,451 per acre to cover the long run higher production and market risk of the organic practices. Combining these two values implies an annual payment of $311 per acre for a 10-year contract (the maximum contract length allowed under the CSP).

The original CSP limit of $45,000 for total annual payments implies a farm could receive a payment of $311 per acre per year for up to 144 acres. However, to qualify for payments, farmers would have needed to implement conservation practices on their entire operation and this could have served as an important barrier to participation given the size of the average soybean farm. According to the 2002 Census of Agriculture, the average soybean farm harvested 228 acres of soybeans. This provides a lower bound on the size of the farm itself, since other crops may have been harvested as well. Thus, our estimates imply that the Conservation Security Program payment levels were too low to offset the option value associated with postponing conversion in 2001 for the majority of conventional soybean growers. This helps to explain why organic soybean acreage has not increased from its peak in 2001.

5. Conclusions

We have developed a formal model to assess the dollar compensation required for widespread conversion of land use from conventional to organic farming. Integrating production and market uncertainty into the model, given sunk costs, implies that the organic
price premium which implicitly defines the optimal conversion trigger may be much higher than the trigger that would be predicted by the conventional net present value model. A difference in triggers leads to inertia in conversion: farmers will wait until the option value trigger is exceeded before switching to organic. We demonstrate that a new policy which favours organic can have opposing effects on the rate of conversion. An increase in relative returns to organic today will increase conversion rates. However, if the future of the policy program is uncertain, its introduction can increase the value of waiting to switch which will decrease conversion rates. These theoretical results have potentially significant implications for (a) the design of future agri-environmental programs that rely on land conversion, and for (b) the evaluation of program performance.

An empirical switching regime model was developed to test key implications from the theory. This model enables direct estimation of the transition cost and the option value. In future work, the estimator could be applied to other environmentally friendly land use practices or bundles of practices in order to provide information on the payment levels needed to trigger participation. The model could also be relevant for situations where the expected returns to the conventional practice under consideration are unobserved but can be constructed based on biophysical and agronomic information, as in Schatzki (2003) and Kurkalova, Kling, and Zhao (2006).

Applying the model to conventional and organic soybeans provided strong evidence that is consistent with the real option value theory. However, it is also important to acknowledge that other theoretical explanations can be offered for the empirical results including adjustment of expectations, each farmer’s unique degree of risk aversion, and
market imperfections. Thus our result adds to other economic, social and psychological explanations of the friction in adoption of organic practices. Previous studies cite organic marketing constraints, the social acceptance of organic technologies, and farmers’ attitude towards environmental problems.

Our estimate for the level of payments that would trigger conversion helps to explain why organic soybean acres did not increase following the introduction of the Conservation Security Program. Assuming an interest rate of 10%, our results imply that the typical conventional soybean grower would have needed to receive an annual payment of $311 per acre for a 10 year contact to induce them to convert to organic. This is much larger than the maximum payment that could have been obtained by the average soybean farmer through the Conservation Security Program ($197).

There are several ways in which this research could be extended. One possibility would be to generalize the empirical model to include a drift term. Bentsson and Olhager (2002) suggest that real option values may drift systematically according to a product’s maturity. Our estimates for the real option value decreased substantially between 1997 and 2001. One explanation for this phenomenon is that the value of postponing conversion decreased due to major changes in farm policy in the summer and fall of 2000. An alternative explanation is that the increased supply of organic soybeans during this period changed growers’ expectations about future returns. Adding a drift term to the model would make it possible to distinguish these two explanations using a longer panel of data. A longer panel would also make it possible to test the maintained assumption of geometric Brownian motion against alternative hypotheses for the stochastic process that guides the evolution of
A second task for future research would be to generalize both the theoretical and empirical models to develop a more flexible treatment of option values and policy uncertainty. While we model the impact of uncertainty on the decision to convert from conventional to organic, farmers also have the option to revert back to conventional practices. Both options however interact and thus could be analysed simultaneously. We did not go this route because our data revealed that organic returns were generally higher than for conventional. In the longer run this could change. A ‘third regime’ would also have policy implications. In particular, unanticipated changes in incentive policies could leave some organic growers ‘stranded’ in organic where the negative real option value keeps them from converting back to conventional despite losing money.

Our empirical specification depicts markets with and without option values, but treats the option value as being constant across space at a point in time. Increasing the resolution of the data to include information on the heterogeneous prices and costs faced by the population of growers would make it possible to extend the estimator to attempt to identify parameters describing a continuous distribution of option values. Finally, another way in which heterogeneity may be important is that farmers may assign different risk probabilities to the continuation of different farm programs. Investigating the implications of this heterogeneity would require extending the theoretical and empirical models to address the impact of multiple sources of rare events.
Footnotes

1. The $\delta$ parameter can be interpreted generally as any deviation from the equilibrium rate of return that creates an incentive for investment. Martzoukos and Trigeorgis (2002) discuss alternative interpretations of $\delta$ in the finance literature.

2. Estimation of the structural parameters ($\mu, \sigma, \lambda, \theta$), while ambitious, offers the potential for counterfactual policy simulation, making it an important topic for future research.

3. The option value implications of combining payments with policy uncertainty can also help to explain other investment phenomena. For example, Miller and Zhang (1996) suggest that a similar combination of circumstances can help to explain the impact of the 1990-1991 Gulf War on oil field development.

4. Key features of the program were outlined in two senate bills, S. 3223 and S. 3260, introduced in October 2000. A less detailed version of the proposal (S. 1426) was introduced in July 1999, and a subsequent revision (S. 932) was introduced in May 2001.

5. The 2002 farm bill also funded “agricultural management assistance” which could be used to pay for 75% of the cost of conversion to organic, up to $50,000 per producer. However, it is not clear that conventional growers could have foreseen the introduction of this provision during our study period.

6. By virtue of their certification, organic farmers automatically satisfied a subset of the conservation practices required to receive funding. It was widely believed that the average organic farm would be closer to enrollment in the program than the average conventional farm and certified and in-transition organic farmers would be likely to qualify for the highest level of payments (Lohr 2001).
7. A full description of the three tiers and their associated payment levels is provided in USDA (2004). The key details about the conservation practices and payment levels were included in the bills introduced to the senate in October 2000.

8. When we first started our project, the OFRF was considering a survey for 2006/7 and we would have included this material in our analysis. However, the survey was postponed and eventually discontinued.

9. Delate et al. (2003), Delate and Cambardella (2004) find that higher seed, fieldwork and production costs incurred in organic corn and soybean production are slightly outweighed by higher pesticide and fertilizer costs for conventional varieties. We choose to take a conservative stance and assume equal production costs for conventional and organic. This is in line with McBride and Greene (2008, table 2) who report statistically insignificant differences in the operating cost for organic and conventional soybeans in dollars per planted acre based on survey data.

References


pp. 133-146.


Table 1. Summary Statistics, Conventional and Organic Soybeans, 1997 and 2001 \(^a\)

<table>
<thead>
<tr>
<th>Year</th>
<th>Yield (bu/acre)</th>
<th>Price ($/bu)</th>
<th>Gross Returns ($/acre)</th>
<th>Organic Acres/Farm</th>
<th># Organic Observations</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>organic</td>
<td>conventional</td>
<td>organic</td>
<td>conventional</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>1997</td>
<td>30.7</td>
<td>41.5</td>
<td>17.6</td>
<td>7.1</td>
<td>538.1</td>
</tr>
<tr>
<td></td>
<td>(9.9)</td>
<td>(5.7)</td>
<td>(3.7)</td>
<td>(0.2)</td>
<td>(221.0)</td>
</tr>
<tr>
<td>2001</td>
<td>27.4</td>
<td>38.7</td>
<td>12.3</td>
<td>5.3</td>
<td>339.7</td>
</tr>
<tr>
<td></td>
<td>(11.0)</td>
<td>(7.6)</td>
<td>(2.6)</td>
<td>(0.00)</td>
<td>(155.2)</td>
</tr>
</tbody>
</table>

\(^a\) Note: prices and returns are reported in $2001. Standard deviations are in brackets. The 2001 Soybean price is based on the loan rate.
Table 2. Option Value Estimates for Soybean Data

<table>
<thead>
<tr>
<th>Parameter</th>
<th>symbol</th>
<th>Discount Rate</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td>r = 0.05</td>
</tr>
<tr>
<td>Option value parameter</td>
<td>a</td>
<td>1.77***</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(0.28)</td>
</tr>
<tr>
<td>Mean conversion cost/acre</td>
<td>T</td>
<td>801.38**</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(219.25)</td>
</tr>
<tr>
<td>LLF probability</td>
<td>/</td>
<td>0.91***</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(0.05)</td>
</tr>
<tr>
<td>Standard deviation</td>
<td>σ_v</td>
<td>0.42***</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(0.05)</td>
</tr>
<tr>
<td>Implied option value ($/acre)</td>
<td></td>
<td>3,882</td>
</tr>
</tbody>
</table>

Note: standard errors are in brackets. ***, **: Significant at 2.5%, 5% (two-sided).
Table 3. Option Value Estimates for Soybean Data, with Policy Uncertainty

<table>
<thead>
<tr>
<th>Parameter</th>
<th>symbol</th>
<th>Discount Rate</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td>$r = 0.05$</td>
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<tr>
<td>Option value parameter</td>
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<td>1.74***</td>
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<tr>
<td>Mean conversion cost/acre</td>
<td>$T$</td>
<td>909.73**</td>
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<tr>
<td>LLF probability</td>
<td>$l$</td>
<td>0.92***</td>
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<tr>
<td>Standard deviation</td>
<td>$\sigma_v$</td>
<td>0.40***</td>
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<tr>
<td>Dummy, year=2001</td>
<td>$\gamma$</td>
<td>-0.30**</td>
</tr>
<tr>
<td>1997 Option Value ($/acre)</td>
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<tr>
<td>2001 Option Value ($/acre)</td>
<td></td>
<td>2,922</td>
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</tbody>
</table>

Note: standard errors are in brackets. ***, **: Significant at 2.5%, 5% (two-sided).