Assembling the model

Numerical results

Policy Uncertainty

Using real option analysis to quantify the effects of ethanol policy on production A PDE-based stochastic control approach

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Ethanol background

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 - The facility
 - Stochastic price model
 - Stochastic optimal control

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- Increased correlation
- Subsidy policy

Policy Uncertainty

- Policy History
- A simple model
- Conclusions

Assembling the model

Numerical results

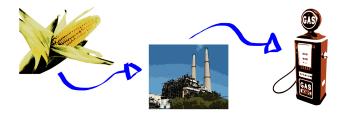
Policy Uncertainty

Ethanol biofuel production in North America has grown in popularity due to environmental and economic reasons. It has become politically attractive to subsidize biofuel programs (ethanol in particular), however subject to much policy uncertainty.

Idea

Can real options and mathematical finance aid as a tool for policy analysis?

Corn ethanol in a few words: You make money on the spread.



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Assembling the model

A model for studying policy impact on ethanol production will require:

- A model for the plant including flexibility control, capitalized construction costs, profit as a function of ethanol/corn, costs to pause/resume production
- A stochastic model for corn and ethanol prices
- An optimal operating rule which maximizes expected future profits

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The facility

Management flexibility (control)

Depending on price conditions, management has the option to:

- Enter into the investment with a capitalized cost of B dollars per gallon of production capacity
- Start or stop production depending on expected profitability (i.e. to avoid running at a loss over sustained periods and to recapture profits if price become favourable again)
 - To start production on from off costs a penalty D_{01} dollars per gallon
 - To stop production incurs a penalty of D_{10}

The two main states are on (1) and off (0).

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The facility			
The process			

When in regime *i* our running profit is $f_i(...)$

 $\underbrace{\text{profit}}_{f_1} = \underbrace{\text{yield}}_{\kappa} \left(\underbrace{\text{ethanol}}_{X_t} + \underbrace{\text{subsidy} - \text{running costs}}_{s-p=-K_1} \right) - \underbrace{\text{corn}}_{Y_t}$ $f_1 = \kappa(X_t - K_1) - Y_t$

1 bushel of corn produces κ gallons of ethanol
K_i is the fixed running cost per gallon in state *i* While idle, the running costs are fixed f₀ = -κK₀.

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Stochastic price model			
Model paramete	ers		

Statistical tests found weak evidence for mean reversion or seasonality (Kirby & Davison, 2010), hence a joint GBM diffusion was chosen.

Ethanol: X_t ---Ethanol, L,, \$/gallon Corn: Y_t Corn. C., \$/bushel $dX_t/X_t = \mu dt + \sigma dW_{1t}$ rice $dY_t/Y_t = adt + bdW_{2t}$ $\operatorname{Corr}[W_{1t}, W_{2t}] = \rho$ 10

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Parameters were estimated using basic regression analysis.

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Stochastic optimal control			

Expected earnings

Total expected discounted earnings

$$J_i(t,\alpha,x,y) = E\left[\int_t^T e^{-r(s-t)} f_{l_s}(s,X_s,Y_s) ds - \sum_{k=1}^n e^{-r(\tau_k-t)} D_{i_{k-1},i_k} \middle| \mathcal{F}_t\right]$$

where $\alpha = (\tau_k, i_k)$ is a switching control sequence between the on and off states

- t ≤ τ_k ≤ T denotes an increasing sequence of stopping times (when to switch),
- and corresponding states *i_k* (where to switch),

• with
$$I_t = \sum_{0 \le k \le n} i_k \mathbb{1}_{[\tau_k, \tau_{k+1})}$$

We assume the facility has no salvage value at the end of its useful life $V_i(T, X_T, Y_T) = 0$

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Stochastic optimal control			

Optimal choice

The value function in state *i* at time *t* assuming subsequent optimal control is $V_i(t, x, y) = \sup_{\alpha} J_i(t, \alpha, x, y)$. Dynamic programming principle:

$$V_{i}(t, x, y) = \sup_{\tau} E\left[\int_{t}^{\tau} e^{-r(s-t)} f_{i}(s, X_{s}, Y_{s}) ds + \max_{0 \le j \le m} e^{-r(\tau-t)} \left\{ V_{j}(\tau, X_{\tau}, Y_{\tau}) - D_{ij} \right\} \mathbb{1}_{\{\tau < T\}} \middle| \mathcal{F}_{t} \right]$$

Ethanol background	Assembling the model	Numerical results 000	Policy Uncertainty 0000000
Stochastic optimal control			

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So either (A) it is optimal to stay in the current state *i* and the PDE is satisfied C_i :

$$rac{\partial V_i}{\partial t} + \mathcal{L}[V_i] + f_i(t, x, y) - rV_i = 0, \quad V_i \ge V_j - D_{ij}$$

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Stochastic optimal control			

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or (B) it is optimal to switch S_{ij}

$$V_i = \max_{0 \le j \le m} V_j - D_{ij}$$

where \mathcal{L} is the generator associated with the diffusion.

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Stochastic optimal control

This is a system of free boundary PDEs (or linear complimentary problem)

Variational inequalities

$$\max\left(\underbrace{\frac{\partial V_i}{\partial t} + \mathcal{L}[V_i] + f_i(t, x, y) - rV_i}_{C_i}, \underbrace{\max_{0 \le j \le m} [(V_j - D_{ij}) - V_i]}_{S_{ij}}\right) = 0$$

where the generator $\boldsymbol{\mathcal{L}}$ is

$$\mathcal{L} = \mu x \frac{\partial}{\partial x} + \frac{1}{2}\sigma^2 x^2 \frac{\partial^2}{\partial x^2} + \rho \sigma x b y \frac{\partial^2}{\partial x \partial y} + \frac{1}{2}b^2 y^2 \frac{\partial^2}{\partial y^2} + a y \frac{\partial}{\partial y}$$

In other words:

$$PDE \leq 0$$
, $Constraint \leq 0$, $PDE \times constraint = 0$

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Stochastic optimal control

Entry into investment

Suppose the firm has a finite lease time on the green field site to build the facility. The decision to enter is analogous to an American call struck at B on the potential future earnings (compound chooser option).

$$\max\left\{\frac{\partial V}{\partial t} + \mathcal{L}[V] - rV, \quad \max\left[V_1(x, y, t), V_0(x, y, t)\right] - B\right\} = 0$$

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Stochastic optimal control

Numerical method - Finite differences

We call the values at the grid points $V(x_i, y_j, t_k) = V_{i,j}^k$

• $t_{k} = t_{0} + k\Delta t$ • $x_{i} = x_{0} + i\Delta x$ • $y_{j} = y_{0} + j\Delta y$ $\frac{V^{k+1} - V^{k}}{\Delta t} + \underbrace{LV^{k}}_{L=differentiation matrix} = f^{k} \leq 0$

Leads to coupled linear complimentary problem

$$MV - b \le 0, V \ge g, (MV - b)^T (V - g) = 0$$

where obstacle $g = V_* - D_*$, $b = -V^{k+1} - f^k$, $M = (\Delta tL - 1)$.

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Stochastic optimal control

Effects of ρ : lessons from exchange options

If there are no switching costs, then this is a running exchange option:

$$V(x, y, t) = E_{x, y} \left[\int_{t}^{T} (\kappa X_{s} - Y_{s})^{+} ds \middle| \mathcal{F}_{t} \right]$$

which has solution

$$V(t,x,y) = \int_t^T e^{-r(s-t)} \left[\kappa x e^{\mu(s-t)} \Phi(q_+) - y e^{a(s-t)} \Phi(q_-) \right] ds$$

where

$$q_{\pm} = \frac{\ln\left(\frac{\kappa x e^{\mu(s-t)}}{y e^{s(s-t)}}\right)}{\nu \sqrt{s-t}} \pm \frac{1}{2} \nu \sqrt{s-t}$$
$$\nu^2 = \sigma^2 - 2\rho \sigma b + b^2$$

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Outline of policy investigation

We investigate the effects of:

- increased correlation
- increased corn prices
- reductions in ethanol subsidy

on the following:

- plant value
- operating characteristics
- decision to enter into investment

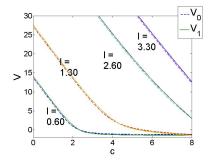


Figure: Level lines of ethanol L vs corn C.

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Effects of a			

Losses in value due to ρ on a 10 year horizon problem.

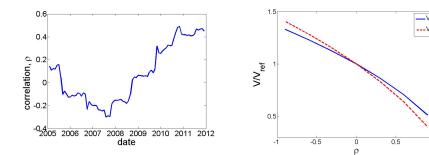


Figure: 3-year rolling correlation Jan/04–Dec/11.

Figure: $V_i/V_{i,ref}$, the option loses much value as *rho* increases.

Assembling the model

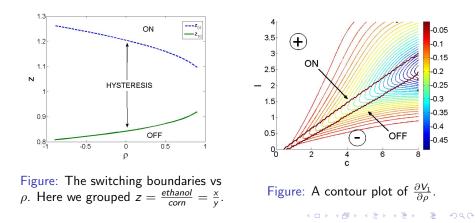
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Increased correlation

Loss in value and effects on operating decisions ρ

As ethanol policy induces more players into the market, the projects become less valuable due to increased correlation between corn/ethanol driven by ethanolic biofuel production demand.



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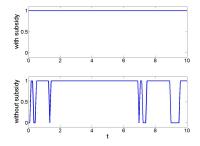
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Subsidy policy

Effects of subsidy policy on operation



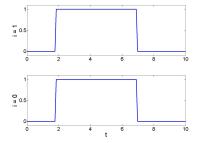


Figure: Positive profits with/without subsidy using historical price series $1_{f_1>0}$.

Figure: An operating signal using the historical realized price series in the absence of subsidy.

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Subsidy policy

Effects of subsidy policy on entry into investment

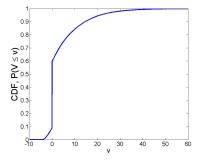


Figure: A large number of projects are stalled in the absence of subsidy indicated by spike at zero.

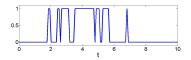


Figure: An investment signal using the historical realized price series in the absence of subsidy.

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Policy History

Policy uncertainty in ethanol subsidy

The level of subsidy has varied over the years. This subjects investment decisions to added uncertainty.

Act	Year	Subsidy (\$/gallon)
Energy Tax Act	1978	0.40
Surface Transportation Assistance Act	1983	0.50
Tax Reform Act	1984	0.60
Omnibus Budget Reconciliation Act	1990	0.54
1998 change effective 2001	2001	0.53
1998 change again effective 2003	2003	0.52
Extension to 2007 but reduced in 2005	2005	0.51
Farm Bill	2008	0.45

Can we account for this?

It is not easily hedged, but we can try model it stochastically...

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Numerical results

Policy Uncertainty

A simple model

Subsidy jump process

Choose a simple Merton style jump model for the subsidy S_t

$$dS_t = (J - S_t)dq, J \sim \log N(a, b)$$

and dq is a Poisson process with arrival rate λ

To simplify the analysis, group $\kappa X - Y$ into Z. Choose Z to be a GBM and estimate parameters by shifting the series up by z_{min} .

$$\underbrace{\underset{f}{\text{profit}} = \underbrace{\text{yield}}_{\kappa} \quad (\underbrace{\text{ethanol}}_{X_t} + \underbrace{\text{subsidy}}_{S_t} - \underbrace{\text{running costs}}_{p} - \underbrace{\text{corn}}_{Y_t}$$
$$f = \underbrace{\kappa X_t - Y_t}_{Z_t} - \kappa p + \kappa S_t$$

The profit function is then: $f(Z_t, S_t) = Z_t - z_{min} - \kappa(p - S_t)$

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A simple model

PIDVI and numerical method

The coupled control problem for the operator is given by PIDVI

$$\max\left(\frac{\partial V_i}{\partial t} + \mathcal{L}[V_i] + f_i(z,s) + \lambda E[V_i(J)] - (r+\lambda)V_i, \quad (V_j - D_{ij}) - V_i\right) = 0$$

To account for the integral, just discretize the sum and throw it into differentiation matrix:

$$E\left[V(x, J, t)\right] = \int_0^{S_{max}} V(z, s, t)g(s)ds = \sum_{j=0}^{S_{max}} V_{i,j}^k g(j\Delta S)\Delta S$$

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A simple model

Value and decision boundaries

Here are the baseline cases at different subsidy levels. Recall

 $Z = \kappa X - Y$

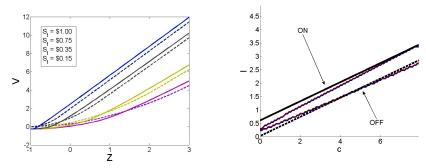


Figure: Values at different spreads and subsidies.

Figure: 1D switching boundaries overlaid onto 2D model: $X = \frac{1}{\kappa} (\theta + Y_{s}) = 0.00$

Assembling the model

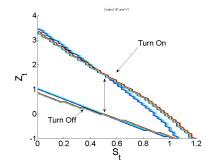
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A simple model

Effects of policy uncertainty

Effects of uncertainty: Makes you switch 'sooner' and pushes value to 'expected' policy level.



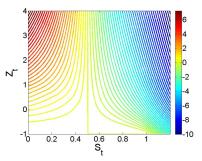


Figure: Switching decisions with and without uncertainty.

Figure: Change in value $V_{risk}(Z_t, s, t) - V_{no risk}(Z_t, S_t, t)$

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Conclusions			
Recap			

We were able to use real options analysis and stochastic control to

- quantitatively determine how price uncertainty affects operating decisions over life of facility (i.e. optimal switching/entry times)
- measurably assess how external policy factors affect these decisions and the economic viability of the facility

Conclusion:

Real options analysis is an insightful tool for studying energy policy.

Thank you!

Ethanol background	Assembling the model	Numerical results	Policy Uncertainty ○○○○○●
Conclusions			
Model Extens	sions:		

ABANDONMENT:

If the option to abandon exists, this limits the potential losses to the salvage value F at abandonment.

$$\max\left[\frac{\partial V_0}{\partial t} + L[V_0] + f_0 - rV_0, \max\left\{(V_1 - D_{01}) - V_0, F\right\}\right] = 0$$

STOCKPILING:

Possibility of stockpiling corn during "low" corn prices. Adds complexity in terms of determining optimal rate to

- purchase (stockpile) excess corn during low prices
- consume corn stockpile during high corn prices

The result is to reduce the need to ever pause production and increases the overall facility value.