



The Value of Climate Information when Farm Programs Matter

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Introduction

- There is a need to estimate value of forecasts
- Agriculture can benefit from forecasts
- Farm decisions include government policies and regulations
- Few studies addressed impacts of Farm Programs to forecasts value (Mjelde et al., 1996; Bosch, 1984)
- Knowledge gap between synergies and conflicts between Farm Programs and forecasts value



jective/Hypothesis



 Estimate impacts of Farm Programs on the value of ENSO forecasts in a maize-cotton-peanut rainfed farm located in Jackson Co., FL

 Government interventions might enhance or limit the usefulness of the climate information



LIMATE CONSORTIUM

Man Representative Farm

- 128.7 ha farm with soils type Dothan Loamy Sand
- Rainfall = 1466 (1143) mm
- T = 19.3 (21.7) °C
- ENSO intra-phase variability impacts crop yields with considerable overlap
- E.g., higher peanut yields early La Niña or late El Niño plantings



SOUTHEAST CLIMOTE CONSODTIUM

IGRICULTURE CLIMATE IIII

Mepresentative Farm

30.774N, 85.226W Representative Farm

Marianna JACKSON

Image © 2005 DigitalGlobe

Pointer 30°47'39.06" N 85°15'35.71" W elev 175 ft

Streaming ||||||||| 100%

|| 100% Eye alt 20747 ft





Crop Yield Simulation

- Chipley weather station (30.783N, 85.483W)- 65 yr records (1939-2003)
- 14 El Niño, 16 La Niña phases
- DSSAT crop simulations (Jones et al., 2003)

 Contemporary and local practices of varieties, fertilization, and planting dates (H.E. Jowers, pers. comm.)



Synthetic Yield Generation

- Needed more ENSO realizations
- Stochastic yield generator (990 yr x ENSO phase)
- Re-sampling technique:
 - Sort simulated yields
 - Function to fit a curve
 - Re-sampling function

Repeated for each planting date management
 each crop, in each ENSO phase REMAND







Synthetic Yields





Synthetic Price Generation

- 2970 price series to match our yields
- Multivariate distribution respecting price covariance among crops based on historical price variability
- Jan96-Jan05 USDA prices, deflated, detrended, decomposed, separated and sampled, back transformed, validated, and seasonally adjusted



Whole Farm Model

- Stochastic non-linear optimization and simulation model
- 325 yr sample for optimizations, all 2970 yr for simulations
- MINOS5 algorithm GAMS (Gill et al. 2000)
- Constant Relative Risk of Aversion (Rr) of 0, 0.5, 1, 2, 3, 4 (Hardaker et al., 2004)



Continities Composed and Compos

$$\max_{x} E\{U(W_{f})\} = \sum_{n=1}^{N} \sum_{i=1}^{3} q_{i}U(W_{0} + \prod_{i,n})/N \quad (1)$$

$$\sum_{m=1}^{22} X_m = 1; X_m \ge 0$$
 (2)

$$\sum_{j=1}^{10} X_m * L_{m,j} \le \overline{L}_j \quad (3)$$

 $U(W_f) = W_f^{1-R_r} / (1-R_r) \quad (4)$



Estimated Value of Information

- Net margins 2970 yr (990 x ENSO phase)
- EVOI = Net Margin With Forecast
 Net Margin Without Forecast
- EVOI = certainty equivalent units (US\$) over different planning horizons
 Repeated for each Rr





Commodity Loan Programs

- Commodity Loan Programs are based on actual production and do not require decision before planting
- The 1996 FAIR Farm Act set LDP of \$1.14 kg⁻¹ for cotton
- The 2002 FSRIA Farm Act set MLB of \$0.39 kg⁻¹ for peanut and \$0.08 kg⁻¹ for maize





Crop Insurance Programs

- Most common crop insurance product by crop in Jackson County (2004)
 - Peanut 70% MPCI, 0.3935 \$ kg⁻¹
 - Cotton 65% CRC, 1.4991 \$ kg⁻¹
 - Maize 50% MPCI, 0.0964 \$ kg⁻¹
- Premiums added to variable costs
- Indemnity payments added to objective function

Findings Optimal Land Allocation

- Forecast value is inherently probabilistic
- Negative value of information exists and is not negligible
- As hypothesized, Farm Programs impact substantially EVOI
- CLP & CIP decrease EVOI
- Further research: synthetic weather generator, other locations: AL, GA

