

Using Public Data to Forecast USDA Farm Subsidies

Derek Araujo • Insight Data Science • Fall 2017

Crop Subsidies: ~ \$25 B annually



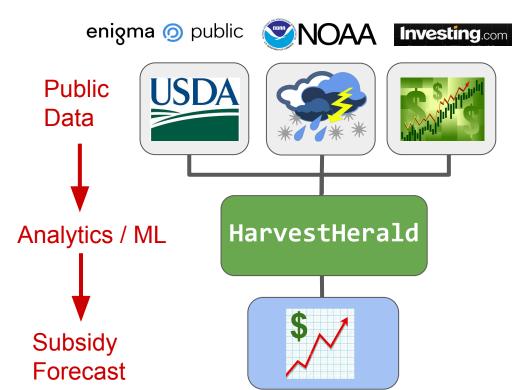
US Gov:

- Budgeting



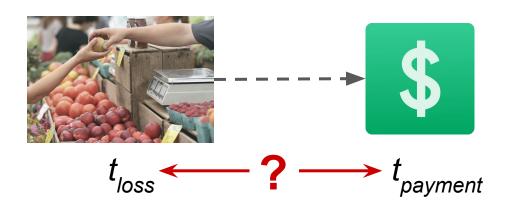
Insurers:

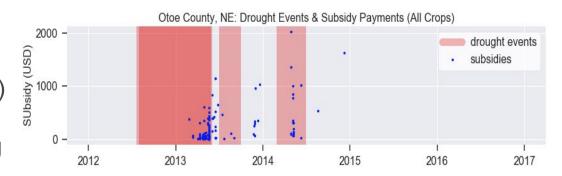
Subsidy = risk mitigation



Data Challenges:

- Unknown time of loss →
 - Unknown sale price, vol.
- Per county: subsidy is sparse/noise-dominated
- → Aggregate for each crop:
 - Monthly bins
 - Avg subsidy (all counties)
 - Focus on drought events (\$\$)
- → Few data points for forecasting

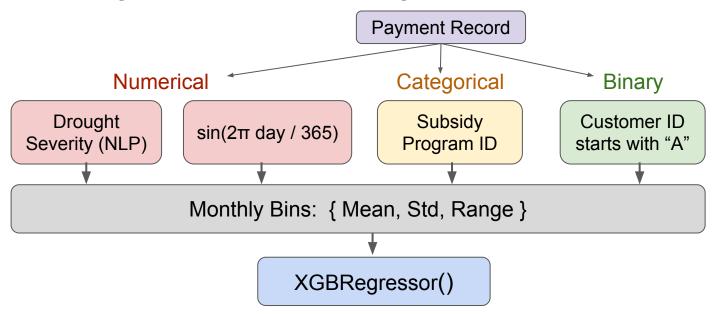




Model-Driven Feature Engineering:

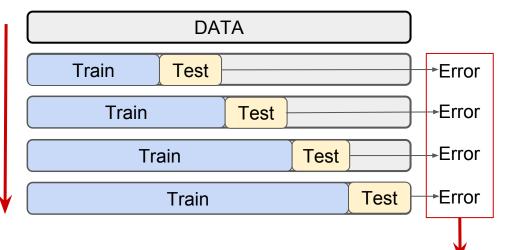
Two models:

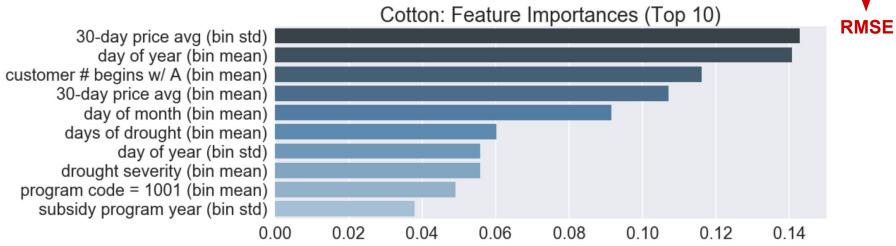
- ARIMA: Auto-Regressive Integrated Moving Average (no engineering)
- XGBoost Regression Tree: > 600 engineered features



Methodology:

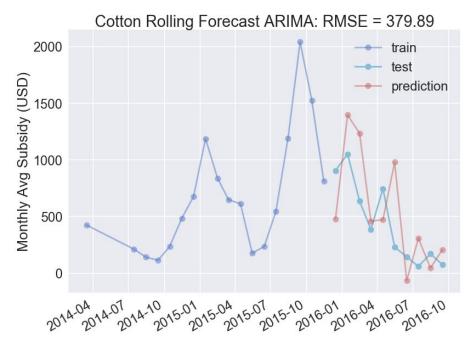
- <u>Target</u>: log(Avg Monthly Subsidy)
- Hyperparams: grid search
- <u>Validation</u>: walk-forward forecast, expanding window
- XGB: examine feature importance



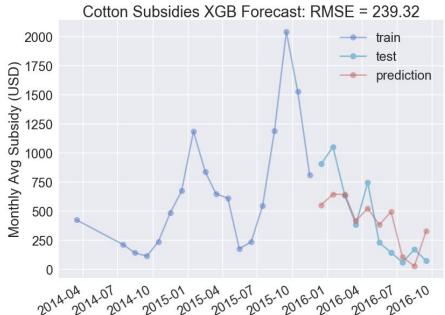


Results: Cotton

ARIMA:

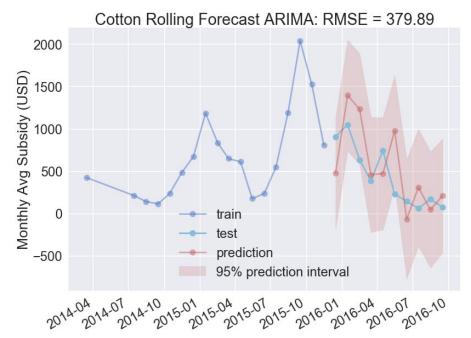


XGB Regressor:

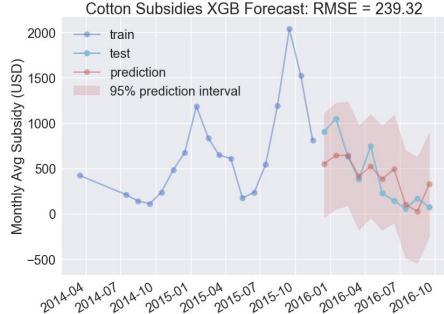


Results: Cotton

ARIMA:



XGB Regressor:

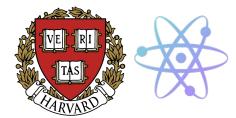


Utility Assessment / Improvements:

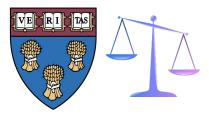
- Wide error bars; but indicative of trend
- <u>Find missing data</u>: (volume, date of sale)
- Other paths forward:
 - ARIMA → ARIMAX (add exogenous variables)
 - Feature selection / PCA



About Derek:



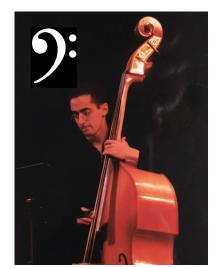
Harvard College A.B., Physics

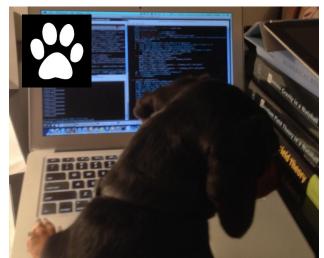


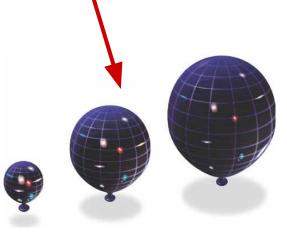
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Inflationary Big Bang Cosmology