

# **Scaling Up Agricultural Insurance**

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## Agriculture is Risky

Long and rich tradition in development economics studying effects of risk, especially in agricultural settings (Paxson 1992; Rosenzweig and Binswanger 1993; Udry 1994, 1995; Mobarak and Rosenzweig 2013; Donovan 2021)

- Risk lowers utility because households prefer stable consumption
- Risk can lower output if high-return investments are risky
- This may explain apparent underutilization of inputs and inefficient crop choice

A potential solution is to offer insurance

- Microeconomic research studying effects of insurance (Karlan et al. 2014)

In parallel, macroeconomic work thinking about GE effects of scaling up interventions

- e.g. Buera et al. (2021), Lagakos et al. (2023), Bergquist et al. (2025); see JEL review by Buera et al. (2023)

## Insurance Unlocks Investment (Karlan et al. 2014)

### Our paper begins with the RCT of Karlan, Osei, Osei-Akoto, and Udry (2014)

- Study rainfall insurance, which pays out if there is not enough rain (or too much)
  - Sidesteps moral hazard/adverse selection: rain is exogenous and common knowledge
- Randomize grants of insurance, and insurance prices
- Insurance raises farmer's agricultural investment
- Insurance shifts crop choice towards risky, high-return crop (maize vs. mangoes)

*Strong evidence that risk aversion plays a role in constraining farmers*

In the background, Ghanaian rainfall insurance market is maturing:

- Experimental product in Years 1-2
- Commercial product in Year 3 and beyond

# Our Paper: Scaling Up Agricultural Insurance

## Should government “scale up” agricultural insurance?

- In practice, this means subsidizing it or offering it at below-market cost

Experiment identifies a partial equilibrium effect:

- Compares treated and control households within the same economy
- If scaled up, there will be important GE effects

To determine optimal policy, we need to understand:

- Is there a market failure?
- *If so, does subsidizing insurance fix the incompleteness?*

**Pecuniary externality in GE: Price changes redistribute resources across states**

## Outline of the Talk

To answer these questions:

- Model: Agricultural households that can (imperfectly) share risk
- Optimal Policy: Subsidy balances “fiscal” vs. “pecuniary” externality
- (Preliminary) Calibration: Use experimental moments, plus additional data
- Counterfactuals: What happens when we subsidize insurance?

This project is work in progress:

- Later, plan to add dynamic model and tighten calibration
- **Your comments and suggestions are much appreciated!**

# Model

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## Environment and Utility

Continuum of ex-ante identical regions,  $i \in [0, 1]$

Representative household in each region, all agents price takers, competitive markets

In each region,  $s_i^H$  is good state (rain) and  $s_i^L$  is bad state (drought)

States are drawn iid across regions, so no aggregate uncertainty

Household utility in region  $i$ :

$$\begin{aligned} U_i &= \mathbb{E}[u(c_i(s_i))] \\ &= P(s_i^H) \cdot u(c_i(s_i^H)) + P(s_i^L) \cdot u(c_i(s_i^L)) \end{aligned}$$

where  $u(c) = \frac{c^{1-\gamma}-1}{1-\gamma}$  is CRRA

## Farm Production

Household makes income by farming;  $G$  different crops, indexed by  $g$

Before knowing the state, allocates unit of land across crops:

$$y_{g,i}(s_i, l_{g,i}) = z_g(s_i) \cdot l_{g,i}^{\theta_g}$$

$$\sum_{g=1}^G l_{g,i} = 1$$

Curvature  $\theta_g$  reflects that plots have heterogeneous suitability

- As more land is allocated to maize, marginal land is less suited to maize

**Crop choice entails risk-return tradeoff:**

- For farmers in Karlan et al. (2014), maize has higher expected return, but mangoes are drought-resistant

## Aggregation and Agricultural Profits

Farmer takes prices as given, agricultural profits are:

$$\pi_i \left( s_i, \{l_{g,i}\}_{g=1}^G \right) = \sum_{g=1}^G p_g \cdot y_{g,i}(s_i, l_{g,i})$$

Agricultural goods are aggregated, CES, into a final good

$$Y_g = \int_0^1 y_{g,i}(s_i) di \quad \forall g$$

$$Y = \left( \sum_{g=1}^G Y_g^{\frac{\sigma-1}{\sigma}} \right)^{\frac{\sigma}{\sigma-1}}$$

Final good is numeraire, so  $P = 1$ . Competitive pricing:

$$p_g = \frac{dY}{dY_g} = \left( \frac{Y_g}{Y} \right)^{-1/\sigma}$$

## Agricultural Insurance

Each household can buy insurance,  $x_i$ :

- Costs  $p_{x,i} \cdot x_i$ , paid regardless of state
- Pays out  $x_i$  in the bad state, 0 in good state
- For simplicity, constrain  $x_i \geq 0$  (household doesn't sell insurance)

## Insurance Firm

Insurance is sold by a representative insurance firm

- Actuarially fair price of insurance is  $P(s_i^L)$
- Firm pays per-unit service cost  $\tau$ , and earns markup  $\mu$

Firm thus offers prices:

$$p_{x,i} = P(s_i^L) + \tau + \mu \forall i$$

and makes profits

$$\Pi' = \int_0^1 (p_{x,i} - P(s_i^L) - \tau) \cdot x_i di = \mu \cdot \int_0^1 x_i di$$

Firm is owned by households, so profits are rebated to households

*Note that  $\tau$  and  $\mu$  are both frictions, but they are distinct:*

- Service cost  $\tau$  is a resource cost; can't just undo it
- Markup  $\mu$  is a pure wedge; can undo it with a subsidy

## Household's Problem

Household budget constraint is:

$$c_i(s_i^H) = \pi_i(s_i^H) + \Pi^I - p_{x,i} \cdot x_i$$

$$c_i(s_i^L) = \pi_i(s_i^L) + \Pi^I - p_{x,i} \cdot x_i + x_i$$

Each household solves:

$$\max_{x, \{l_g\}_{g=1}^G} \mathbb{E}[u(c_i(s_i))]$$

$$s.t. \sum_{g=1}^G l_{g,i} = 1$$

## Separation and Efficiency

Household first order conditions:

$$\text{Production: } \mathbb{E} \left[ \frac{u'(c_i(s_i))}{\mathbb{E}[u'(c_i)]} \cdot p_g \frac{\partial y_{g,i}}{\partial l_{g,i}} \right] = \lambda_i \forall g$$

$$\text{Insurance: } \mathbb{E}[u'(c_i)] \cdot p_{x,i} = u' \left( c_i \left( s_i^L \right) \right) P \left( s_i^L \right)$$

*As long as  $x_i > 0$ , we get a nice separation result:*

$$\mathbb{E} \left[ \frac{u'(c_i(s_i))}{\mathbb{E}[u'(c_i)]} \cdot p_g \frac{\partial y_{g,i}}{\partial l_{g,i}} \right] = p_{x,i} \cdot p_g \frac{\partial y_{g,i}(s_i^L)}{\partial l_{g,i}} + (1 - p_{x,i}) \cdot p_g \frac{\partial y_{g,i}(s_i^H)}{\partial l_{g,i}}$$

- Farm maximizes marginal-utility-weighted profits
- With insurance, weight by state-prices,  $p_{x,i}$  and  $1 - p_{x,i}$
- With actuarially fair insurance, maximize (true) expected profits

**Distortion in insurance market leads to production inefficiency**

## **Welfare and Optimal Policy**

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## Subsidies as Markups

To scale up agricultural insurance, government provides a subsidy

Subsidy lowers the price consumer pays, and is funded through lump-sum taxation

Markups work in exactly the same way!

- Markup raises price, and profits returned lump-sum to households as dividend

For clarity, we will lump them together:

- Using subsidies as a tool, government chooses the markup

$$\mu = \mu^{\text{Baseline}} - \text{Insurance Subsidy}$$

## Indirect Utility and Consumption

How do subsidies affect household welfare, in general equilibrium?

- Differentiate welfare with respect to the markup
- Household treats prices and insurer profits as exogenous
- Land and insurance are choices, so fall out with Envelope Theorem

$$\begin{aligned}\frac{dU}{d\mu} &= \frac{\partial U}{\partial p_x} \cdot \frac{dp_x}{d\mu} + \frac{\partial U}{\partial \Pi^I} \cdot \frac{d\Pi^I}{d\mu} + \sum_{g=1}^G \frac{\partial U}{\partial p_g} \cdot \frac{dp_g}{d\mu} \\ &= \mathbb{E} \left[ u'(c) \cdot \left( \underbrace{\frac{\partial c}{\partial p_x} \cdot \frac{dp_x}{d\mu}}_{\text{Cost of Insurance}} + \underbrace{\frac{\partial c}{\partial \Pi^I} \cdot \frac{d\Pi^I}{d\mu}}_{\text{Insurer Profits}} + \underbrace{\sum_{g=1}^G \frac{\partial c}{\partial p_g} \cdot \frac{dp_g}{d\mu}}_{\text{Farmer Profits}} \right) \right]\end{aligned}$$

Note:  $\frac{d}{d\mu}$  reflects the derivative of a variable in general equilibrium

## Optimal Subsidies

At the optimal subsidy:

$$\underbrace{-\mu \cdot \frac{dx}{d\mu}}_{\text{Fiscal Externality}} = \mathbb{E} \left[ \underbrace{\frac{u'(c)}{\mathbb{E}[u'(c)]} \cdot \left( \sum_{g=1}^G y_g \cdot \frac{dp_g}{d\mu} \right)}_{\text{Pecuniary Externality/Risk-Sharing}} \right]$$

- When  $\mu = 0$ , marginal benefit to consumers from lower  $p_x$  exactly cancels marginal cost to the insurer/government
- When  $\mu < 0$ , subsidy imposes a fiscal externality
- Pecuniary externality: change in crop prices transfers resources across states
- Formula very similar to Baily-Chetty: balance fiscal externality against risk term

## Pecuniary Externality Reduces Consumption Risk

$$\underbrace{-\mu \cdot \frac{dx}{d\mu}}_{\text{Fiscal Externality}} = \mathbb{E} \underbrace{\left[ \frac{u'(c)}{\mathbb{E}[u'(c)]} \cdot \left( \sum_{g=1}^G y_g \cdot \frac{dp_g}{d\mu} \right) \right]}_{\text{Pecuniary Externality/Risk-Sharing}}$$

### Why does pecuniary externality reduce consumption risk?

- Under incomplete insurance, marginal utility higher in the bad state
- Insurance subsidy  $\Rightarrow$  more production of risky crop
- $\Rightarrow$  lower relative price of risky crop  $\Rightarrow$  transfers resources to bad state

Note that  $\mathbb{E} \left[ \sum_{g=1}^G y_g \cdot \frac{dp_g}{d\mu} \right] = 0$ , so no effect on average income

- Pecuniary externality is just about transfers across states through relative prices

## Pecuniary Externalities Under Incomplete Markets

### When and why are pecuniary externalities non-zero?

Can rewrite risk-sharing/pecuniary externality term:

$$\mathbb{E} \left[ \frac{u'(c)}{\mathbb{E}[u'(c)]} \cdot \left( \sum_{g=1}^G y_g \cdot \frac{dp_g}{d\mu} \right) \right] = (\mu + \tau) \cdot \sum_{g=1}^G \left( y_g(s^L) - y_g(s^H) \right) \cdot \frac{dp_g}{d\mu}$$

- Holds for arbitrary production functions and (homothetic) crop demand
- Under full insurance,  $\mu + \tau = 0$ , the pecuniary externality is zero
- When markets are competitive ( $\mu = 0$ ) and complete ( $\tau = 0$ ), both fiscal and pecuniary externality are zero

### Special case of more general principle:

- First Welfare Theorem: Competitive and complete markets achieve the first-best  
 $\implies$  pecuniary externalities cancel out
- Theory of the Second Best: Away from first-best, they no longer cancel out!

## The Optimal Subsidy Implements A Negative Markup

$$\underbrace{-\mu \cdot \frac{dx}{d\mu}}_{\text{Fiscal Externality}} = \mathbb{E} \left[ \underbrace{\frac{u'(c)}{\mathbb{E}[u'(c)]} \cdot \left( \sum_{g=1}^G y_g \cdot \frac{dp_g}{d\mu} \right)}_{\text{Pecuniary Externality/Risk-Sharing}} \right]$$

### What markup, $\mu$ , will solve the above FOC?

- If  $\tau > 0$ , then pecuniary externality of a subsidy will be beneficial as long as insurance is less than actuarially fair
- At  $\mu = 0$ , fiscal externality is zero; at  $\mu + \tau = 0$  the pecuniary externality is zero
- So, optimal  $\mu < 0$  (subsidy larger than markup), but still not full insurance

*Result highlights importance of macroeconomic analysis for optimal policy*

- Without pecuniary externality: just set subsidy to undo markup
- The interesting term is inherently a general equilibrium object
- Micro RCTs inform calibration, but cannot directly tell us optimal subsidy

## **Calibration**

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## Calibration Strategy

Calibrate to agricultural economy in Ghana

### Three sources for calibration:

- Experimental moments from Karlan et al. (2014)
- Supplemental moments from Ghana Living Standards Survey (GLSS)
- External calibration to parameters from Sotelo (2020)

Model is just-identified; search for parameters that exactly hit target moments

- In principle, joint identification from all moments
- But, highlight which moments most closely linked to each parameter

## Identification of Risk Aversion ( $\gamma$ )

Karlan et al. (2014) offer farmers insurance at (randomized) prices, to estimate insurance demand elasticity

- Estimate demand elasticity of 2.0 between market price and actuarially fair price

We use this to identify  $\gamma$

- If farmer is nearly risk-neutral, demand will quickly drop to zero away from actuarially fair price
- If farmer is very risk averse, demand will be less elastic

This is a partial equilibrium elasticity

- To obtain moment from model, first solve in GE
- Then vary  $p_{x,i}$ , while holding fixed  $\Pi'$  and  $\{p_g\}_{g=1}^G$

*Not quite same as  $\frac{dx}{d\mu}$  in optimal subsidy formula (PE vs. GE), but very closely related!*

## Identifying Production Parameters

For each crop, need to identify three production function parameters:

- Curvature (returns-to-scale) parameter  $\theta_g$
- Productivity,  $z_g(s_i)$ , in good state and bad state

We identify these using data on land use and yields

First, suppose you know  $\theta_g$ :

$$z_g(s_i) = y_{g,i}(s_i) / l_{g,i}^{\theta_g}$$

Need data on land shares for each crop, and (state-specific) output

*Subtlety: We measure revenue, rather than output.*

- Back out  $p_g$  from  $\mathbb{E}[p_g \cdot y_g]$ , then can back out  $y_g(s_i)$
- Normalization: Demand has no crop-specific multipliers, so need to infer “units”

## Identifying Production Parameters ( $\theta_g$ )

How do we identify curvature parameter  $\theta_g$ ?

Risk-neutral farmers would equalize marginal revenue product of land across crops

$$\begin{aligned}\frac{d}{dl_g} \mathbb{E}[p_g y_{g,i}] &= \theta_g \cdot p_g \cdot \mathbb{E}\left[z_g \cdot l_{g,i}^{\theta_g - 1}\right] \\ &= \theta_g \cdot \underbrace{\frac{p_g \cdot \mathbb{E}[y_{g,i}]}{l_g}}_{\text{Average Yield for Crop } g}\end{aligned}$$

So, ratio of average yields gives ratio of  $\theta_g$  parameters

- Pin down  $\theta_{\text{Maize}} = 0.4$  from Sotelo (2020)

*With risk aversion,  $\mathbb{E}[\cdot]$  replaced with risk-adjusted probabilities*

- If demand for insurance is positive, this is pinned down by  $p_x$

# Ghana Living Standards Survey (GLSS) Data

## Pin down production parameters with moments from GLSS

Data on revenue and land allocation by crop

- Revenue data available from rounds 4-7 (1998-1999, 2005-2006, 2012-2013, 2016-2017)
- Land data is only available for a subset of crops
- For round 7, we are able to impute area harvested; verify accuracy for the crops where we observe land directly

We focus on the top ten crops by area harvested, and then drop cocoa

- Maize, sorghum, groundnut (peanut), beans, rice, cassava, plantain, yam, cocoyam (taro)

## Rainfall Data

Daily rainfall data from Google Earth Engine (UC Santa Barbara CHIRPS data):

- Data from 1981-2025, with 0.5 degree ( $\approx 5.5$  km) pixel resolution
- For each pixel, compute “Bad Weather” dummy for a year  
Following exact definition used in Karlan et al.’s Year 2 insurance product
  - Growing season from June-September
  - Day is “dry” if less than 1mm rain, and “wet” if more than 1mm
  - Bad weather trigger if 12 or more consecutive dry days and/or 7 or more consecutive wet days

**Take average to get  $P(s^L) = 0.68$**

## Productivity Regressions

To back out average yields, we run (in round 7):

$$\log(\text{Yield}_{irg}) = \alpha_r + \delta_g + \varepsilon_{irg}$$

To back out how yields depend on weather, we regress on Bad Weather<sub>irt</sub> (rounds 4-7)

- Bad Weather<sub>irt</sub> variable is average across all pixels in that region

For crops where we can observe yields:

$$\log(\text{Yield}_{irgt}) = \beta_g \cdot \text{Bad Weather}_{irt} + \gamma_{rg} + \theta_{gt} + \varepsilon_{irgt}$$

For crops where we cannot observe yields:

$$\log(\text{Harvest Value}_{irgt}) = \beta_g \cdot \text{Bad Weather}_{irt} + \gamma_{rg} + \theta_{gt} + \varepsilon_{irgt}$$

where  $i$  is a household,  $r$  is a region,  $g$  is a crop,  $t$  is a year

## Insurance Servicing Costs ( $\tau$ ) and Markups ( $\mu$ )

The supply side of insurance depends on three terms:

- Probability of bad state,  $P(s^L)$ : Actuarially fair price of insurance
- Servicing cost,  $\tau$ : Additional resource cost of providing insurance
- Markup,  $\mu$ : Pure markup between price and resource cost of insurance
- $p_x = P(s^L) + \tau + \mu$

We calibrate  $\tau + \mu$  using the actuarially fair price and estimated market price of the Year 2 insurance product in Karlan et al. (2014):

- $$\frac{\text{Market Price}}{\text{Actuarially Fair Price}} = \frac{P(s^L) + \tau + \mu}{P(s^L)} = \frac{14}{10} \implies \frac{\tau + \mu}{P(s^L)} = \left(\frac{14}{10} - 1\right) = 0.4$$
- We will calibrate  $\mu^{\text{Baseline}} = 0$ , so  $\tau = 0.4 \times P(s^L)$

## Elasticity of Substitution Across Crops $\sigma$

For elasticity of substitution across crops, we use estimate from Sotelo (2020)

- Estimate  $\sigma = 2.4$  in Peruvian data

Elasticity of substitution across crops is important for the pecuniary externality

- Insurance subsidy will increase quantity of risky crops
- How that translates into price effect,  $\frac{dp_g}{d\mu}$ , depends on  $\sigma$
- If  $\sigma$  is very high (near-perfect substitutes), then price effect is small

Estimate of  $\sigma = 2.4$  is on the lower end of literature

- Gives the pecuniary externality the best chance to matter

## Calibration Table

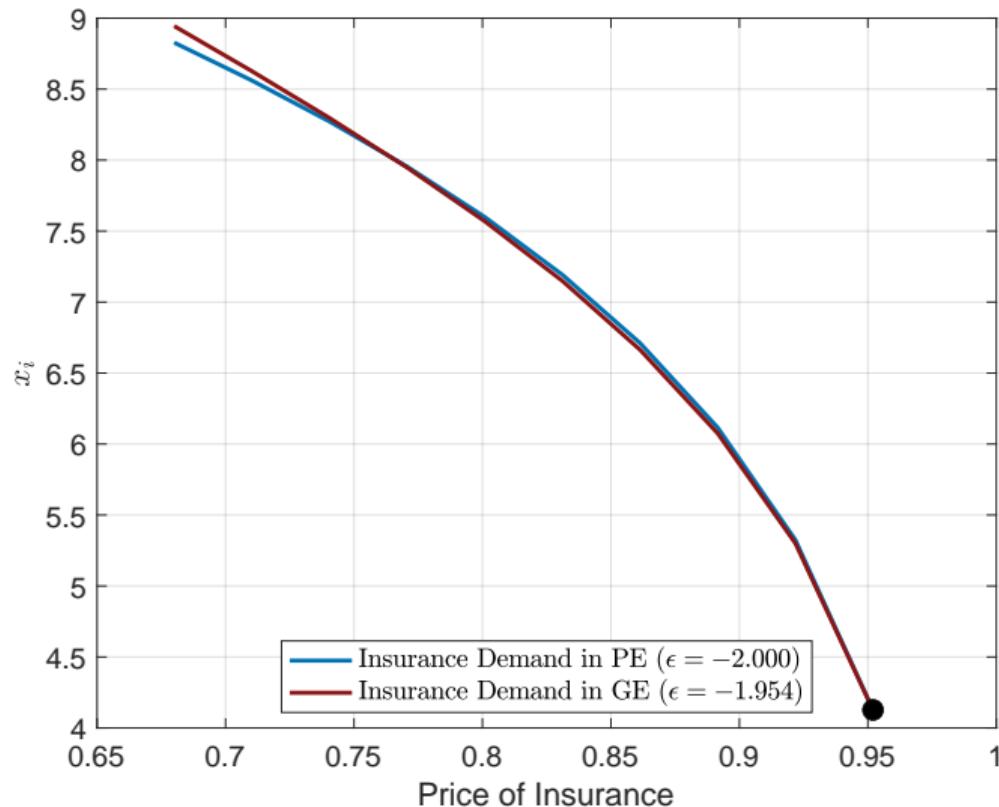
Parameter	Description	Value	Calibration Target
$\sigma$	Elasticity of Substitution Across Crops	2.4	Sotelo (2020)
$\gamma$	Coefficient of Relative Risk Aversion	7.4	Elasticity of Demand for Insurance
$P(s^L)$	Probability of Low Rainfall	0.68	Rainfall Data
$\{\theta_g\}_{g=1}^G$	Curvature Parameter by Crop	—	Relative Yields Across Crops
$\{z_g(s_i)\}_{g=1}^G$	State-Dependent Productivity by Crop	—	Yield Regressions and Land Shares
$\tau$	Servicing Cost of Insurance	0.4	Computed in Karlan et al. (2014)
$\mu^{\text{Baseline}}$	Markup on Insurance	0	By Assumption

**Table 1:** Calibration Table for Model Parameters

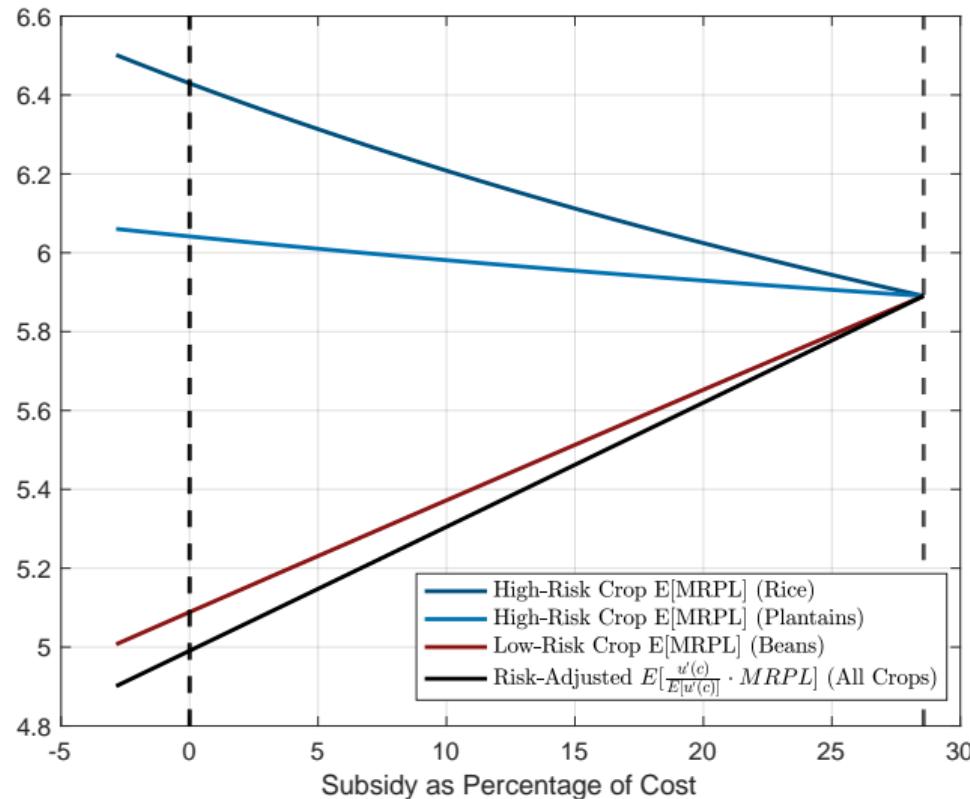
## **Results and Counterfactuals**

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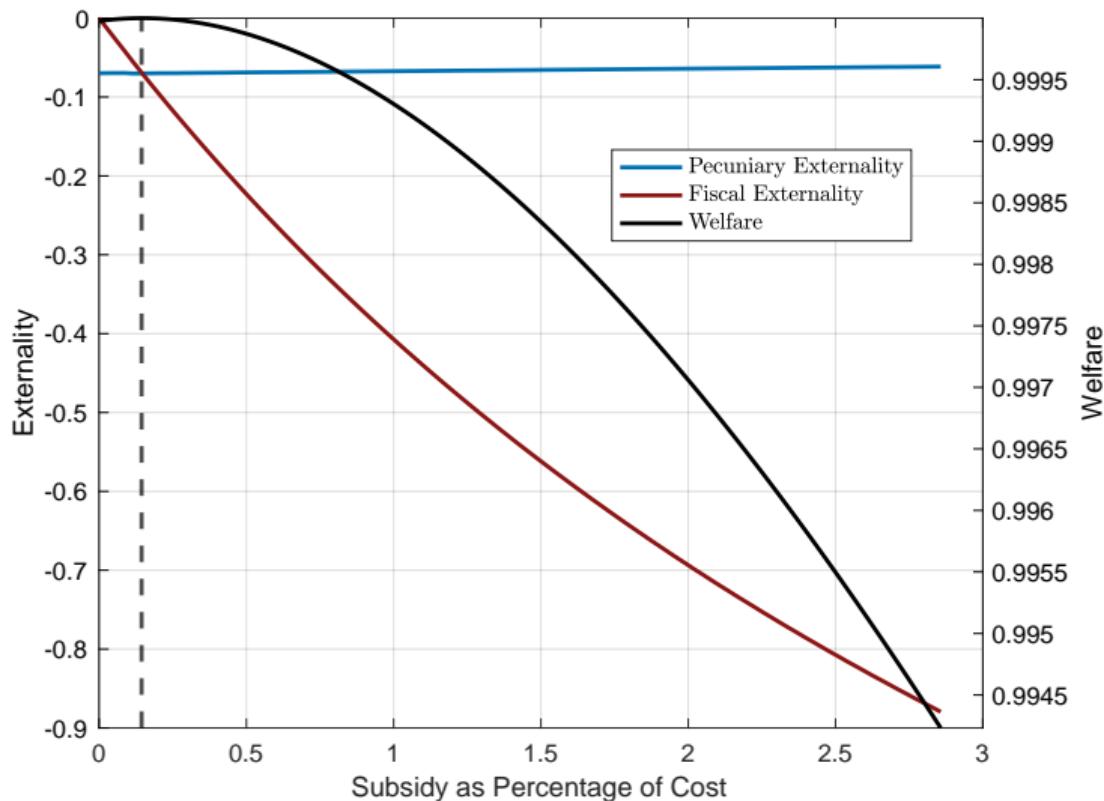
## Insurance Demand (PE vs. GE)



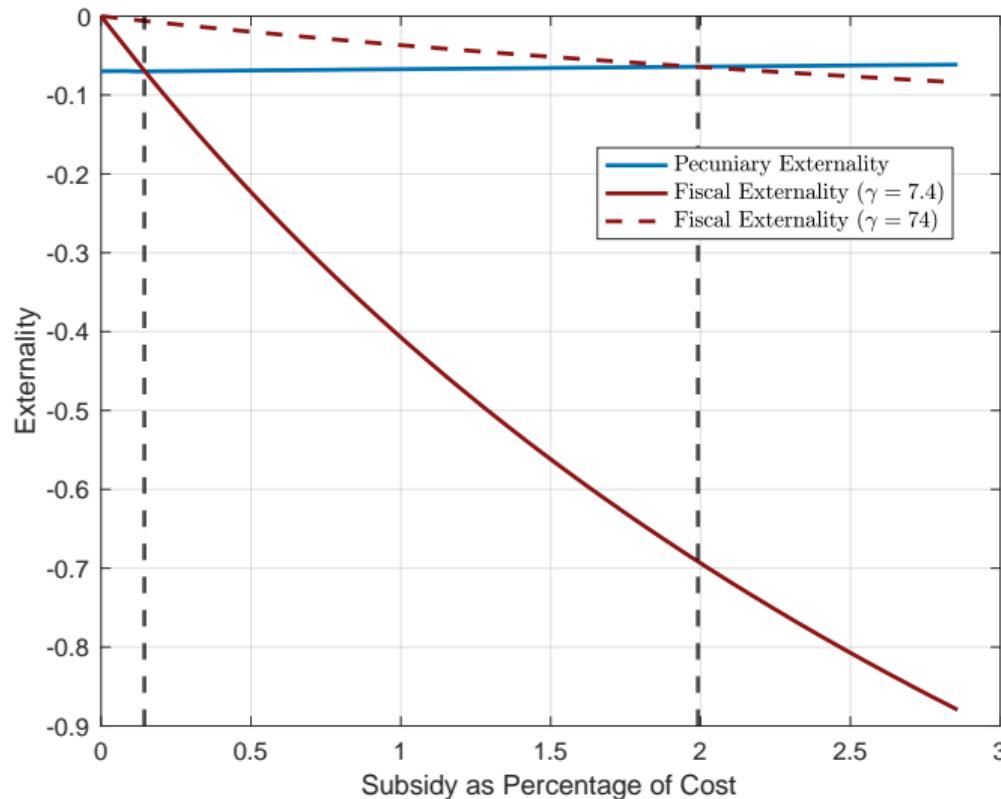
## Marginal Revenue Product of Land



## Optimal Subsidy



## The Optimal Increases with Risk Aversion



## Conclusion

The optimal insurance subsidy solves the following equation:

$$\underbrace{-\mu \cdot \frac{dx}{d\mu}}_{\text{Fiscal Externality}} = \mathbb{E} \left[ \underbrace{\frac{u'(c)}{\mathbb{E}[u'(c)]} \cdot \left( \sum_{g=1}^G y_g \cdot \frac{dp_g}{d\mu} \right)}_{\text{Pecuniary Externality/Risk-Sharing}} \right]$$

**Optimal subsidy undoes the markup, plus a little more:**

- Macroeconomic modeling is essential to measuring the pecuniary externality
- Microeconomic experiments can help calibrate the model
- But, just the fact that farmers respond to insurance does not tell us that we should subsidize it

Quantitatively small: If  $\mu^{\text{Baseline}} = 0$ , optimal subsidy is 0.1% of cost

**Thank you for your comments and suggestions!**

## **Appendix**

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## Imputation of Land in GLSS

GLSS divides crops into two groups, based on typical harvest frequency

- For frequently harvested crops, does not ask household for total acreage
- But, we have total harvest value for all crops

In round 7, we have area of plot, and all crops harvested on it (may be multiple)

Imputation procedure:

- Trim harvest value by crop (1st and 99th percentile)
- For each region-crop, compute yield among single-cropped plots
- Divide harvest value by relevant imputed yield to get acreage

Confirm accuracy for the crops where we observe actual acreage [Back](#)

## Data Cleaning for GLSS Regressions

For productivity regressions:

- Trim harvest value by crop-year at 1st and 99th percentile
- Compute yields (actual or imputed acreage)
- Then trim yields by crop-year at 1st and 99th percentile

For crop fixed effect regressions use round 7 yields

For rainfall regressions use rounds 4-7

- Use yields for crops where we can compute them
- Harvest value is the outcome for other crops

## GLSS Estimates

Crop	Acreage	Imputed Acreage	$\delta_g$	$\beta_g$
Beans	2,815.74	3,093.89	0.00	0.23
Cassava		4,907.58	0.51	-0.66
Cocoyam		2,557.52	0.06	-0.65
Groundnut	4,031.86	4,916.80	0.56	-0.51
Maize	11,053.42	11,691.24	0.37	-0.21
Plantain		2,960.37	0.55	-0.59
Rice	2,061.25	3,037.53	0.69	-0.37
Sorghum	5,542.16	4,184.56	-0.04	-0.21
Yam		2,889.68	1.66	-0.53

**Table 2:** Estimated Moments from Ghana Living Standards Survey

## Calibration of Productivity Parameters (Details)

Calibrate model to match:

- $\frac{p_g \cdot \mathbb{E}[y_{g,i}]}{l_g} = \exp(\delta_g)$
- $\frac{p_g \cdot y_{g,i}(s_i^L)}{l_g} = \exp(\beta_g) \cdot \frac{p_g \cdot y_{g,i}(s_i^H)}{l_g}$
- $\mathbb{E}[l_g]$  from land shares in round 7

Separation means we can solve for production decisions without solving household problem

- Solve for  $p_x = P(s^L) + \tau$ , since  $\mu^{\text{Baseline}} = 0$
- Solve land FOCs to get  $\theta_g$  from  $\delta_g$  and  $\beta_g$
- Use  $\sigma$  to get  $p_g$  from  $\delta_g$
- Solve for state-specific productivities
- Everything is closed-form :-)