Dynamic Field Experiments & Index Insurance Comprehension: Risk valuation in Morocco, Kenya & Peru



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Index Insurance Benefit Chain

Product Design

Reduced Covariate Risk

The "build it and they will come" short cut

Uptake

Livelihood Benefit

Reduced Vulnerability

& Poverty









Intro

Overview

OBJECTIVE

Explore in detail the use of economic experiments:

- To build comprehension among prospective beneficiaries
- To better understand their valuation of risk and risk reduction
- In order to improve the design and delivery of index insurance

OUTLINE

- **Risky Benefits**: Discuss stochastic relative benefit streams, their dynamic implications, and the comprehension and valuation challenges they can pose to the rural poor
- **Experiments**: Introduce economic experiments as a means to assess risk valuation, including those with built-in dynamics
- The Paper: Describe and analyze dynamic risk experiments conducted in *Morocco, Kenya* and *Peru*
- **Discussion**: Offer concluding thoughts about the merits and limitations of economic experiments in the 'index insurance benefit chain'

TAKE HOME: Economic experiments can be an effective tool for building comprehension of index insurance, assessing risk preferences and potential demand

Risky Benefits in India

- Bt cotton provides important 'higher moment benefits' that are quite different than classic first moment yield improvements
- Many other crop traits in the pipeline confer similar higher moment benefits
- Will poor farmers value propoor seeds that reduce production risk? Implications for delivery and uptake?





Risky Benefits Experiments Kenya Peru Stochastic Relative Benefits & Welfare Dynamics

Morocco

Discussion



Drought Tolerance

Intro

Dynamic Benefits: Safety & Cargo Nets

- With asset dynamics, targeting social protection can be key (Barrett et al. 2008)
 - Safety nets can protect the poor from falling below dynamic asset thresholds
 - **Cargo nets** may be needed to lift the persistently poor above dynamic asset thresholds
 - The dynamic efficacy of transfers can be heavily conditioned on the recipient's proximity to critical thresholds

Taxonomy of Economic Experiments

Ē	Туре	Context	Setting	Control	
onomics	Conventional Lab Experiment	Abstract	Lab	Direct	
ental Eco	Artefactual Field Experiment		Field Lab		
xperim	Framed Field Experiment	Familiar			
] پ ا	Natural Field Experiment (RCT)		Natural (subjects	Indirect (via third party)	
_	Natural Experiment		participation)	None	

Experimental Economics in the Field

Used to simulate 'incentive compatible' economic behavior in a controlled and relevant environment

Guiding Principles

- 1. Real incentives
- 2. No deception
- 3. Framing & context in field experiments
- 4. Repetition to allow learning

Field Experiments in Development

- Binswanger's (1980) coin toss risk experiments
- Pender's (1996) delayed rice gifts and discount rates
- In 2000s:
 - Standard lab experiments taken to the field: Risk (EU v CPT), public goods, norms of fairness and punishment, etc.
 - Framed field experiments: Microfinance, technology adoption, etc.

Dynamic experiments

- Explicit linkages across experimental rounds
 - Cumulative earnings in an account Ο
 - Reputation building across rounds Ο
- A dynamic threshold changes key payoff parameters at a known point in endowment or earnings space
- E.g., Microfinance field experiments:
 - Future loans conditioned on past repayment \rightarrow Dynamic Ο incentives
 - These may matter more than group loans, monitoring, etc. Ο (Abbink et al. 2006, Gine et al. 2009)
 - Individuals' response to these dynamic incentives is positively Ο correlated with their 'static' risk preferences (Gine et al. 2009)

Why Index Insurance Experiments?





- So prospective beneficiaries can better understand products with stochastic and/or dynamic benefits
- So we can understand their valuation and potential demand and refine product design and delivery accordingly
- If we think index insurance can provide dynamic benefits, we should consider valuation in a dynamic settings

Objectives by Location

	Project Objective	Team
Morocco	Assess drought risk & valuation of drought tolerance	UCD: T.Lybbert, Y.Kusunose, N.Magnan, J.E.Taylor INRA: A.Fadlaoui, R.Mrabet ICARDA: A.Aw-Hassan CIMMYT: E.Meng
Kenya	Assess feasibility & valuation of NDVI index insurance	Cornell: C.Barrett, P.Chantarat Syracuse: J.McPeak ILRI: A.Mude Wisc: M.Carter
Peru	Assess feasibility & valuation of area yield index insurance	UCD : S.Boucher, C.Mullally Wisc : M.Carter, F.Galarza





Obj: Assess drought risk at HH level and valuation of drought tolerant crops



- 20 villages in rainfed cereal production
- Initial sample size 290 HHs

Summer 07	Summer 08	Season 08–09
Village survey	Village survey	SMS survey (2/mo.)
Detailed HH survey	Detailed HH survey	
Economic experiment		

Experimental Design

- Valuation of payoff distributions, i.e., 'seed types'
 - Open-ended valuation via BDM
 - o Dichotomous 'seed' choice
- Static, then dynamic rounds with cumulative earnings and plot thresholds at 0dh and 140dh



Context, Calibration & Comprehension

- Contextualizing risk
- Calibrating payoffs
- A familiar context with well-calibrated payoffs can improve comprehension
- But presentation and practice are still critical



Morocco

Kenya

Peru Discussion

Conducting the Experiment



Practice then high-stakes rounds WTP for each seed in isolation Choice between seeds (static) Choice between seeds (dynamic)





Dynamic Treatment Effects

- We estimate an ordered Probit model of seed choice in dynamic rounds
- Key results
 - Evidence of dynamic risk response both above 0dh (greater risk aversion) and below 140dh thresholds (greater risk seeking)
 - Risk taking with second plot
 - Proximity interacted with static risk aversion → dynamic risk seeking just below threshold is magnified by static risk aversion

Morocco

Kenya Peru -

Discussion

On Location: Kenya



NDVI-based index insurance for Kenyan pastoralists





- Starting period herd sizes randomly assigned: 6, 8, 10 Tropical Livestock Unit
- Shared risk determined by a ball drawn from a bag with 16 balls, each ball is a rainy season – dry season pair:
- This shared risk is adjusted for each person based on the individual risk draw

-30%	-20%	0%	10%	20%
(1/16)	(1/16)	(2/16)	(7/16)	(5/16)
$\overline{\mathbf{S}}$				
Ball-10%		I+0%	Ball +10	

Experimental Design

- Five sets of rounds
 - 1. Covariate risk only ("pasture quality")
 - 2. Individual risk only ("luck")

Covariate and individual risk

- 3. Without insurance
- 4. With insurance
- 5. Covariate and individual risk with <u>chosen level of insurance</u>
- Subjects paid according to end-of-round herd size in randomly drawn round from set 5

Kenya

Actual Asset Dynamics

- With nonlinear asset dynamics, insurance may be especially valuable to the poor
- Payoffs in experiments can be linked across rounds in a way that captures the essence of such dynamics
- This can facilitate comprehension and better assess risk attitudes and ultimately demand
- There exists a dynamic threshold at ~7 TLU
- Herders are aware of the threshold
- To replicate these dynamics impose A fixed consumption requirement of 0.5 TLU per round An appropriately calibrated herd growth rate Random starting herd sizes

Experimental Asset Dynamics

- To replicate these dynamics in the experiment impose
 - A fixed consumption requirement of 0.5 TLU (5 sheep or goats) per round
 - An appropriately calibrated herd growth rate
 - Random starting herd sizes
- → E(herd growth)<0 below 6.6 TLU E(herd growth)>0 above 6.6 TLU

Morocco

Kenya

Conducting the Experiment



5 Locations207 participants





 During the experiment, the subsistence requirement seemed to capture a mechanism behind non-linear herd dynamics that was familiar to the subjects
 Some pleaded to be able to 'go hungry' in a given round in order to keep their herd above ~7 TLU



Mean herd size 'with' greater		Mean variance in herd size across rounds	Mean share of herd insured	t-statistic on test of different means [*]	
	than 'without'	'with' versus 'without'	across rounds	6 – X	8 - 10
6 TLU start	+3%	-40%	81%	-	-
8 TLU start	+2%	-29%	78%	2.40	-
10 TLU start	+3%	-22%	68%	8.47	6.45





Discussion

Peru

Determinants of Insuring

	Coefficients	Std Errors
Round number (2,3,,10)	0.012***	0.0046
Dirib Gumbo dummy	0.11***	0.0378
Karare dummy	0.13***	0.0346
Kargi dummy	0.14***	0.0354
North Horr dummy	0.09**	0.0365
Start 6 TLU dummy	1.25***	0.0771
Start 8 TLU dummy	1.38***	0.0938
Start 10 TLU dummy	1.39***	0.1022
Beginning round herd size	-0.0910***	0.0139
Beginning round herd size ²	0.0020***	0.0005
Covariate shock previous round	-0.0013*	0.0009
Idiosyncratic shock previous round	-0.0028**	0.0015
\mathbf{R}^2 :	0.47	
N (207 subjects x 9 rounds each):	1863	

Table 5 Tobit estimation results for fraction of herd insured in Kenya

Logologo is omitted site dummy. All three start sizes included instead of an overall constant. Estimated as double censored tobit model (45% (2%) of observations censored at 100% (0%) of herd insured). Subject fixed effects are not included so that start sizes can be.

*** (**) [*] denote significance at 5% (10%) [15%] levels.

On Location: Peru

- Area yield index insurance project
- 25,000 irrigated hectares
- Smallholder cotton dominates
- Natural risks
 - o Drought
 - Excess rain (El Niño)
 - Temperature/pests



Discussion

Experimental Design (Baseline)

Project A: Uninsured Loan

- Intensive cotton technology with loan
- High return/high risk option

Project B: Self-Finance

- Low-intensity, cotton technology without loan
- Low return/low risk option
- Farmer's Payoff in each round depends on
 - Project chosen
 - Valley-wide weather shock and Individual shock
 - Credit history (defaults reduce land value by 50%)
- 6 low-stakes rounds (learning), 6 high-stakes rounds

Experimental Design (Insurance)

- Project A: Uninsured Loan
- Project B: Self-Finance
- Project C: Insured Loan
 - Intensive cotton technology with loan
 - Area-yield insurance
- Farmer's Payoff in each round depends on
 - Project chosen
 - Covariate and individual weather shock
 - Credit history
- 6 low-stakes rounds (learning), 6 high-stakes rounds

Calibration & Context in Peru

Color de la Ficha	Rendimiento Promedio	Numero de Fichas
Negro	Muy Bajo (23 QQ)	1
Rojo	Bajo (30 QQ)	2
Blanco	Normal (37 QQ)	4 🔅 🔅
Azul	Alto (43 QQ)	2
Verde	Muy Alto (48 QQ)	1

Fichas del Rendimiento Promedio en el Valle

Color de la Pelota	Suerte Individual	Numero de Pelotas
Morado	Mala	1
Verde	Normal	2
Amarillo	Buena	1 🙂



Pelotas de la Suerte Individual

- Morocco

C Julio Cesar Casma

Kenya

Peru 📙 Dis

Discussion

Calibration & Context in Peru







Morocco

Kenya

Peru Discussion

Conducting the Experiment



12 Locations 412 Subjects



Dynamic Treatment Effects

 Insurance can resolve risk-rationing, especially among relatively educated subjects

		Insurance Game				
		Uninsured loan (A)	Self-finance (B)	Insured loan (C)	Total	%
ame	Uninsured loan (A) %	109 <i>38.0</i>	14 4.9	164 57.0	287 100.0	75.9
ine G	Self-finance (B) %	20 22.0	19 20.9	52 57.1	91 100.0	24.1
Basel	Total %	129 <i>34.1</i>	33 8.7	216 57.1	378 100.0	100.0

Intro	Risky Benef	its Experiments	- Morocco Ken	ya Peru Discussion
		MOROCCO	PERU	KENYA
Dynamie element experim	c Subject of the thr ent consect cumula lose th season plot) if earning (above	ts choose between ee 'seeds' for seven putive rounds with ative earnings. They eir plot for one (gain a second their cumulative gs are below 0Dh 140Dh).	Subjects with uninsured loan face dynamic risk of default, which eliminated their access to credit in future rounds and depreciated the value of their land.	Subjects required to consume 0.5 livestock units each round, which creates positive (negative) expected herd growth above (below) 6.6 livestock units.
Effect of dynamic element	Farme jus thr agg the Statica far agg 14 Farme wit	rs are conservative t above the 0Dh eshold and gressive just below 140Dh. Ily risk averse mers are especially gressive just below 0Dh. rs take greater risks th the second plot ce they have it.	 In pre-testing, losing land as default consequence too dominant as a dynamic incentive. 57% of risk rationed farmers opt for insured loan when available. Statically risk averse farmers tend to stick with self-financing. 	 Herders clearly understood the nonlinearity introduced by the consumption requirement. Mean share of herd insured higher for those starting below 6.6 threshold. Share of herd insured increases with initial herd size but decreases as herds grow, which requires linked rounds.



- With repeated seasons, subjects can appreciate stochastic and dynamic benefits
- In all three experiments, subjects' risk decisions respond to dynamics
- Connection to products and place provides a useful context to aid comprehension
- But complex experimental designs can still pose comprehension problems

External Validity of Dynamics?

- Within-subject design with both static and dynamic treatments imply that the dynamic treatment effect may be of interest even without strong external validity
- When building comprehension for a soon-to-bereleased insurance product, external validity matters more...

...but strong context with specific product strengthens context-specific external validity

E.g., ILRI getting calls from pastoralists wondering when NDVI insurance will be available

 External validity of experiments that elicit responses to broader, more general interventions conditional on poverty dynamics may be more challenging



- Computer-based platforms promising due to computational demands of dynamic experiments (especially when combined with other logistical challenges of conducting field experiments
- These can improve subject learning / comprehension
- Kenya part II
 - Local wifi linked netbooks
 - GameWeb software
 - Browser interface
 - Display of outcomes
 - Near real time feedback



Final Thoughts

- Economic experiments can be an effective tool for building comprehension of index insurance and assessing risk preferences
- Experiments may help to characterize demand for an insurance product but only after comprehension is solid
- More complex production settings (higher risk dimentionality) raise important tradeoffs between basis risk and comprehension

Once they understand index insurance, do subjects need to understand the mechanics or just the level of basis risk?

Even the best experimental design will require a cognitive leap to reality

The potential role of progressive locals, leaders, and suppliers *Intra-seasonal index information sent via SMS?*

