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Most attention has been focused on (2), which is not necessarily fortunate.

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DUNCAN THOMAS "Intrahruschold ellocation: an inferential approach"

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Table 2

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Effect of Mother's and Father's Unearned Income on Household and Child Health

-	Log (caloric intake)	Log (protein intake)	# Children Ever Born	Survival Rate	Log (weight for height)	Log (height for age)
Linear Model					······································	
Unearned income of						
Mother	0.456 [2.4] 0.063	1.218 [3.7] 0.170	-5.911 [3.7] -0.734	0.437 [3.3] 0.024	0.317 [3.6] 0.039	0.110 [2.0]
Father						
Other	[1.8] 0.100 [0.7]	[2.0] 0.074 [0.7]	[2.3] - 1.169 [2.6]	[1.9] 0.066	[1.7] -0.027	[2.8] 0.022
Tests for equality of income effects		-		[1.7]	[1.3]	[2.7]
Mother = father [p-value] Tests for joint	2.75 [9.7]	14.17 [0.0]	10.26 [0.1]	9.69 [0.2]	9.18 [0.2]	2.17 [14.0]
significance All coefficients Income	102.80 8.96	132.87 26.34	339.85 26.50	i68.82 17.30	38.84 12.49	187.12 7.68

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Difficulties:

→ Labor income:

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 - → Unobserved differences between households
 - → Marriage market

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Men always prefer women with more assets, so that a woman with more assets will have more choices, and will select a match that is better aligned to her preferences.

Therefore, controlling for the total assets level, the higher a woman's assets at the time of marriage, the better the allocation will reflect women's preference, despite the fact that the household, once formed, is dictatorial, and not collective. **IDEAL EXPERIMENT**

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DEAL EXPERIMENT

Randomly give an income transfer to women or to men, and check whether it has the same impact on expenditure depending on who gets it. We don't have such an experiment... (PROGRESA, in Mexico is random transfer, but it went only to women!).

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- → Duflo (2000). Pension in South Africa.

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- → Permanent income shock; gender of recipient vary.
- → Eligible families are poorer (extended families, more likely to be rural, poor), therefore children in these families would have been in worst shape without the pension.

WEIGHT FOR HEIGHT
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Idea: compare children who live in 3 generations households (with an elderly), where nobody is eligible yet, to children who live in 3 generation households where a man, or a woman, is eligible.

	Percentage	% children
	receiving pension	living with
	(1993)	
_	(1)	(2)
PANEL A: Men		
Age in 1993		
50-54	2.8	9.77
55-59	4.7	7.62
60-64	22	5.5
65 and above	60	8.02
PANEL B: Women		
Age in 1993		
50-54	13.6	8.24
55-59	16.4	10.86
60 and above	77	21.4

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$$w_{ifk} = \pi_w E_f + \pi_m E_m + \sum_{j=1}^4 \gamma_j 1_{(j=k)} + W_{ifk} \lambda + X_{ifk} \delta + \epsilon_{ifk}, \quad (1)$$

		Dep	endent varial	ole: Weight f	or Height Z-	score	
_			0	LS			2SLS
_	(1)	(2)	(3)	(4)	(5)	(6)	(7)
Panel A: GIRLS							
Eligible household	0.14	0.35*	0.34*				
	(0.12)	(0.17)	(0.17)				
Woman eligible				0.24*	0.61*	0.61*	1.19*
(in col. 7: woman receives pension)				(0.12)	(0.19)	(0.19)	(0.41)
Man eligible				-0.011	0.11	0.056	-0.097
(in col. 7: man receives pension)				(0.22)	(0.28)	(0.19)	(0.74)
N. Obs	1574	1574	1533	1574	1574	1533	1533
Panel B: Boys							
Eligible household	0.0012	0.022	0.030				
0	(0.13)	(0.22)	(0.24)				
Woman eligible				0.066	0.28	0.31	0.58
(in col. 7: woman receives pension)				(0.14)	(0.28)	(0.28)	(0.53)
Man eligible				-0.059	-0.25	-0.25	-0.69
(in col. 7: man receives pension)				(0.22)	(0.34)	(0.35)	(0.91)
N. Obs	1670	1670	1627	1670	1670	1627	1627
Presence of older members	No	Yes	Yes	No	Yes	Yes	Yes
Family background variables	No	No	Yes	No	No	Yes	Yes
Child Age dummies	Yes	Yes	Yes	Yes	Yes	Yes	Yes

Table 3: Effect of the program on weight for height OLS and 2SLS regressions

Notes: Standard errors (robust to correlation of residuals within households and heteroscedasticity) in parentheses. Indicator for presence of old men and women: presence of a woman above 50, a man above 50, a woman above 56, a man above 56, a man above 61

Family background variables: father's age and education, mother's age and education and rural or metro residence. Member age variables: family size, number of members aged 0 to 5, 6 to 15, 15 to 24, 24 to 49.

Age dummies: Dummies for whether the child is born in 1991, 1990, or 1989.

The instruments in column (7) are woman eligible and man eligible (the first stage is in table A).

Weight for height is a flow measure of nutrition, will respond fast.

Idea: compare children who live in 3 generations households (with an elderly), where nobody is eligible yet, to children who live in 3 generation households where a man, or a woman, is eligible.

Hope: households who have a woman above 60 not so different from household who have a woman between 55 and 60, yet they are much more likely to receive the pension (22% vs 78%)

$$w_{ifk} = \pi_w E_f + \pi_m E_m + \sum_{j=1}^4 \gamma_j 1_{(j=k)} + W_{ifk} \lambda + X_{ifk} \delta + \epsilon_{ifk}, \quad (1)$$

Results: grandmothers feed girls.

Table 4: Effect of eligibility by gender of the intermediate generation. OLS regressions

	GIRLS	BOYS
	(1)	(2)
Mother's mother	0.48*	0.099
eligible	(0.21)	(0.27)
Father's mother	0.15	0.29
eligible	(0.25)	(0.30)
Mother's father	0.097	0.00052
eligible	(0.34)	(0.43)
Father's father	0.22	0.25
eligible	(0.48)	(0.44)
Control variables:		
Presence of older members	Yes	Yes
Family background variables	Yes	Yes
Age dummies	Yes	Yes
N. Obs.	1457	1552

Notes: Standard errors (robust to correlation of residuals within households and heteroscedasticity) in parentheses.

Family background variables: father's age and education, mother's age and education and rural or metro residence.

family size, number of members aged 0 to 5, 6 to 15, 15 to 24, 24 to 49, Age dummies: Dummies for whether the child is born in 1991, 1990, or 1989. Presence of older members: Dummies for whether there is a woman above 50, a man above 50, a woman above 55, a man above 55, a man above 60.

Potential problems

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Height for age is a stock measure of nutrition, will respond slowly. Difference in difference-type estimate, with older children serving as control group : they have been exposed to better nutrition a smaller fraction of their lives than younger children.



Figure 1: Height for age of children living with eligible womenn, eligible men, no eligible member

Date of Birth

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Height for age is a stock measure of nutrition, will respond slowly. Difference in difference-type estimate, with older children serving as control group : they have been exposed to better nutrition a smaller fraction of their lives than younger children.

$$h_{ifk} = \pi_w (YOUNG * E_f) + \pi_m (YOUNG * E_m) + \beta_w E_f + \beta_m E_f(2)$$

$$\sum_{j=1}^4 \gamma_j 1_{(j=k)} + X_{ifk} \delta + \sum_{j=1}^4 1_{(k=j)} * X_{ifk} \lambda_j + \epsilon_{ifk}, \qquad (3)$$

		Pensio	n Variable	
	Eligibility	Eligibility	Old	Receives
			Grandparent	Pension
		OLS		2SLS
	(1)	(2)	(3)	(4)
Panel A: Girls				
Eligible household*YOUNG	0.68* (0.37)			
Woman pension variable *YOUNG		0.71*	0.40	1.16*
Man pension variable*YOUNG		(0.34) 0.097 (0.57)	(0.27) -0.12 (0.35)	(0.56) -0.071 (0.95)
Eligible household	-0.17 (0.16)			
Woman pension variable	× /	-0.15 (0.17)	-0.039 (0.13)	-0.15 (0.17)
Man pension variable		-0.11 (0.24)	0.027	-0.11 (0.24)
N.obs Panel B: Boys	1533	1533	1533	1533
Eligible household*YOUNG	0.11 (0.31)			
Woman pension variable *YOUNG		0.18 (0.32)	0.026	0.28 (0.47)
Man pension variable*YOUNG		-0.30 (0.32)	0.18 (0.30)	-0.47 (0.71)
Eligible household	-0.15 (0.15)			
Woman pension variable	× /	-0.14 (0.32)	-0.084 (0.69)	-0.15 (0.17)
Man pension variable		-0.073 (0.21)	-0.011 (0.14)	-0.057 (0.21)
N. Obs	1627	1627	1627	1627
Age dummies	Yes	Yes	Yes	Yes
Family background variables Family background variable*age dummies	Yes Yes	Yes Yes	Yes Yes	Yes Yes

Table 5: Effect of eligibility on height for age. OLS regressions

Notes: Standard errors (robust to correlation of residuals within households and heteroscedasticity) in parentheses.

Family background variables: father's age and education, mother's age and education and rural or metro residence, family size, number of members aged 0 to 5, 6 to 15, 15 to 24, 24 to 49, above 50 Age dummies: Dummies for whether the child is born in 1991, 1990,or 1989.

Potential problems

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- → Endogenous family recomposition

Height for age is a stock measure of nutrition, will respond slowly. Difference in difference-type estimate, with older children serving as control group : they have been exposed to better nutrition a smaller fraction of their lives than younger children.

$$h_{ifk} = \pi_w(YOUNG * E_f) + \pi_m(YOUNG * E_m) + \beta_w E_f + \beta_m E_f(2)$$

$$\sum_{j=1}^4 \gamma_j 1_{(j=k)} + X_{ifk} \delta + \sum_{j=1}^4 1_{(k=j)} * X_{ifk} \lambda_j + \epsilon_{ifk}, \qquad (3)$$

Results: Same as weight for height

Browning and Chiappori (1998), Bourguignon, Chiappori and Lechene (1993). Test generally pass.

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- PSID data, State level sex ratio. State level divorce law.
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- Sex ratio have the same effects.

		Hours/1000									
	Unre	stricted	Unre	stricted	Ge	neral	Col	lective	Sharing		
	Mod	el With	Mode	el With	Coll	lective	Mod	el With	Rule		
	Divo	rce Law	Aggı	regated	М	odel	Caring		With Caring		
	Du	mmies	Law D	Jummies					_		
	Wives	Husbands	Wives	Husbands	Wives	Husbands	Wives	Husbands	•		
$\log \omega_f$	1.409	-0.810	1.427	-0.756	1.427	-0.760	0.873	-1.056	-56.638		
	(0.346)	(0.321)	(0.340)	(0.323)	(0.340)	(0.322)	(0.289)	(0.315)	(29.524)		
$\log \omega_h$	0.782	-0.597	0.749	-0.564	0.748	-0.568	0.271	-0.827	-25.346		
	(0.296)	(0.287)	(0.296)	(0.288)	(0.296)	(0.288)	(0.258)	(0.273)	(22.543)		
$\log \omega_f \times \log \omega_h$	-0.440	0.273	-0.433	0.255	-0.433	0.257	-0.215	0.374	20.063		
	(0.126)	(0.123)	(0.125)	(0.124)	(0.125)	(0.123)	(0.104)	(0.119)	(10.744)		
Nonlabor	-0.009	-0.006	-0.008	-0.006	-0.008	-0.006	-0.007	-0.006	0.698		
Income/1000	(0.004)	(0.004)	(0.003)	(0.004)	(0.003)	(0.004)	(0.003)	(0.004)	(0.170)		
Sex Ratio	-1.796	4.549	-2.143	4.379	-2.283	4.267	-2.314	4.034	216.280		
	(0.965)	(1.177)	(0.956)	(1.139)	(0.700)	(1.024)	(0.727)	(1.032)	(88.221)		
Divorce Laws Index			-45.685	80.672	-43.994	81.894	-46.004	79.733	4309.954		
			(14.136)	(15.529)	(11.769)	(14.337)	(12.579)	(14.679)	(1713.692)		
Divorce Laws											
Property Division	-0.102	0.047									
(Community=1)	(0.084)	(0.082)									
Mutual/Unilateral	-0.117	0.022									
(Mutual=1)	(0.050)	(0.053)									
Enforcement	-0.050	0.091									
(Court payment=1)	(0.036)	(0.035)									
Spousal Interest	0.003	0.112									
(Degree as asset=1)	(0.029)	(0.027)									
Intercept	1.174	1.102	1.326	1.071	1.391	1.134	2.720	1.970			
	(0.849)	(0.941)	(0.832)	(0.927)	(0.777)	(0.883)	(0.570)	(0.914)			
Children (≤ 6)	-0.539	0.126	-0.510	0.129	-0.512	0.127	-0.592	0.092			
	(0.158)	(0.112)	(0.155)	(0.112)	(0.155)	(0.111)	(0.151)	(0.112)			
Children (7–17)	-0.098	0.036	-0.087	0.041	-0.087	0.041	-0.098	0.031			
	(0.039)	(0.038)	(0.037)	(0.037)	(0.037)	(0.037)	(0.037)	(0.038)			
Education	-0.018	0.036	-0.023	0.036	-0.022	0.036	-0.019	0.037			
	(0.018)	(0.012)	(0.018)	(0.012)	(0.018)	(0.012)	(0.018)	(0.012)			
Age	-0.128	0.064	-0.130	0.065	-0.131	0.064	-0.160	0.047			
	(0.048)	(0.042)	(0.047)	(0.042)	(0.046)	(0.042)	(0.045)	(0.043)			
White	-0.017	-0.021	-0.010	-0.015	-0.005	-0.011	-0.018	-0.013			
	(0.049)	(0.051)	(0.049)	(0.051)	(0.043)	(0.048)	(0.044)	(0.047)			
Value of Function	22	2.902	23	.473	23	.497	26	.057			
Newey-West Test					0.	024	2.	.584			

TABLE 2 **GMM** PARAMETER ESTIMATES

Notes: • Asymptotic standard errors in parentheses. • Instruments: Second order polynomial in age and education (M-F), Father Education (M-F), White (M-F), Spanish (M-F), City size (3 dummies), North-East, North-Central, West, Protestant (M-F), Jewish (M-F), Catholic (M-F), Sex ratio, Divorce Laws.
The parameters of the sharing rule are divided by 1,000 (except the one associated with nonlabor income).
Each regression includes three region dummies (North East, North Central and West).

Browning and Chiappori (1998), Bourguignon, Chiappori and Lechene (1993). Test generally pass. Specification problems are the same as those encountered above.

Chiappori-Fortin-Lacroix: Divorce laws and sex ratio

- PSID data, State level sex ratio. State level divorce law.
- Divorce laws that are more favorable to women leads to lower female labor supply, higher male labor supply.
- Sex ratio have the same effects.
- Test: ratio of the coefficient of sex ratio and divorce law should be the same for men and for women.

		Hours/1000									
	Unre	stricted	Unre	stricted	Ge	neral	Col	lective	Sharing		
	Mod	el With	Mode	el With	Coll	lective	Mod	el With	Rule		
	Divo	rce Law	Aggı	regated	М	odel	Caring		With Caring		
	Du	mmies	Law D	Jummies					_		
	Wives	Husbands	Wives	Husbands	Wives	Husbands	Wives	Husbands	•		
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$\log \omega_h$	0.782	-0.597	0.749	-0.564	0.748	-0.568	0.271	-0.827	-25.346		
	(0.296)	(0.287)	(0.296)	(0.288)	(0.296)	(0.288)	(0.258)	(0.273)	(22.543)		
$\log \omega_f \times \log \omega_h$	-0.440	0.273	-0.433	0.255	-0.433	0.257	-0.215	0.374	20.063		
	(0.126)	(0.123)	(0.125)	(0.124)	(0.125)	(0.123)	(0.104)	(0.119)	(10.744)		
Nonlabor	-0.009	-0.006	-0.008	-0.006	-0.008	-0.006	-0.007	-0.006	0.698		
Income/1000	(0.004)	(0.004)	(0.003)	(0.004)	(0.003)	(0.004)	(0.003)	(0.004)	(0.170)		
Sex Ratio	-1.796	4.549	-2.143	4.379	-2.283	4.267	-2.314	4.034	216.280		
	(0.965)	(1.177)	(0.956)	(1.139)	(0.700)	(1.024)	(0.727)	(1.032)	(88.221)		
Divorce Laws Index			-45.685	80.672	-43.994	81.894	-46.004	79.733	4309.954		
			(14.136)	(15.529)	(11.769)	(14.337)	(12.579)	(14.679)	(1713.692)		
Divorce Laws											
Property Division	-0.102	0.047									
(Community=1)	(0.084)	(0.082)									
Mutual/Unilateral	-0.117	0.022									
(Mutual=1)	(0.050)	(0.053)									
Enforcement	-0.050	0.091									
(Court payment=1)	(0.036)	(0.035)									
Spousal Interest	0.003	0.112									
(Degree as asset=1)	(0.029)	(0.027)									
Intercept	1.174	1.102	1.326	1.071	1.391	1.134	2.720	1.970			
	(0.849)	(0.941)	(0.832)	(0.927)	(0.777)	(0.883)	(0.570)	(0.914)			
Children (≤ 6)	-0.539	0.126	-0.510	0.129	-0.512	0.127	-0.592	0.092			
	(0.158)	(0.112)	(0.155)	(0.112)	(0.155)	(0.111)	(0.151)	(0.112)			
Children (7–17)	-0.098	0.036	-0.087	0.041	-0.087	0.041	-0.098	0.031			
	(0.039)	(0.038)	(0.037)	(0.037)	(0.037)	(0.037)	(0.037)	(0.038)			
Education	-0.018	0.036	-0.023	0.036	-0.022	0.036	-0.019	0.037			
	(0.018)	(0.012)	(0.018)	(0.012)	(0.018)	(0.012)	(0.018)	(0.012)			
Age	-0.128	0.064	-0.130	0.065	-0.131	0.064	-0.160	0.047			
	(0.048)	(0.042)	(0.047)	(0.042)	(0.046)	(0.042)	(0.045)	(0.043)			
White	-0.017	-0.021	-0.010	-0.015	-0.005	-0.011	-0.018	-0.013			
	(0.049)	(0.051)	(0.049)	(0.051)	(0.043)	(0.048)	(0.044)	(0.047)			
Value of Function	22	2.902	23	.473	23	.497	26	.057			
Newey-West Test					0.	024	2.	.584			

TABLE 2 **GMM** PARAMETER ESTIMATES

Notes: • Asymptotic standard errors in parentheses. • Instruments: Second order polynomial in age and education (M-F), Father Education (M-F), White (M-F), Spanish (M-F), City size (3 dummies), North-East, North-Central, West, Protestant (M-F), Jewish (M-F), Catholic (M-F), Sex ratio, Divorce Laws.
The parameters of the sharing rule are divided by 1,000 (except the one associated with nonlabor income).
Each regression includes three region dummies (North East, North Central and West).

Browning and Chiappori (1998), Bourguignon, Chiappori and Lechene (1993). Test generally pass. Specification problems are the same as those encountered above.

Chiappori-Fortin-Lacroix: Divorce laws and sex ratio

- PSID data, State level sex ratio. State level divorce law.
- Divorce laws that are more favorable to women leads to lower female labor supply, higher male labor supply.
- Sex ratio have the same effects.
- Test: ratio of the coefficient of sex ratio and divorce law should be the same for men and for women.
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- Test does not reject equality of ratio.
- Specification check for Singles: no similar effects.

	0	LS	GN	/M
	Wowen	Men	Wowen	Men
$\log \omega$	-0.036	-0.040	-0.177	0.171
	(0.049)	(0.048)	(0.253)	(0.207)
Nonlabor Income (/1000)	-0.001	-0.001	-0.001	-0.003
	(0.001)	(0.001)	(0.004)	(0.002)
Sex Ratio	4.187	1.121	5.857	0.695
	(2.569)	(2.070)	(2.819)	(2.488)
Divorce Laws Index	-0.018	0.015	-0.152	-0.025
	(0.039)	(0.034)	(0.160)	(0.118)
Intercept	-0.374	1.186	-0.739	1.405
-	(1.243)	(1.020)	(1.294)	(1.137)
Education	0.077	0.038	0.095	0.000
	(0.020)	(0.021)	(0.035)	(0.045)
Age	0.052	-0.015	0.079	-0.047
-	(0.038)	(0.030)	(0.062)	(0.036)
White	0.123	0.182	0.111	0.206
	(0.111)	(0.089)	(0.166)	(0.110)
North East	-0.083	-0.052	-0.094	-0.114
	(0.104)	(0.082)	(0.123)	(0.111)
North Central	-0.202	0.038	-0.193	0.015
	(0.078)	(0.075)	(0.081)	(0.080)
West	-0.243	-0.166	-0.184	-0.146
	(0.101)	(0.092)	(0.121)	(0.117)
Value of Function			4.470	9.591
Number of Observations	572	498	572	498

TABLE 3 Parameter Estimates – Singles Hours/1000

PRODUCTION EFFICIENCY: UDRY

Idea: Investment should not be affected by bargaining power. An efficient household should first maximize the total size of the pie, and then divide the pie according to bargaining power.

PRODUCTION EFFICIENCY: UDRY

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Setting: Burkina-Faso. Very poor, semi-arid area. There is on average 1.8 wives for each head of the household. Important characteristic: Women and men each control their own plots.

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Setting: Burkina-Faso. Very poor, semi-arid area. There is on average 1.8 wives for each head of the household. Important characteristic: Women and men each control their own plots.

Test: for a given year, household and crop, is the yield (and potentially the inputs) a function of the gender of the person who owns the plot?

$$Q_{htci} = X_{htci}\beta + \gamma G_{htci} + \lambda_{htc} + \epsilon_{htci}$$

Where h: household, t: time, c: crop, i: plot Q_{htci} : yield on plot X_{htci} : control variable on plot λ_{htc} : household-time-crop fixed effect.

Test: is γ equal to zero?

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TABLE 3

OLS FIXED-EFFECT ESTIMATES OF THE DETERMINANTS OF PLOT YIELD AND Ln(Plot Output) (× 1,000 FCFA) Dependent Variable: Value of Plot Output/Hectare

	Household-			HOUSEHOLD-CROP	YEAR EFFECTS
	Year-Crop Effects: All Crops (1)	HOUSEHOLD- Millet Only (2)	White Sorghum (3)	Vegetables (4)	All Crops: CES* (5)
		31	41	134	1.67
Mean of dependent variable Gender: (1 = female) Plot size: 1st decile 2d decile 3d decile 4th decile 6th decile 7th decile 8th decile 9th decile 10th decile Ln(area) Toposequence: Uppermost Top of slope Mid-slope Near bottom	$\begin{array}{r} 89\\ -27.70 (-4.61)\\ 133.99 (3.50)\\ 69.10 (4.38)\\ 63.45 (5.52)\\ 34.08 (2.88)\\ -2.04 (29)\\ -13.44 (-1.78)\\ -17.23 (-2.59)\\ -26.68 (-3.81)\\ -31.52 (-4.49)\\ \end{array}$	$\begin{array}{c} 31\\ -10.36 & (-2.53)\\ -28.35 & (-2.67)\\ 8.64 & (.82)\\ 16.95 & (1.81)\\ 9.79 & (1.12)\\99 & (11)\\ -13.01 & (-1.73)\\ -12.97 & (-1.34)\\ -21.50 & (-2.65)\\ -20.56 & (-2.55)\\ \end{array}$	$\begin{array}{c} -19.38 & (-4.43) \\ -17.90 & (-1.92) \\ 52.30 & (3.16) \\ 47.68 & (4.77) \\ 26.73 & (3.12) \\ -6.38 & (-1.16) \\ -11.31 & (-1.69) \\ -28.58 & (-4.82) \\ \cdots & 28.65 & (-4.98) \\ -37.70 & (-6.03) \\ \end{array}$ $\begin{array}{c} -14.60 & (-1.73) \\ -11.27 & (-1.47) \\ -8.62 & (-1.15) \\ -5.36 & (71) \end{array}$	$\begin{array}{cccc} -34.27 & (-2.21) \\ 237.10 & (4.66) \\ 63.97 & (2.38) \\ 35.87 & (1.52) \\ 4.21 & (.18) \\ -6.65 & (26) \\ -33.54 & (90) \\ 31.04 & (.73) \\ \end{array}$	$\begin{array}{c}20 (-3.56) \\$
Soil types: 11 12 13 31 32 33 37 45 46 51 Location: Compound	$\begin{array}{cccc} -32.20 & (93) \\ 41.82 & (1.11) \\ 102.92 & (1.10) \\ 1.86 & (.36) \\ 6.38 & (.99) \\ 29.42 & (2.14) \\ 7.69 & (1.37) \\ 5.66 & (1.03) \\ -17.03 & (-1.20) \\ 8.57 & (.90) \\ 1.54 & (.19) \end{array}$	$\begin{array}{cccc} -6.13 & (92) \\ 4.92 & (1.18) \\ 7.43 & (1.11) \\ 10.65 & (1.55) \\ 10.26 & (1.23) \\ 8.56 & (.67) \\ 6.20 & (.80) \\ 7.42 & (1.15) \\ -25.95 & (-1.98) \\ 43.77 & (1.72) \\ 9.69 & (2.67) \end{array}$	$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$	$\begin{array}{c} -36.66 & (66) \\ -19.36 & (38) \\ -76.60 & (49) \\ 52.92 & (.46) \\ 12.96 & (.26) \end{array}$	$\begin{array}{c}89 & (-2.34) \\ .23 & (.74) \\ .69 & (1.01) \\ .08 & (.83) \\ .07 & (.74) \\ .18 & (1.14) \\ .13 & (1.36) \\ .06 & (.67) \\32 & (-1.16) \\ .05 & (.42) \end{array}$

NOTE. ---t-ratios (in parentheses) and test statistics reported in the text are based on heteroskedastic-consistent estimates of the variance-covariance matrix. The omitted plot size category is the 5th decile. The omitted toposequence is bottom land. The omitted soil type is "all others," and the omitted location is "bush" (far from the village). • Dependent variable of col. 5 is In(value of plot output).

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$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$			House	CHOLD-YEAR-CROP EFFE	стя	(1.000
$\begin{array}{c c c c c c c c c c c c c c c c c c c $		Male Labor per Hectare	Female Labor per Hectare (2)	Child Labor per Hectare (3)	Nonhousehold Labor per Hectare (4)	Manure (1,000 kg per Hectare) (5)
Gender $(1 = female)$ -668.47 (-9.60) 70.23 (1.53) 1153.16 (210.7) Plot size: $1,209.72$ (2.53) $1,462.21$ (5.71) 740.80 (1.17) 193.35 $(.43)$ 24.79 (2.45) Ist decile $1,209.72$ (2.53) $1,462.21$ (5.71) 740.80 (1.17) 193.35 $(.43)$ 24.79 (2.45) 2d decile 417.18 (3.25) $1,131.01$ (5.82) 143.12 (1.11) 487.39 (1.28) 7.99 (96) 2d decile 245.94 (2.74) 799.12 (6.72) 133.16 (1.53) 689.39 (1.27) 2.58 (44) 3d decile 96.53 (1.71) 407.87 (5.02) 72.51 $(.68)$ 378.18 (1.07) -6.18 (-1.15) 4th decile 55 (01) -69.25 (-1.36) -72.15 (98) 57.48 $(.80)$ -2.14 (33) 6th decile 55 (01) -69.25 (-1.36) -72.15 (98) 55.1 $(.64)$ -11.08 (-1.56) 7th decile 55.12 (2.97) -306.51 (-5.96) -59.53 (60) 65.51 (87) -11.64 (-1.88) 9th decile 375.53 (-6.23) 386.78 (-6.61) -184.61 (-1.61) 43.81 (-1.07) -16.41 (-2.46) 9th decile 413.36 (-6.79) 373.57 (-5.16) -220.64	a'	(1)	(-/	-19546(-2.34)	-428.41 (-1.70)	- 16.33 (-2.54)
Toposequence: 41.62 $(.35)$ -1.92 (02) -35.12 $(.38)$ 144.02 $(.83)$ $.26$ $(.00)$ Uppermost 29.36 $(.30)$ 91.02 (1.07) 35.15 $(.38)$ 144.02 $(.83)$ $.26$ $(.00)$ Top of slope 29.36 $(.30)$ 91.02 (1.07) 35.15 $(.38)$ 144.02 $(.83)$ $.26$ $(.00)$ Mid-slope 36.08 $(.38)$ $.57$ $(.01)$ $.10$ $(.00)$ -15.45 (11) 1.14 $(.10)$ Mid-slope 36.08 $(.18)$ $.57$ $(.01)$ -98.03 (-1.05) 23.27 $(.17)$ 2.88 $(.20)$	Gender (1 = female) Plot size: 1st decile 2d decile 3d decile 4th decile 6th decile 7th decile 9th decile 10th decile 10th decile Toposequence: Uppermost Top of slope Mid-slope	$\begin{array}{c cccc} - 668.47 & (-9.60) \\ 1,209.72 & (2.53) \\ 417.18 & (3.25) \\ 245.94 & (2.74) \\ 96.53 & (1.71) \\55 & (01) \\ -153.12 & (-2.97) \\ -375.53 & (-6.23) \\ -413.36 & (-6.79) \\ -490.11 & (-7.72) \\ \hline 41.62 & (.35) \\ 29.36 & (.30) \\ 36.08 & (.38) \\ \end{array}$	$\begin{array}{ccccc} 70.23 & (1.53) \\ 1,462.21 & (5.71) \\ 1,131.01 & (5.82) \\ 799.12 & (6.72) \\ 407.87 & (5.02) \\ -69.25 & (-1.36) \\ -306.51 & (-5.96) \\ -386.78 & (-6.61) \\ -373.57 & (-5.16) \\ -418.06 & (-6.08) \\ \end{array}$ $\begin{array}{c} -1.92 & (02) \\ 91.02 & (1.07) \\ .57 & (.01) \\ 75.94 & (.86) \end{array}$	$\begin{array}{c} -193.40 & (-2.34) \\ 740.80 & (1.17) \\ 143.12 & (1.11) \\ 133.16 & (1.53) \\ 72.51 & (.68) \\ -72.15 & (98) \\ -59.53 & (60) \\ -184.61 & (-1.61) \\ -269.99 & (-1.83) \\ -219.27 & (-1.86) \\ \hline -55.52 & (51) \\ 35.15 & (.38) \\ .10 & (.00) \\ -98.03 & (-1.05) \end{array}$	$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$	$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$
·						
	Soil Types: 3	103.49 (.60)	-31.68 (23)	235.74 (.86) 21.88 (.44)	175.29 (.50) 66.04 (.47)	-11.80 (-1.18) 07 (01)
Soil Types: 3 103.49 (.60) -31.68 (23) 235.74 (.86) 175.29 (.50) -11.80 (-1.1) 3 21.88 (.44) 66.04 (.47)07 (07)	7 11 12	-65.79 (85) -28.77 (09) 1,051.98 (.82)	-52.06 (34) 367.34 (1.63) -38.50 (29)	-778.86 (-4.36) 62.36 (.44)	262.71 (.70) 368.47 (1.13) - 187.07 (89)	70 (08) 16.32 (1.48)
Soil Types: 103.49 (.60) -31.68 (23) 235.74 (.86) 175.29 (.50) -11.80 (-1.1) 3 -65.79 (85) -30.39 (28) 21.88 (.44) 66.04 (.47) 07 (67) 7 -65.79 (85) -30.39 (28) 21.88 (.44) 66.04 (.47) 07 (67) 11 -28.77 (09) -52.06 (34) -778.86 (-4.36) 262.71 ($.70$) 70 (67) 12 $1.051.98$ ($.82$) $.367.34$ (1.63) 62.36 ($.44$) -187.07 (89)	13 21	274.48 (1.53) 196.37 (.95)	-43.41 (49)	-42.87 (35)	37.73 (.27) 115.56 (1.00)	2.86 (.18) 6.43 (1.29)
Soil Types: 103.49 (.60) -31.68 (23) 235.74 (.86) 175.29 (.50) -11.80 (-1.1) 3 -65.79 (85) -30.39 (28) 21.88 (.44) 66.04 (.47) 07 (6) 7 -28.77 (09) -52.06 (34) -778.86 (-4.36) 262.71 (.70) 70 (6) 11 -28.77 (09) -52.06 (34) -778.86 (-4.36) 262.71 (.70) 70 (6) 12 1.051.98 (.82) 367.34 (1.63) 62.36 (.44) 368.47 (1.13) 16.32 (1.4) 12 1.051.98 (.82) 367.34 (1.63) 62.36 (.44) 368.47 (1.13) 16.32 (1.4) 13 274.48 (1.33) -38.50 (29) -187.07 (89) 2.27 2.86 (.1) 21 196.37 (.95) -43.41 (49) -42.87 (35) 37.73 (.27) 2.86 (.1) </th <td>31</td> <td>83.16 (1.59)</td> <td>68.24 (.92) -10.36 (15)</td> <td>173.14 (1.07)</td> <td>-51.08 (44)</td> <td>.73 (.12)</td>	31	83.16 (1.59)	68.24 (.92) -10.36 (15)	173.14 (1.07)	-51.08 (44)	.73 (.12)
Soil Types: 103.49 (.60) -31.68 (23) 235.74 (.86) 175.29 (.50) -11.80 (-1.1) 3 -65.79 (85) -30.39 (28) 21.88 (.44) 66.04 (.47) 07 (07) 7 -65.79 (85) -30.39 (28) 21.88 (.44) 66.04 (.47) 07 (07) 11 -28.77 (09) -52.06 (34) -778.86 (-4.36) 262.71 (.70) 70 (07) 12 $1,051.98$ (.82) 367.34 (1.63) 62.36 (.44) 368.47 (1.13) 16.32 (1.47) 13 274.48 (1.33) -38.50 (29) -187.07 (89) 21 196.37 ($.95$) -43.41 (49) -42.87 (35) 37.73 ($.27$) 2.86 ($.123$) 31 83.16 (1.59) 68.24 ($.92$) 205.90 (2.29) 115.56 (1.00) 64.33 (1.53) 31 83.16 (1.59) -103.66 (15) 173.14 (1.07) -51.08 (44) $.73$ ($.573$)	32 33	250.40 (2.57)	163.76 (1.36)	206.68 (.78)	-113.92 (37)	-17.28 (1.01) -12.75 (94)
Soil Types:103.49 $(.60)$ -31.68 (23) 235.74 $(.86)$ 175.29 $(.50)$ -11.80 (-1.1) 3 -65.79 (85) -30.39 (28) 21.88 $(.44)$ 66.04 $(.47)$ 07 (67) 7 -28.77 (09) -52.06 (34) -778.86 (-4.36) 262.71 $(.70)$ 70 (61) 11 -28.77 (09) -52.06 (34) -778.86 (-4.36) 262.71 $(.70)$ 70 (61) 12 $1,051.98$ $(.82)$ 367.34 (1.63) 62.36 $(.44)$ 368.47 (1.13) 16.32 (1.42) 13 274.48 (1.33) -38.50 (29) -187.07 (89) 21 196.37 $(.95)$ -43.41 (49) -42.87 (35) 37.73 $(.27)$ 2.86 $(.133)$ 31 83.16 (1.59) 68.24 $(.92)$ 205.90 (2.29) 115.56 (1.00) 6.43 (1.53) 32 24.77 $(.50)$ -10.36 (15) 173.14 (1.07) -51.08 (44) $.73$ $(.43)$ 33 250.40 (2.57) 163.76 (1.36) 206.68 $(.78)$ -113.92 (37) 17.28 (1.53)	35	179.46 (1.50)	303.86 (1.90)	248.38 (2.60)	31.14 (.20)	8.34 (1.44
Soil Types:103.49(.60) -31.68 (23)235.74(.86)175.29(.50) -11.80 (-1.1)7 -65.79 (85) -30.39 (28)21.88(.44)66.04(.47) 07 (6)11 -28.77 (09) -52.06 (34) -778.86 (-4.36)262.71(.70) 70 (6)121.051.98(.82)367.34(1.63)62.36(.44)368.47(1.13)16.32(1.4)13274.48(1.33) -38.50 (29) -187.07 (89)21196.37(.95) -43.41 (49) -42.87 (35) 37.73 (.27)2.86(.12)3183.16(1.59)6.2.50(2.29)115.56(1.00)6.43(1.5)3224.77(.50) -10.36 (15)173.14(1.07) -51.08 (44).73(.13)33250.40(2.57)163.76(1.36)206.68(.78) -113.92 (37)17.28(1.6)35179.46(1.50)303.86(1.90)248.38(2.60)195.14(.58) -12.75 (5)	37	82.49 (.70)	-8.33 (10)	79.85 (1.02)	41.90 (.25)	8.00 (1.83
Soil Types:103.49(.60) -31.68 (23)235.74(.86)175.29(.50) -11.80 (-1.1)7 -65.79 (85) -30.39 (28)21.88(.44)66.04(.47) 07 (6)11 -28.77 (09) -52.06 (34) -778.86 (-4.36)262.71(.70) 70 (6)121.051.98(.82)367.34(1.63)62.36(.44)368.47(1.13)16.32(1.4)13274.48(1.33) -38.50 (29) -187.07 (89)21196.37(.95) -43.41 (49) -42.87 (35) 7.73 (.27)2.86(.1)3183.16(1.59)66.24(.92)205.90(2.29)115.56(1.00)6.43(1.5)3224.77(.50) -10.36 (15)173.14(1.07) -51.08 (44).73(.1)33250.40(2.57)163.76(1.36)206.68(.78) -113.92 (37)17.28(1.6)35179.46(1.50)303.86(1.90)248.38(2.60)195.14(.58) -12.75 (53)3782.49(.70)50.84(.30)114.53(1.19)31.14(.20)8.34(1.4)3782.49(.70)50.84(.30)114.53(1.92)41.90(.25)8.00(.11)	45	-187.14 (-1.84)	141.73 (.76)	42.70 (.09)	223.23 (1.27)	-15.45 (79) 80 (17)
Soil Types:103.49(.60) -31.68 (23) 235.74 (.86) 175.29 (.50) -11.80 ($-1.1)$ 7 -65.79 (85) -30.39 (28) 21.88 (.44) 66.04 (.47) 07 (07 11 -28.77 (09) -52.06 (34) -778.86 (-4.36) 262.71 ($.70$) 70 (6)121,051.98($.82$) 367.34 (1.63) 62.36 (.44) 368.47 (1.13) 16.32 (1.43)13 274.48 (1.33) -38.50 (29) -187.07 (89) -187.07 (89)21196.37(.95) -43.41 (49) -42.87 (35) 37.73 ($.27$) 2.86 ($.133$)31 83.16 (1.59) 68.24 ($.92$) 205.90 (2.29) 115.56 (1.00) 6.43 (1.53)32 24.77 ($.50$) -10.36 (15) 173.14 (1.07) -51.08 (44) $.73$ ($.133$)33 250.40 (2.57) 163.76 (1.36) 206.68 ($.78$) -113.92 (37) 17.28 (1.63)35 179.46 (1.50) 303.86 (1.90) 248.38 (2.60) 195.14 ($.58$) -12.75 (53)37 82.49 ($.70$) 50.84 ($.30$) 114.53 (1.19) 8.34 ($.127$) -15.45 (453)36 -187.74	51	95.73 (1.83)	-27.01 (33)	2.93 (.05)	120.70 (1.03)	
$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	Location:		97.16 (QA)	-18.82(31)	- 162.88 (-1.38)	.99 (.24
Soil Types:103.49(.60) -31.68 (23)235.74(.86)175.29(.50) -11.80 (-1.1)7 -65.79 (85) -30.39 (28) 21.88 (.44) 66.04 (.47) 07 (0)11 -28.77 (09) -52.06 (34) -778.86 (-4.36) 262.71 (.70) 70 (6)121.051.98(.82) 367.34 (1.63) 62.36 (.44) 368.47 (1.13) 16.32 (1.4)13274.48(1.33) -38.50 (29) -187.70 (89)(.13)(.14)(.15) 68.24 (.92) 205.90 (2.29) 115.56 (1.00) 6.43 (1.3)21196.37(.95) -43.41 (49) -42.87 (35) 37.73 (.27)2.86(.13)2224.77(.50) -10.36 (15) 173.14 (1.07) -51.08 (44) $.73$ (1.4)33250.40(2.57)163.76(1.36)206.68(.78) -113.92 (37)17.28(1.6)35179.46(1.50)303.86(1.90)248.38(2.60)195.14(.58) -12.75 (5)35179.46(1.50)303.86(1.90)248.38(2.60)195.14(.58) -12.75 (5)3782.49(.70)50.84(.30)114.53(1.19)31.14(.20)8.34(1.2)3782.49(.70)<	Compound *	35.35 (.78)	12.18 (.45)	42.92 (.93)	25.80 (.30)	5.86 (1.60)
Soil Types: 103.49 (.60) -31.68 (23) 235.74 (.86) 175.29 (.50) -11.80 (-1.1) 3 -65.79 (85) -30.39 (28) 21.88 (.44) 66.04 (.47) 07 (6.7) 11 -28.77 (96) -52.06 (34) -778.86 (-4.36) 262.71 (.70) 70 (6.7) 12 1.051.98 (.82) 367.34 (1.63) 62.36 (.44) 368.47 (1.13) 16.32 (1.4) 13 274.48 (1.33) -38.50 (29) -187.07 (89) 21 196.37 (.95) -43.41 (94) -42.87 (35) 37.73 (.27) 2.86 (.1.3) 31 81.6 (1.59) 68.24 (.92) 205.90 (2.29) 115.56 (1.00) 6.43 (1.53) 33 250.40 (2.57) 163.76 (.190) 248.38 (2.60) 195.14 $(.58)$ -12.75 (55)	Village Mean of dependent variable	427.39	466.18	85.55	84.88	7.78
Soil Types: 3 103.49 (.60) -31.68 (-23) 235.74 (.86) 175.29 (.50) -11.80 (-1.1 7 -65.79 (85) -30.39 (28) 21.88 (.44) 66.04 (.47) 07 (6 7 -28.77 (09) -52.06 (34) -778.86 (-4.36) 262.71 ($.70$) 70 (6 11 -28.77 (09) -52.06 (34) -778.86 (-4.36) 262.71 ($.70$) 70 (6 12 $1.051.98$ (.82) 367.34 (1.63) 62.36 (.44) 368.47 (1.13) 16.32 (1.4 12 $1.96.37$ ($.95$) -43.41 (49) -42.87 (35) 37.73 (27) 28.6 ($.43$) 31 83.16 (1.59) 68.24 ($.92$) 205.90 (2.29) 115.56 (1.00) 6.43 (1.53) 31 83.16 (1.59) 68.24 ($.92$) 205.90 (2.29) 115.56 (1.00) 6.43 (1.53) 31 250.40 (2.57) 163.76 (1.36) 206.68 ($.78$) -113.92 (37) 17.28 (1.43) 33 250.40 (2.57) 163.76 (1.36) 206.68 ($.78$) -113.92 (37) 17.28 (1.43) 35 179.46 (1.50) 303.86 (1.90) 248.38 (2.60) 195.14 ($.58$) -12.75 (58) 35 179.46 (1.50) 303.86 (1.90) 248.38 (2.60) 195.14 (2.5) 8.00 (1.43) 37 82.49 ($.70$) 50.84 ($.30$) 114.53 (1.19) 31.14 (20) 8.34 (1.43) 37 82.49 ($.70$) 50.84 ($.30$) 114.53 (1.29) 41.90 ($.25$) 8.00 (1.44) 36 -187.14 (-18.41) 141.73 ($.760$) 42.70 ($.099$) 223.23 (1.27) -15.45 (51) 51 95.73 (1.83) -27.01 (33) 2.93 (0.55) 126.70 (1.05) $.800$ (1.43) 46 -187.14 (-1.84) 141.73 ($.456$) 42.70 ($.099$) 225.28 (-1.38) $.99$ ($.516$) 51 95.73 (1.83) -27.01 (33) 2.93 (0.55) 126.70 (1.05) $.806$ (1.40) 32.94 (1.259) 32.46 (1.259) 32.48 (1.279) -15.45 (51) 46. -138 19.69 ($.70$) 12.18 ($.455$) 42.88 (31) -162.88 (-1.38) $.99$ ($.529$) 466.18 85.55 91.818 1.70	when >0	506.62	517.17	202.88	213.11	

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TABLE 6

ast-Squares Tobit Fixed-Effect Estimates of the Determinants of Plot Input Intensities 1 -

NOTE --- This is the least-squares implementation of Honore's (1992) fixed-effect Tobit estimator. (-ratios are in parentheses.

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$$Q_{htci} = X_{htci}\beta + \gamma G_{htci} + \lambda_{htc} + \epsilon_{htci}$$

Where h: household, t: time, c: crop, i: plot Q_{htci} : yield on plot X_{htci} : control variable on plot λ_{htc} : household-time-crop fixed effect.

Test: is γ equal to zero?

Estimation of a production function suggests that 5.8% gain in production could be obtained just by reallocating inputs across plots (NB: doing the same exercise in the village would lead to a 13% increase in production).



FIG. 2.-Regression of yield on area with household-year-crop effects
LEAST-SQUARES TOBIT FIXED-EFFECT ESTIMATES OF THE DETERMINANTS OF PLOT INPUT INTENSITIES

TABLE 6

				Hous	EHOLD-YEAD	R-CROP EFFI	ECTS			
	Male I per He	abor ctare	Female per He (2)	Labor ctare	Child per H. (3	Labor ectare	Nonhou Labor per (4	sehold Hectare	Manure kg per I (5	(1,000 Hectare)
Gender (1 = female) Plot size:	- 668.47	(-9.60)	70.23	(1.53)	- 195.46	(-2.34)	-428.41	(-1.70)	-16.33	(-2.54)
1st decile	1,209.72	(2.53)	1,462.21	(5.71)	740.80	(1.17)	193.35	(.43)	24.79	(2.42)
2d decile	417.18	(3.25)	1,131.01	(5.82)	143.12	(11.1)	487.39	(1.28)	7.99	(96)
3d decile	245.94	(2.74)	799.12	(6.72)	133.16	(1.53)	689.39	(1.27)	2.58	(.48)
4th decile	96.53	(1.71)	407.87	(5.02)	72.51	(.68)	378.18	(1.07)	-6.18	(-1.12)
6th decile	55	(01)	-69.25	(-1.36)	-72.15	(98)	57.48	(.80)	-2.14	(33)
7th decile	-153.12	(-2.97)	-306.51	(-5.96)	-59.53	(60)	65.51	(.64)	-11.08	(-1.54)
8th decile	-375.53	(-6.23)	-386.78	(-6.61)	-184.61	(-1.61)	-43.81	(30)	-11.01	(-1.61)
9th decile	-413.36	(-6.79)	-373.57	(-5.16)	-269.99	(-1.83)	-255.15	(87)	-11.64	(-1.80)
10th decile	-490.11	(-7.72)	-418.06	(-6.08)	-219.27	(-1.86)	-220.64	(-1.07)	-16.41	(-2.45)
Toposequence:										
Uppermost	41.62	(.35)	-1.92	(02)	-55.52	(51)	20.20	(.12)	-9.22	(62)
Top of slope	29.36	(.30)	91.02	(1.07)	35.15	(.38)	144.02	(.83)	.26	(.02)
Mid-slope	36.08	(.38)	.57	(10)	.10	(00)	-15.45	(11)	1.14	(11)
Near bottom	16.42	(.18)	75.94	(98.)	-98.03	(-1.05)	23.27	(-17)	2.88	(.27)

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- → Why is the household not pareto efficient?
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 - → The "labor market" within the household is not perfect, because of a lack of secure property rights on the land. Men have more labor, but women don't want men to work on their plots because they fear that the plots will then be confiscated by the husband.

→ Udry and Duflo (2001) look at Cote d'Ivoire, where women and men grow different crop on different plot.

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- → One third player: yams(!!)

	Food consumption	Adult goods	Clothing	Prestige goods Education	
	(1)	(2)	(3)	(4)	(5)
PANEL A: RESTRICTED EXC	LUSION RESTRI	CTION TEST: SI	EMI PARAMETE	RIC FORMULATION	
Prediced male non-yam	-0.037	0.178	0.112	0.550	-0.139
income	(0.029)	(0.464)	(0.267)	(0.233)	(0.099)
Predicted yam	0.047	-0.705	0.094	-0.491	0.212
income	(0.032)	(0.588)	(0.282)	(0.155)	(0.136)
Predicted female	-0.006	0.845	0.214	0.534	-0.210
income	(0.034)	(0.623)	(0.370)	(0.192)	(0.130)
F tests (pvalue) :	1.147	0.837	0.203	7.057	1.895
Predicted income variables	(0.339)	(0.479)	(0.894)	(0.000)	(0.143)
Jointly significant					
Predicted income variables	1.711	1.252	0.041	10.584	2.635
significantly different	(0.190)	(0.294)	(0.960)	(0.000)	(0.082)
Coefficient of female crops	1.268	2.501	0.054	17.596	4.059
and yam income equal.	(0.265)	(0.120)	(0.818)	(0.000)	(0.049)

	Staples	Meat	Vegetables	Processed foods	Purchased foods	Food consumed at home
PANEL A: RESTRICTED EXC	CLUSION RESTR	RICTION TEST	: SEMI PARAM	IETRIC FORM	ULATION	
Prediced male non-yam	0.015	-0.053	3 -0.054	0.004	-0.176	0.068
income	(0.077)	(0.090) (0.142)) (0.131)	(0.090)	(0.133)
Predicted yam	0.142	-0.093	3 -0.167	-0.005	-0.018	0.100
income	(0.061)	(0.073) (0.097)) (0.110)) (0.071)	(0.073)
Predicted female	-0.117	0.19	5 0.574	0.266	0.127	-0.013
income	(0.080)	(0.103) (0.144)) (0.164)	(0.135)	(0.104)
F tests (pvalue) :	2.696	2.880	5.640) 1.014	1.803	0.952
Predicted income variables jointly significant	(0.055)	(0.044) (0.002)) (0.393)) (0.157)) (0.422)
Predicted income variables	3.871	4.280	0 8.229) 1.055	5 1.790	0.630
significantly different	(0.027)	(0.019) (0.001)) (0.355)) (0.177)	(0.537)
Coefficient of female crops	7.066	8.440	0 15.467	2.092	2 1.180	0.996
and yam income equal.	(0.010)	(0.005) (0.000)) (0.154)) (0.282)	(0.323)

 Table 4: Restricted exlusion restriction tests

- → Udry and Duflo (2001) look at Cote d'Ivoire, where women and men grow different crop on different plot.
- → Crops are differently affected by (the same) wheather. These are short run income fluctuation, that are perfectly observed.
- → Income variation predicted by rainfall variation should not have effect on bargaining power (there should be short term insurance).
- → Therefore, controlling for changes in total expenditures, we should not see an impact on predicted female income variation and female income variation on changes in expenditures on particular goods.
- → One third player: yams(!!)
- → Results: male and female income affect private "prestige" goods, presumably investment in bargaining power (pagnes and jewelry). female income affect food purchase. Yam associated with only good things.