Impact of weather on wages and employment in rural India, 2004-2011.

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Motivation

- The incidence of droughts is frequent and lasts longer.
- Drought negatively affects crop yields and hence agricultural incomes.
- Increasing trend in the percentage of subsistence households and wage is the only source of livelihood.
 - Plot Marginal Farmer Table Land holdings Plot workers from subsistence hhs
- What are the labor impacts of rainfall variability in rural India for the period 2004-2011?

Relationship to previous work

- A higher temperature negatively affects crop yields and agricultural incomes (Schlenker and Robert, 2009; Taraz, 2018).
- How labor market responds to weather shocks in low and middle incoming countries through the channel of agriculture?
 - Short-term increase in temperature is negatively associated with rural labor employment and labor reallocate from ag to non-ag sector (Jessose et al. 2016; Colmer 2018).
 - Rainfall shock results in lower wages in rural economies. (Jayachandran, 2006).

Contribution to the climate and development literature

- We explore a non-linear relationship between weather variables and labor demographics.
- A new data set that integrates spatial distribution of agro-climatic variables within a district with the national employment and unemployment situation.
- A critical examination of the relationship between rainfall variability and employment.

Profit Maximization Problem

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Introduction

Theoretical Framework

Two types of agents: a single commercial farmer and nsubsistence households. A commercial farmer with Cobb-Douglas production function $Q = Q(L, \bar{K}; \theta)$ with constant returns to scale.

$$Q = aL^{1-\alpha}$$
 where $a = A\bar{K}^{\alpha}$ and $\alpha \in (0,1)$.

For a given wage, w, and output price, p, the profit maximization implies that

$$p\frac{\partial \pi}{\partial L} = w \Longrightarrow w = \frac{p(1-\alpha)a}{L^{\alpha}} \Longrightarrow L^* = \left[\frac{p(1-\alpha)a}{w}\right]^{\frac{1}{\alpha}}$$

Taking logs and differentiating, we obtain the labor demand elasticity,

$$\frac{\partial ln(L)}{\partial ln(w)} = -\frac{1}{\alpha}.$$

Labor reallocation decision: Set up

Introduction

Following (Maggie Y. Liu et al., 2021), assume that there are two sectors: ag (A) and non-ag (N) in a two period model. The wage of individual i in sector J(J=A,N) is given by:

$$w_i^J = \mu^J + \beta^J \varepsilon_i,$$

Baseline wage depends on climatic determinants, θ , through agricultural productivity. That is, $\mu^J = \mu^J(\theta)$ and $\frac{\partial \mu^A}{\partial \theta} < \frac{\partial \mu^N}{\partial \theta} < 0$.

Switching cost, $c = c(c_{monetary}, c_{non-monetary})$, from sector A to N in period 2. So,

$$\mu^{N}(\theta) + \beta^{N} \varepsilon_{i} - c > \mu^{A}(\theta) + \beta^{A} \varepsilon_{i}$$

Rearranging, we get:

$$\varepsilon_i > \frac{\mu^A(\theta) - \mu^N(\theta) + c}{\beta^N - \beta^A}.$$

Labor reallocation constraints: Incentive and Feasibility

Incentive constraint defined as

$$\frac{Mig}{Pop} = 1 - \Phi\left(\frac{\mu^A(\theta) - \mu^N(\theta) + c}{\beta^N - \beta^A}\right) \tag{1}$$

Feasibility constraint defined as

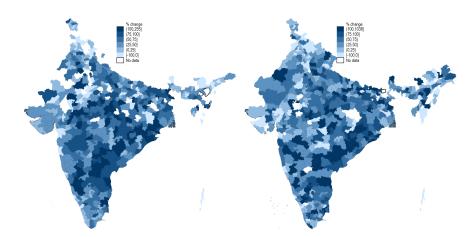
$$\frac{Mig}{Pop} = 1 - \Phi\left(\frac{c - \mu^A(\theta)}{\beta^A}\right) \tag{2}$$

By taking logarithms and log-linearizing both sides of equations 1 and 2, we obtain the basic empirical equation:

$$ln\left(\frac{Mig}{Pop}\right) = \alpha + \beta ln(\theta) + \gamma c.$$

Introduction

District level percentage change in rural average day wage in real terms, NSS EUS, 2004-11



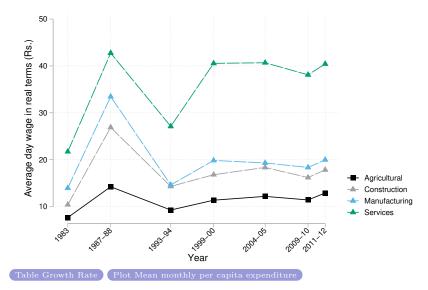
(a) Casual wage labor: Agriculture

(b) Casual wage labor: Non-agricultural

Data

- Unit-level data from five rounds of National Sample Surveys on Employment and Unemployment Situation in India (NSS EUS). Descriptive statistics
- Climate variables were obtained from the University of East Anglia's Climate Research Unit (CRU)
 - 30 years monthly precipitation and monthly mean temperature
- Weather variables were obtained from the Government of India's National Centre for Medium Range Weather Forecasting (NCMRWF)
 - June-September daily accumulated rainfall
 - June-September daily mean temperature
 - June-September daily relative humidity

Real average day wage trend, Rural All-India



Empirical Specification: Wages and Employment

Base specification:

$$ln(y_{dt}) = f(\theta_{dt}) + \alpha_d + \phi_s t + \varepsilon_{dt}$$

 y_{dt} represents the outcome of interest in district d in year t; $f(\theta_{dt})$ is a function of rainfall and temperature; Z_{dt} represents controls at the district level; α_d is a vector of district fixed effects; ϕ_{st} is a vector of state-year fixed effects; ε_{dt} represents error terms.

Effects of weather on the share of employment by sector at the

LIIC	Effects of weather on the share of employment by sector at the				
dist	rict level (log): Panel estimates				
		Agriculture	Non-agricultural		
	Degree days (thousands) $(8-32^{\circ}C)$	-1.438	-1.227		
		(1.381)	(1.500)		
		[1.159]	[1.252]		
	Degree days squared	0.473	0.102		
		(0.289)	(0.301)		
		[0.026]**	lo orei		

 $|0.236|^{**}$ |0.256|Square root of HDD $(34^{\circ}C)$ 0.006-0.006(0.008)(0.009)[0.008][0.008]Monsoon Rainfall_t (Kg/m2, \log) 0.0910.040 $(0.047)^*$ (0.049)[0.043]** [0.045]

Observations 15,890 24,235 Number of Districts 568 561 Controls for lagged weather and long-run climatic variables;

District and State-year FEs. 12 / 19

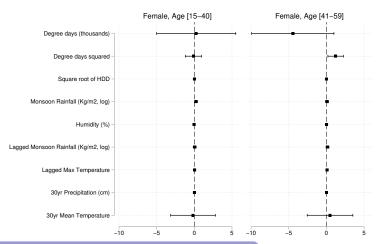
Effects of weather on the average day wage at the district level

	Agriculture	Non-agricultura
Degree days (thousands) $(8-32^{\circ}C)$	1.441	1.675
	(1.872)	(1.350)
	[1.749]	$[1.022]^*$
Degree days squared	-0.251	-0.269
	(0.325)	(0.276)
	[0.327]	[0.226]
Square root of HDD $(34^{\circ}C)$	0.006	0.0002
	(0.008)	(0.009)
	[0.007]	[0.007]
Monsoon Rainfall _t (Kg/m2, log)	0.026	-0.003
	(0.032)	(0.043)

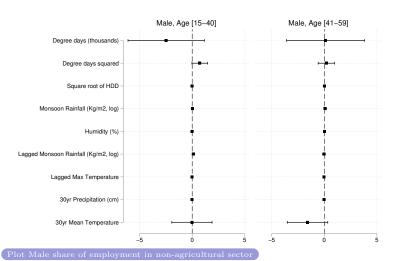
(0.040)[0.033][0.037]Observations 5,883 11,045 Number of Districts 428 544 Controls for lagged weather and long-run climatic variables;

District and State-year FEs.

Impact of weather on female share of rural employment in agriculture at the district level (log): Coefplots



Impact of weather on male share of rural employment in agriculture at the district level (log): Coefplots



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Precipitat	ion variability m	easures: share of	employme	ent in	

0.473**

(0.236)

0.108**

(0.050)

-0.015(0.021)

15,890

568

Background and Data

Specification

0.472**

(0.237)

0.095*

(0.055)

0.035(0.270)

15,890

568

Results

0.476**

(0.236)

0.087*

(0.045)

-0.008(0.012)

15,890

568

16 / 19

agriculture at the district level (log).					
		Rainfall CV	Shannon	Dry 10-day	
	Degree days ₁₀₀₀ $(8-32^{\circ}C)$	-1.438	-1.436	-1.490	
		(1.158)	(1.160)	(1.157)	

grio	Rainfall CV Shannon			
		Rainfall CV	Shannon	
	Degree days ₁₀₀₀ $(8 - 32^{\circ}C)$	-1.438	-1.436	
		(1.158)	(1.160)	

Controls for lagged weather and long-run climatic variables;

Theoretical Framework

Degree days squared

Shannon Diversity

Number of Districts

District and State-year FEs.

Rainfall CV

Dry 10-day

Observations

Monsoon Rainfall_t (Kg/m2, \log)

Introduction

Introduction 000	Theoretical Framework	Background and Data	Specification 0	Results 00000•0	
Precipitat	ion variability m	easures: share	of employm	ent in	

(0.061)

-0.646** (0.300)

24,235

561

(0.045)

0.014 (0.013)

24,235

561

17/19

Rainfall CV Shannon	Drv 1
non-agricultural sector at the district level (log).	
Precipitation variability measures: share of employment	in

Rainfall CV

Dry days 10

Observations

Shannon Diversity

Number of Districts

District and State-year FEs.

non-agricultural sector at the district level (log).					
			Rainfall CV	Shannon	Dry 10
	Degree days ₁₀₀₀ $(8-3)$	$32^{\circ}C)$	-1.190	-1.235	-1.12
			(1.240)	(1.949)	(1.25

on-agricultural sector at the district level (log).					
	Rainfall CV	Shannon	Dry 10		
Degree days ₁₀₀₀ $(8 - 32^{\circ}C)$	-1.190	-1.235	-1.1		
	(1.249)	(1.243)	(1.2)		
Degree days squared	0.101	0.126	0.0		

Controls for lagged weather and long-run climatic variables;

	Rainfall CV	Shannon	Dry 10-day
Degree days ₁₀₀₀ $(8 - 32^{\circ}C)$	-1.190	-1.235	-1.126
	(1.249)	(1.243)	(1.256)
Degree days squared	0.101	0.126	0.096
	(0.255)	(0.254)	(0.256)
Monsoon Rainfall _t (Kg/m2, log)	-0.018	-0.036	0.048

(0.058)

0.051*(0.027)

24,235

561

Robustness check: share of employment at the district level (log).

	Agriculture	Non-agricultural
Degree days (thousands) $(8-32^{\circ}C)$	-1.602	-1.037
	(1.202)	(1.279)
$DD \times High GWR$	-0.133	0.271
	(0.204)	(0.276)
Monsoon Rainfall (Kg/m2, log)	0.100**	0.011
	(0.046)	(0.049)
Rain x High Discharge	-0.026	0.064
	(0.048)	(0.049)
Observations	15,630	23,460

Controls for lagged weather and long-run climatic variables;

District and State-year FEs.

Main Findings:

- Degree days increases the share of the labor in construction sector at the decreasing rate.
- Young and elderly male, and young female allocates labor to the construction sector in response to the rainfall variability.
- Construction sector provides a cushion for casual wage labor impacted due to climatic distress.

Future Directions:

• Using India Human Development Survey repeat the analyses at the individual level.

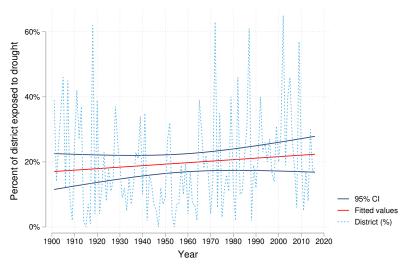


Figure: 1. Incidence of drought (rainfall below 20th percentile of long-run historical average) at the district level (based on 2011 India Census district geographic boundaries) between 1901-2016. (Source: CRU)

Impact of drought on yields for cereals and pulses

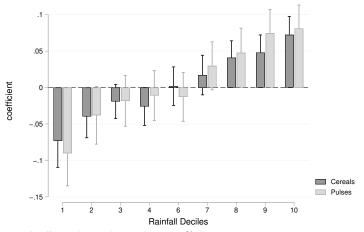


Figure: 2. Coefficient for rainfall deciles and 95% CI in India. The dependent variable is the natural logarithm of annual crop yield (kg per hectare) from 2001 to 2015. The specification include district and year fixed effects. The 5th decile is selected as reference. (Source: ICRISAT)



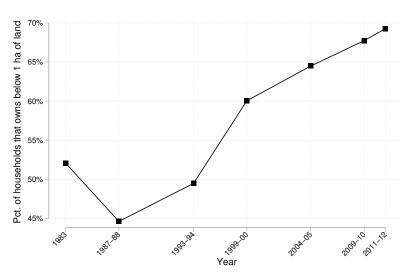


Figure: 3. Percentage of rural households that owns less than 1 hectare of land. Source: Unit-level data from various rounds of National Sample Survey on Employment and Unemployment Situation (NSS EUS) in India.

return

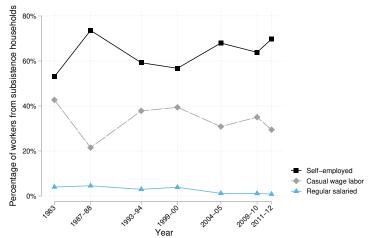
Distribution of average land holdings in rural India.

	Subsisten	ce (Below 1 ha)	Commerci	al (10 and above ha)
Year	Pct.(%)	Mean (sd)	Pct.(%)	Mean (sd)
1983	52.08	0.29 (0.30)	1.71	31.75 (180.23)
1987-88	44.62	0.33(0.30)	9.87	$16.48 \ (09.57)$
1993-94	49.48	0.33(0.30)	8.89	$17.30\ (29.59)$
1999-00	60.05	0.29(0.30)	1.99	$16.03\ (10.96)$
2004-05	64.49	0.27 (0.30)	4.98	16.79 (70.33)
2009-10	67.73	$0.23 \ (0.28)$	4.68	14.97 (07.09)
2011-12	69.26	$0.24 \ (0.28)$	4.95	$14.36 \ (05.27)$

Source: NSS EUS data.



Distribution of types of workers engaged in agricultural sector from subsistence households, Rural All-India





Real Agricultural Earnings, Rural All-India (Rs., in 1983 Rural Maharashtra Prices)

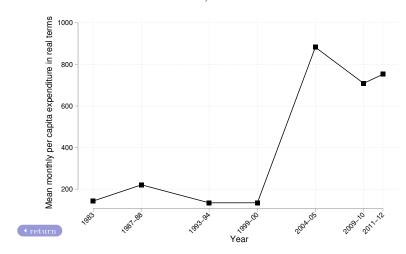
Year	Daily Earnings (Rs.)	Growth Rate (%)	AGR^a (%)
1983	7.65	-	-
1987-88	14.22	85.77	17.15
1993 - 94	9.26	-34.89	-6.98
1999-00	11.35	22.63	4.53
2004-05	12.16	7.16	1.43
2009-10	11.40	-6.24	-1.25
2011-12	12.88	12.97	6.48

a: AGR stands for annualised rates of growth.

Source: NSS EUS data.



Mean monthly per capita expenditure, Rural All-India (Rs., in 1983 Rural Maharashtra Prices)

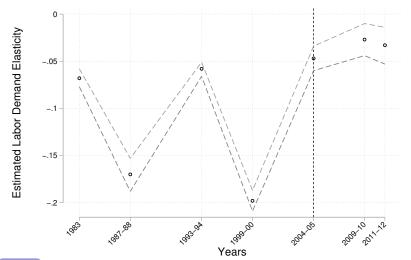


District Level Descriptive Statistics: Local Labor Market in rural India (2004-2011).

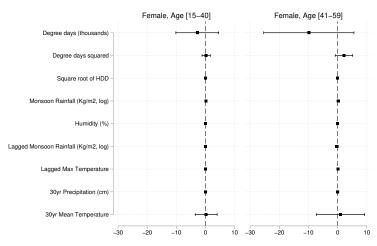
	Mean	Std. Dev.	Std. Dev.
		(within)	(between)
Panel A: Wage (Rs.)			
Avg. Day Wage: Agriculture	75.48	40.90	32.65
Avg. Day Wage: Non-agricultural	189.66	113.06	112.74
Panel B: Employment			
Employment Share: Agriculture	0.09	0.03	0.07
Employment Share: Non-agricultural	0.04	0.02	0.03
Unemployment Share of Labor Force	0.05	0.02	0.03
Panel C: Meteorological			
Daily Max. Temperature (°C)	32.15	0.70	3.63
Monsoon Rainfall (Kg/m2)	1253.25	250.81	788.09

Source: NSS EUS data.

Rural Labor Demand Elasticity for Agriculture: Number of days employed in agriculture in a week.



Impact of weather on female share of rural employment in non-agricultural sector at the district level (log): Coefplots





Impact of weather on male share of rural employment in non-agricultural sector at the district level (log): Coefplots

