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**Gender Disparities in Completing School Education in India
Analyzing Regional Variations**

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Abstract

Is gender disparity greater in North India? This paper seeks to answer this question by examining gender differences in probability of completing school education across regions in India. A Gender Disparity Index is calculated using National Sample Survey Organization unit level data from the 61st Round and regional variations in this index analyzed to examine the hypothesis that gender disparity is greater in the North, comparative to the rest of India. This is followed by an econometric exercise using a logit model to confirm the results of the descriptive analysis after controlling for socio-economic correlates of completing school education. Finally, the Fairlie decomposition method is used to estimate the contribution of explanatory variables in explaining differences in probabilities of completing schooling across regions. The results reveal that gender disparities are greater in North India, for total and rural population, and in Eastern India, for urban population. However, the ‘residual effect’ after accounting for effect of explanatory variables - often referred to as ‘discrimination effect’, as opposed to disparity – is higher in Eastern India, irrespective of the place of residence.

Key Words: discrimination, disparity, gender, Oaxaca decomposition, school education, India.

Gender Disparities in Completing School Education in India

Analyzing Regional Variations

1. Introduction

Is gender discrimination greater in North India? Based on an analysis of infant and child mortality rates, sex ratios and fertility trends, Dyson and Moore (1983) concluded that, relative to their South Indian states counterparts, women in Northern states *were* subjected to higher levels of discrimination. Despite disagreement over the explanation,¹ the empirical finding that gender disparity is stronger in northern India has not been contested.²

However, Dyson and Moore (1983) fail to distinguish between disparity and discrimination. Disparity simply refers to differences in the outcome under consideration (wages, mortality rates, educational attainments, or any such indicator). Such disparities may be caused by differences in socio-economic characteristics. Alternately, differences in outcomes may occur due to socio-cultural forces independent of these characteristics – the inferior outcome may be a consequence of deliberately unfair treatment from the family or society. This is referred to as discrimination – the practice of treating members of a group less fairly than others, simply because the person(s) belong to a particular race,

¹ This observation was explained by Dyson and Moore (1983) in terms of cultural practices – the prevalence of endogamous marriages in South India implied that women had more access to her kin, thereby increasing autonomy. Alternative explanations have been offered for this phenomenon. Bardhan (1974) and Miller (1981) have argued that the prevalence of wet rice cultivation in southern states has created demand for women labour, increasing their participation in economic activities; this has empowered South Indian women. Rosenzweig and Schultz (1982) analysis also associates differences in child survival rates to differences in male and female labor force participation rates. Jeffrey (1993), on the other hand, links lower levels of gender disparities in the south to higher levels of State investment on education and health. Murthi et al (1995) and Dasgupta et al (2004) also argue that public investment in these spheres in states like Kerala and Karnataka have promoted female agency and reduce gender differences in demographic outcomes.

² Basu's (1992) largely qualitative analysis comparing female agency among South and North Indian migrants to Delhi slums finds that the former enjoy greater mobility and freedom of expression than their Northern counterparts. Jejeebhoy's (2001) quantitative study concludes that Tamil women in the South have more mobility and authority than women in Uttar Pradesh. Southern states also performed better in terms of the Rural Human Development Index - a weighted average of expenditures, literacy, formal education, life expectancy, and infant mortality rate - developed by the Planning Commission (GOI 2002). Shariff's study also observes higher level of human development in the south (Shariff 1999). However, these analyses have not rigorously tested for different levels gender discrimination between northern and southern states. Nor has there been an attempt to examine this research issue in the context of educational disparities.

social class, religion or gender. This will result in disparity levels, even if the two groups are similar with respect to social and economic characteristics. It is not easy to segregate the individual effects of discrimination and differences in socio-economic characteristics in explaining group (in this case, gender) differences in outcome. However, certain econometric techniques have emerged that address this issue. These techniques identify the residual effect - after taking into account the effect of explanatory variables in explaining disparity levels between groups - with discrimination (Blinder 1973; Fairlie 2005; Oaxaca 1973).

This paper examines regional variations in gender differences in the probability of completing school education,³ and employs econometric tests to examine whether such differences are indeed greater in North India. Having established the regional pattern of disparity, we will then proceed to estimate the contribution of the residual effect in explaining gender gap in regional outcomes.

Although gender disparity and discrimination is present in different spheres, this paper focuses specifically on education because of its importance in human development and as a determinant of the quality of life. The importance of education in economic growth (Schultz 1961) and human development (Sen 1985, 1993) has been widely recognized. Recently, the Government of India has made the right to education a fundamental right, under its Constitution, of every child. However, the focus of policy makers and researchers generally has been on the primary level. The four-five years of education imparted as primary education is undoubtedly useful in ensuring the functional literacy of recipients of such education. Despite the importance of functional literacy, economic returns to primary education – in terms of increasing probability of securing work, getting better jobs or bargaining for higher wages – is much less. In comparison, completion of schooling marks an important landmark in the educational career, and makes children better equipped to fend for themselves in the labour market.

³ In India, this consists 12 years of schooling.

The paper is based on unit level data from the quinquennial survey on Employment and Unemployment (2004-05) undertaken by National Sample Survey Organization. The survey was spread over 7,999 villages and 4,602 urban blocks covering 1,24,680 households (79,306 in rural areas and 45,374 in urban areas) and enumerating 6,02,833 persons (3,98,025 in rural areas and 2,04,808 in urban areas). A two-stage stratified sample design, with census villages and urban blocks as the first-stage units for the rural and urban areas respectively, and households as the second-stage units was adopted. The fieldwork for the survey was handled by the Field Operations Division of NSSO.

2. Context, objective and methodology

2.1 Determinants of educational attainments

The literature on socio-economic determinants of educational attainments has mainly focused on enrolment and primary education. Generally employing limited dependent regression models, studies have identified factors like family income or wealth, parental education, empowerment and education of mother, credit constraints, age of the child, family size or presence of siblings, caste affiliations, place of residence and educational infrastructure as determinants of enrolment and primary school completion rates (Akhtar 1996; Deolalikar 1997; Tansel 1998; Brown and Park 2002; Connelly and Zheng 2003; Boissiere 2004; Desai and Kulkarni 2008; SIS/DPP 2005; Okumu et al 2008; Husain and Chatterjee 2009). These studies have also found the presence of strong gender differences.

In India, the education of girls has historically lagged behind that of boys (Aggarwal 1987; Agrawal and Aggarwal 1994). In addition studies have shown that certain communities and classes fare much worse than the others. Though some researchers have recently attempted to lay down the determinants of the inequality in educational attainment for boys and girls, only a handful of these (Bandopadhyay and Subrahmaniam 2008; Das and Mukherjee 2007, 2008; Sengupta and Guha 2002; Raju 1991; Burney and Irfan 1991), have explicitly looked at the factors responsible for the relative gender inequality in educational attainment. But none of these works have examined variations in gender discrimination over regions.

Similarly, Vaid's analysis of trends in gender discrimination across the schooling career of children finds that transition probabilities of girls increase, relative to that of boys, at higher levels of education. Although she finds that locality specific effect decreases at higher levels, except for Rajasthan, variation in gender discrimination across regions is not examined. In a similar study, Husain and Sarkar (2010) found that, on an average, gender disparity is lower across educational levels in southern states. Econometric analysis revealed that, after controlling for socio-economic characteristics, gender disparity decreases. However, region-specific effects were not incorporated in the econometric model.

The finding that gender disparity reduces at higher levels of education is interesting. In fact, Husain and Sarkar (2010) found a reversal of gender disparity at the secondary and higher secondary levels in several states. Unfortunately, gender disparities at higher levels of education have been rarely examined in the Indian context. Hasan and Mehta's study (2006) of college education focuses on disparities across social castes, but ignores gender dimensions, as also does Sundaram (2006). Thorat (2006) notes gender differences in access to higher education but does not look at regional variations. Overall, studies have tended to neglect the study of regional dimensions in gender disparities at higher levels of education. This lacuna forms the motivation of this paper.

2.2 Research hypothesis and methodology

The hypothesis being tested in this paper is that gender disparities in school completion rates is higher in North Indian states, compared to the rest of India. However, instead of using a binary classification, we have grouped Indian states into three groups – Northern states (comprising of Jammu & Kashmir, Punjab, Haryana, Himachal Pradesh, Rajasthan, Chandigarh, Bihar, Jharkhand, Uttar Pradesh, Chattisgarh, Madhya Pradesh, Uttaranchal), Eastern states (consisting of West Bengal, Assam, Orissa, Tripura, Mizoram, Meghalaya, Nagaland, Sikkim) and Southern states (including Kerala, Karnataka, Andhra Pradesh, Maharashtra, Gujarat, Tamil Nadu, Pondicherry, Goa, Dadra, Nagar Haveli, Andaman & Nicobar).

The paper uses a Disparity Index suggested by Sopher (1980), and modified by Kundu and Rao (1986). The index measures disparity between two groups in their possession of a particular property (in this case completion of school education) in terms of the logarithm of the odds ratio - that is, the ratio of the odds that any member of one group (male) has completed school to the odds that any member of the other group (female) does. In brief, if p and q are the probabilities of males and females completing school respectively, then the disparity index (DI) is given by:

$$DI = \log \left[\frac{p^*(1-q)}{q^*(1-p)} \right].$$

The objective of taking log is to reduce the leveling off effect (states with high levels of attainments may show a lower level of disparity than states with low levels of attainments even though the gender gap is the same for both states) (Sopher 1980).

Kundu and Rao (1986) have shown that the above index fails to satisfy the additive monotonicity axiom.⁴ They have, therefore, proposed a modification to this Index as follows:

$$DI_{KR} = \log \left[\frac{p^*(2-q)}{q^*(2-p)} \right].$$

Based on this Index we examine regional variations in the probabilities of completing school education. Cohort-wise analysis is also undertaken to get an idea of the trend in disparities across regions.

The descriptive analysis is followed by an econometric analysis that seeks to identify the factors causing discrimination. A logit model is used for this purpose as the dependant variable is binary (whether respondent has completed schooling or not). Based upon the literature cited in Section 2.1, as well as availability of data from the NSSO database, we hypothesize that – apart from gender and regional location - the probability of completing school depends upon personal traits (age and socio-religious identity), household characteristics (place of residence, household size, expenditure levels, and sex and

⁴ The additive monotonicity axiom specifies that if a constant is added to all observations in a non-negative series, *ceteris paribus*, the inequality index must report a decline.

educational level of household head). These variables are included in the regression model. Now, the gender and region variables are incorporated in the initial model as dummy variables. Statistically significant coefficients of the gender and regional dummies will indicate the presence of gender discrimination and regional effects.

However, as our objective is to examine regional variations in gender differences, we re-estimated the regression for each individual region. The statistical significance and signs of the coefficients will help to establish the validity of Dyson and Moore's hypothesis, viz., that gender disparity is more accentuated in the north.

2.3 Measuring discrimination

Now this analysis merely establishes whether the difference in outcome is significant after controlling for socio-economic characteristics. We also have to measure the contribution of discrimination in explaining the observed gender disparities.

Oaxaca (1973) and Blinder (1973) have shown that the difference in outcomes between two groups may be attributed to differences in explanatory variables or endowments (referred to as *explained component*) and differences in coefficients of explanatory variables (referred to as *residualy* or *unexplained component*). The latter is commonly accepted as a measure of discrimination in literature. Their work has resulted in the development of a methodology to estimate the contribution of discrimination in explaining disparities in outcome. Having established regional patterns in gender disparity, we next attempt to estimate how *the extent of discrimination* varies across regions. This may be undertaken as follows.

For the regression model:

$$y = \alpha + \beta x \quad [1]$$

estimated after pooling a superior group (in terms of having a 'better' outcome, denoted by S) and an inferior group (in terms of having a relatively 'worse' outcome, denoted by W), the difference in mean outcomes can be decomposed as follows:

$$\bar{y}^S - \bar{y}^W = \Delta x \beta^W + \Delta \beta x^S, \quad [2]$$

$$\text{or, } \bar{y}^S - \bar{y}^W = \Delta x \beta^S + \Delta \beta x^W. \quad [3]$$

This method has been used to measure gender ‘discrimination’ in educational attainments (Kingdon 2002). A generalized form of the decomposition for a multi-variate case is:

$$\bar{y}^S - \bar{y}^W = \Delta X [D \beta^W + (I - D) \beta^S] + \Delta \beta [X^S D + X^W (I - D)] \quad [4]$$

when I is the identity matrix and D is the matrix of weights. It is easy to see that [2] and [3] are special cases of D equal to I and 0 respectively. In addition, alternative weights have been suggested by researchers.⁵

Now the above decomposition is based on the relation: $\bar{y} = \hat{\beta} \bar{x}$. Unfortunately, this does not hold for discrete choice models – predicted probability evaluated at means of the independent variables is not necessarily equal to the proportion of ones. Instead we have the relation that average value of the dependant variable equals the average values of predicted probabilities in the sample.⁶ Therefore, Fairlie (1999, 2005) argues, for the non-linear model $Y = F(X \hat{\beta})$, the decomposition may be written as:

$$\bar{Y}^S - \bar{Y}^W = \left[\sum_{i=1}^{N^S} \frac{F(X_i^S \hat{\beta}^S)}{N^S} - \sum_{i=1}^{N^W} \frac{F(X_i^W \hat{\beta}^S)}{N^W} \right] + \left[\sum_{i=1}^{N^W} \frac{F(X_i^W \hat{\beta}^S)}{N^W} - \sum_{i=1}^{N^W} \frac{F(X_i^W \hat{\beta}^W)}{N^W} \right] \quad [5]$$

While [5] corresponds to [2], the equivalent to [3] in the non-linear case is:

$$\bar{Y}^S - \bar{Y}^W = \left[\sum_{i=1}^{N^S} \frac{F(X_i^S \hat{\beta}^W)}{N^S} - \sum_{i=1}^{N^W} \frac{F(X_i^W \hat{\beta}^W)}{N^W} \right] + \left[\sum_{i=1}^{N^S} \frac{F(X_i^S \hat{\beta}^S)}{N^S} - \sum_{i=1}^{N^S} \frac{F(X_i^S \hat{\beta}^W)}{N^S} \right] \quad [6].$$

Again the difference lies in the weighting system, with the alternative weighting systems suggested by Cotton (1988), Reimers (1983), Oaxaca and Ransom (1994), and Neumark (1988) being applicable here also.

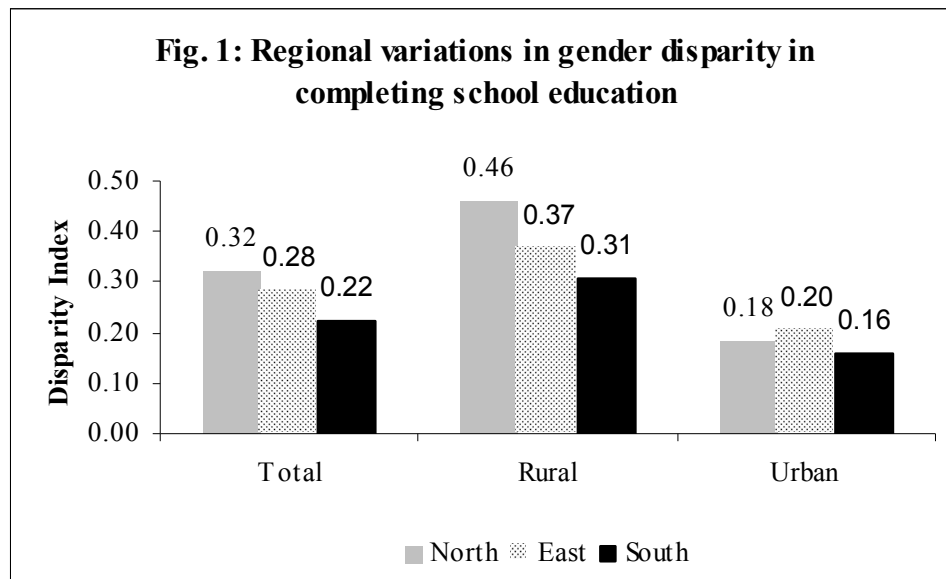
3. Regional patterns in disparity levels

At the all-India level, the probability of a girls completing schooling is 0.11, compared to 0.20 for a boy. This implies a disparity of 0.30. Predictably, disparity is higher in rural

⁵ Cotton (1988) suggests that weights should be mean of the coefficient vector, Reimers (1983) argues that weights should reflect proportion of the two groups, while Neumark (1988) and Oaxaca and Ransom (1994) opt for coefficients estimated from pooled sample.

⁶ This holds exactly for the logit model, which is another reason for preferring logit models to probit models. In the latter, the relation does not hold exactly, but approximates the relation as an empirical regularity (Fairlie, 2005: 307).

areas (0.39), relative to urban areas (0.18). What is interesting, however, is the regional variation in gender disparity in completion of school education. Fig. 1 reveals that gender disparity is higher in Northern states, relative to the rest of India. Further, disparity is lowest in South India. This is also true for rural areas. In urban areas, variations in disparity levels are lower than in rural areas. The interesting finding is that, it is in *Eastern* states that girls lag behind boys to a greater extent than in Northern or Southern states.



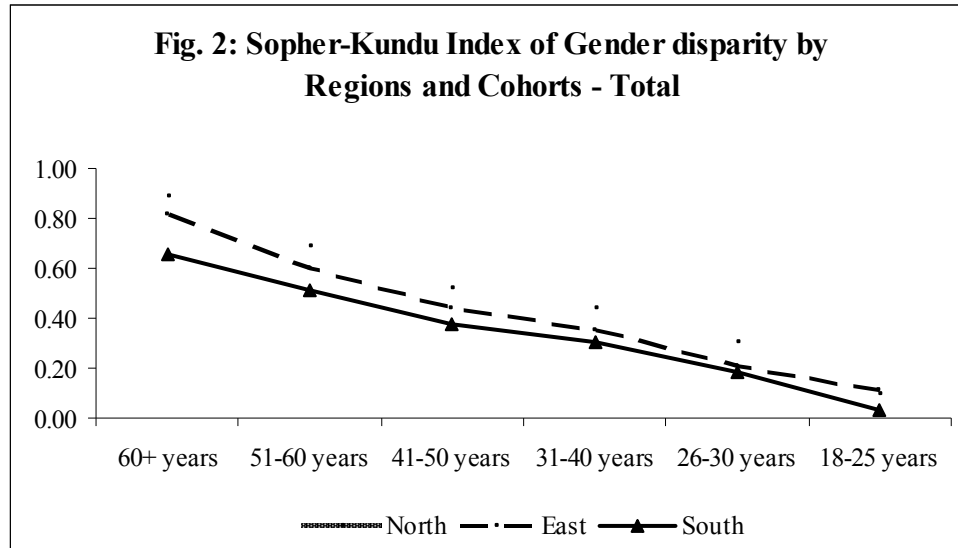
3.1 Cohort-wise analysis

Now these estimates were based on a sample containing several generations. It would be interesting to decompose the sample by generations, and study trends in disparity over time. The sample is therefore divided into the following groups – 18-25 years, 26-30 years, 31-40 years, 4-50 years, 51-60 years and 61 years and above.⁷

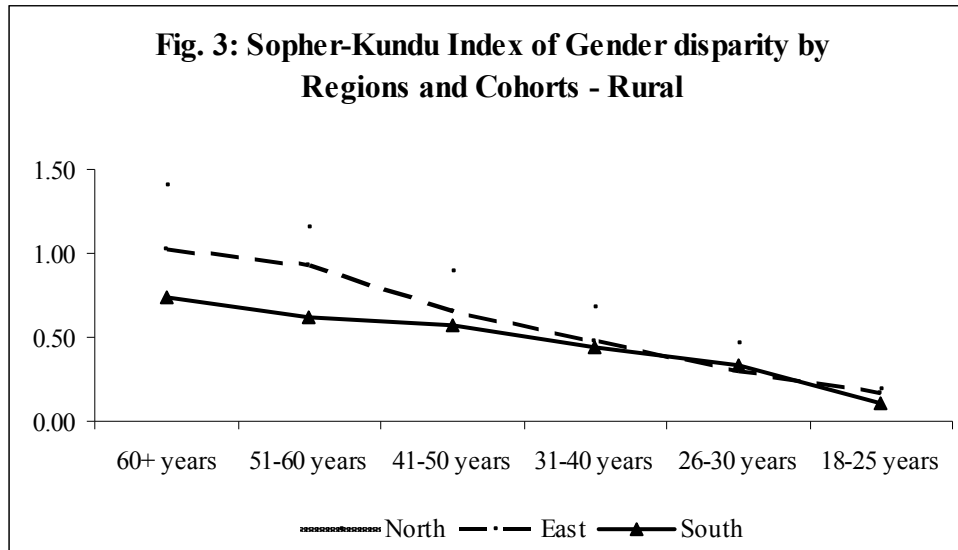
In Fig. 2 we present trends in disparity across the generations for the total population. We can see that disparity levels have fallen in all regions. This is consistent with Husain and Sarkar's finding that disparity levels has fallen across all educational levels at the all-India level (Husain and Sarkar 2010). Disparity in the South has traditionally remained

⁷ The 18-25 years group has been formed to maintain parity with subsequent econometric analysis.

lower than in other regions. In North, on the other hand, disparity levels have always been relatively higher than the rest of India. But the gap between East and North is decreasing and, for the ‘current’ generation (18-26 year group), differences are marginal.

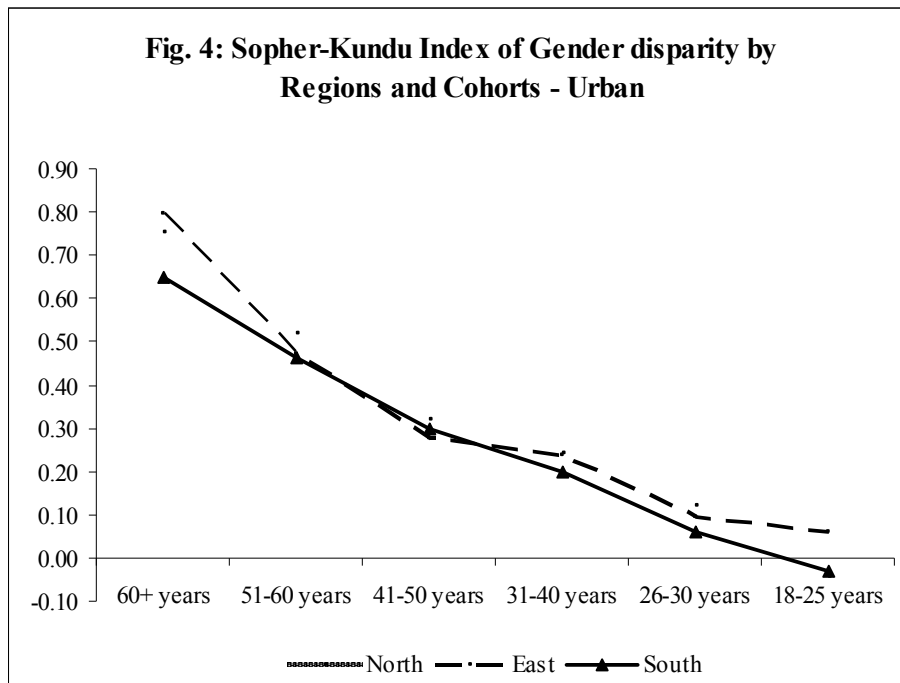


Similar trends are observed in rural areas, though a faster convergence rate is seen (Fig. 3).



In urban areas, on the other hand, regional differences have always been marginal (Fig. 4). Although Southern India has tended to display lower disparity levels, the performance of North India is better for the ‘current’ generation. Interestingly, we observe a negative value of the disparity index for these two regions, implying that girls ‘out-perform’ boys. Eastern states, on the other hand, have performed somewhat erratically, starting off with

higher levels of disparity, improving thereafter, but then falling behind even Northern states.



3.2 State-wise analysis

While the regional result is interesting in itself, we should also examine whether disaggregative state-wise analysis reveals any variation *within* regions. The results of the state-wise analysis are presented in Table 1.

In rural areas of Northern India, Punjab and Himachal Pradesh has low levels of gender disparity, comparable to that of even Southern states. This contrasts with substantial levels of disparities observed in Rajasthan, Bihar, Jharkhand, Madhya Pradesh, Chattisgarh and Uttar Pradesh. In urban areas disparities are low states like Himachal Pradesh, Haryana and Chandigarh. On the other hand, states like Bihar, Rajasthan, Jharkhand, Chattisgarh, and Bihar exhibit high levels of disparity.

State-wise variation is less marked in Eastern states. An exception is rural West Bengal, which has a disparity level of 0.48.⁸ This is surprising, given the long period of rule by a

⁸ Dropping, West Bengal, the Disparity Index for the remaining Eastern states is 0.35.

coalition of Leftist parties and the impressive record of land reforms in the state. Cohort-wise analysis, however, reveals that gender disparities were extremely high after Independence (possibly because of the migration patterns after partition), fell gradually since then, with a sharp fall in the 1980s – when the positive effect of land reforms would take place with a lag.

Table 1: Probabilities of completing schooling and disparity between gender across State

Region/State	Rural			Urban		
	Male	Female	SK Index	Male	Female	SK Index
North	0.151	0.055	0.46	0.322	0.223	0.18
Jammu & Kashmir	0.137	0.051	0.45	0.252	0.213	0.08
Himachal Pradesh	0.168	0.103	0.23	0.411	0.366	0.06
Punjab	0.146	0.100	0.17	0.311	0.306	0.01
Chandigarh	0.228	0.135	0.25	0.487	0.420	0.08
Uttaranchal	0.195	0.096	0.33	0.405	0.310	0.14
Haryana	0.157	0.069	0.38	0.293	0.245	0.09
Delhi	0.330	0.182	0.30	0.418	0.348	0.10
Rajasthan	0.118	0.020	0.79	0.282	0.145	0.32
Uttar Pradesh	0.169	0.057	0.49	0.291	0.199	0.19
Bihar	0.153	0.026	0.79	0.326	0.126	0.46
Jharkhand	0.099	0.026	0.60	0.320	0.169	0.32
Chhattisgarh	0.189	0.055	0.57	0.378	0.217	0.28
Madhya Pradesh	0.134	0.039	0.56	0.327	0.206	0.23
East	0.131	0.058	0.37	0.292	0.194	0.20
Sikkim	0.097	0.060	0.21	0.175	0.169	0.02
Arunachal Pradesh	0.142	0.061	0.39	0.394	0.221	0.30
Nagaland	0.286	0.121	0.41	0.430	0.309	0.18
Manipur	0.203	0.094	0.36	0.377	0.216	0.28
Mizoram	0.133	0.059	0.37	0.206	0.167	0.10
Tripura	0.099	0.045	0.35	0.292	0.190	0.21
Meghalaya	0.071	0.039	0.27	0.346	0.250	0.16
Assam	0.114	0.048	0.39	0.347	0.220	0.23
West Bengal	0.120	0.041	0.48	0.278	0.187	0.19
Orissa	0.120	0.056	0.34	0.258	0.149	0.27
South	0.146	0.075	0.31	0.279	0.201	0.16
Gujrat	0.136	0.059	0.38	0.276	0.194	0.17

Region/State	Rural			Urban		
	Male	Female	SK Index	Male	Female	SK Index
Daman & Diu	0.269	0.113	0.41	0.491	0.236	0.39
Dadra & Nagar Haveli	0.120	0.056	0.35	0.375	0.287	0.14
Maharashtra	0.185	0.072	0.44	0.328	0.245	0.15
Andhra Pradesh	0.112	0.033	0.55	0.275	0.139	0.33
Karnataka	0.131	0.051	0.42	0.256	0.177	0.18
Goa	0.265	0.207	0.12	0.346	0.293	0.09
Lakshadweep	0.090	0.000	-	0.126	0.077	0.23
Kerala	0.171	0.170	0.00	0.226	0.212	0.03
Tamil Nadu	0.132	0.074	0.26	0.265	0.194	0.15
Pondicheri	0.192	0.076	0.48	0.213	0.155	0.15
Andaman & Nicobar	0.135	0.124	0.04	0.300	0.311	-0.02

Source: Estimated from unit level NSS data, 61st Round, 2004-2005.

Southern states, too, exhibit some variation in disparity levels. While the record of Kerala is remarkable – it has very low levels of disparity in urban areas, and *no* disparity in rural areas – disparity levels are high in specific areas. Rural areas of Andhra Pradesh, Maharashtra and Karnataka, along with urban Andhra Pradesh, are found to display relatively high levels of disparity. In fact, disparity levels in Andhra Pradesh are substantially higher than the average disparity in North India.

Thus, the picture for gender disparity observed for school education is more complex than the over-simplified hypothesis formed on the basis of Dyson and Moore's paper. There are considerable variations within regions, and occasionally even between rural and urban areas of the same state (Madhya Pradesh and Uttar Pradesh are examples).

3.3 Regional variation across correlates

Now the differences in gender disparity across regions may be partly explained by differences in socio-economic structure. Table 2 indicates, for instance, smaller sized families and higher urbanization levels in the South. Differences in share of socio-religious communities may also be observed between the South and North. It is necessary, therefore, to examine variations in disparity levels across regions after decomposing the sample by the socio-economic correlates.

Table 2: Variations in socio-economic characteristics over regions

Groups	North	East	South	Total
Monthly per capita expenditure groups				
BPL HHs	21.0	18.0	18.4	19.4
DBPL HHs	51.4	52.3	48.9	50.8
Affluent HHs	27.6	29.7	32.7	29.9
Total	182,856	109,391	152,865	445,112
Household size				
0-3 members	13.7	16.9	23.1	17.7
4 members	13.5	18.6	22.4	17.8
5 members	17.1	20.7	19.4	18.8
6 members	15.3	15.9	13.5	14.8
7-10 members	30.1	23.4	17.4	24.1
More than 10 members	10.3	4.5	4.4	6.8
Total	182,856	109,391	152,865	445,112
Socio-religious identity				
Muslims	12.4	11.8	11.2	11.8
H-SC	5.9	8.7	5.5	6.4
H-ST	17.0	12.3	12.8	14.4
H-Others	56.0	39.3	63.7	54.5
All Others	8.72	27.99	6.89	12.83
Place of residence				
Rural	71.1	72.7	58.5	67.2
Urban	29.0	27.3	41.5	32.9
Total	182,856	109,391	152,865	445,112

Source: Estimated from unit level NSS data, 61st Round, 2004-2005.

Note: [1] BPL is acronym for Below Poverty Line households, while DBPL stands for Households below Double Poverty Line. The justification for taking DBPL is that these households are also targeted in some Government programmes.

[2] Planning Commission poverty lines for each state has been taken. See

Table 3 presents the results of bi-variate analysis. Once again, the results challenges our starting proposition, viz. that disparity levels are greater in North India. Gender disparity levels are higher in Eastern India, than in Northern India, among affluent households, households with 4-5 members, Muslim households and in urban areas. South displays lowest disparity levels in all cases.

Table 3: Regional variation in disparity across some socio-economic correlates

Socio-economic correlates	North			East			South		
	Male	Female	SK Index	Male	Female	SK Index	Male	Female	SK Index
Monthly per capita expenditure groups									
BPL HHs	0.077	0.026	0.49	0.058	0.031	0.28	0.064	0.036	0.26
DBPL HHs	0.160	0.067	0.40	0.118	0.059	0.31	0.129	0.069	0.28
Affluent HHs	0.380	0.224	0.27	0.346	0.198	0.28	0.385	0.266	0.19
Household size									
0-3 members	0.216	0.106	0.34	0.198	0.110	0.28	0.241	0.139	0.26
4 members	0.235	0.135	0.26	0.193	0.100	0.31	0.223	0.150	0.19
5 members	0.208	0.115	0.28	0.167	0.088	0.30	0.193	0.125	0.20
6 members	0.192	0.100	0.31	0.159	0.084	0.29	0.175	0.108	0.22
7-10 members	0.179	0.086	0.34	0.160	0.096	0.24	0.168	0.102	0.23
More than 10 members	0.220	0.081	0.47	0.184	0.099	0.29	0.171	0.089	0.30
Socio-religious identity									
Muslims	0.127	0.059	0.35	0.103	0.041	0.42	0.140	0.090	0.20
H-SC	0.108	0.028	0.61	0.089	0.035	0.42	0.109	0.051	0.34
H-ST	0.107	0.037	0.48	0.094	0.038	0.41	0.121	0.073	0.23
H-Others	0.258	0.131	0.32	0.239	0.139	0.26	0.229	0.137	0.24
All Others	0.209	0.149	0.16	0.180	0.100	0.28	0.283	0.241	0.08
Place of residence									
Rural	0.151	0.055	0.46	0.131	0.058	0.37	0.146	0.075	0.31
Urban	0.322	0.223	0.18	0.293	0.194	0.20	0.279	0.201	0.16

Source: Estimated from unit level NSS data, 61st Round, 2004-2005.

Note: [1] BPL is acronym for Below Poverty Line households, while DBPL stands for Households below Double Poverty Line. The justification for taking DBPL is that these households are also targeted in some Government programmes.

[2] Planning Commission poverty lines for each state has been taken. See

To sum, up, gender disparities in the probability of completing school education is lowest in Southern states. Comparison between Northern and Eastern states, however, does not reveal any clear picture. While disparity levels *tend* to be higher in North India, it is higher in East India for specific socio-economic groups. To get a clearer picture, therefore, we turn to an econometric analysis.

4. Econometric analysis

Since disparity is estimated for region/state, while our unit of analysis is individual child, we have to abandon the Sopher-Kundu index and use a different method to test for gender disparity. We therefore regress probability of completing school education upon gender of the child, controlling for individual and household traits like age of child, socio-religious identity, monthly per capita expenditure groups, household size, place of residence and region. Gender of the child is a dummy and we use boys as the reference category. A statistically significant coefficient of this dummy variable indicates the presence of gender disparity, while a negative sign indicates that girls have a lower probability of completing school than boys. Apart from this main regression, we have also estimated separate models for rural and urban areas.

4.1 Gender disparity in schooling

The results of this basic model are given in Table 4. Note that we have reported odd ratios and not coefficients. Thus, odd ratios lower (higher) than unity corresponds to a negative (positive) coefficient. For each of the models reported in Table 4, the odd ratios for girls are statistically significant and lower than unity. This indicates the presence of gender disparities in completing school education, even after controlling for socio-economic traits. Predictably, gender differences are higher in rural areas.

Table 4: Results of Logit Regression Model – All India

Independent variables	Model 1: Total		Model 2: Rural		Model 3: Urban	
	Odds Ratio	z	Odds Ratio	z	Odds Ratio	z
Male (RC)	1.00		1.00		1.00	
Female	0.45	-83.96	0.36	-75.23	0.56	-41.66
Age Cohort	0.72	-106.10	0.69	-83.42	0.74	-66.46
Muslims (RC)	1.00		1.00		1.00	
H-SC	1.31	9.14	1.15	3.72	1.50	7.81
H-ST	1.15	6.38	1.26	7.42	0.96	-1.11
H-Others	2.30	48.94	2.09	29.49	2.45	38.58
All Others	1.62	23.41	1.47	13.28	1.81	20.23
Household size	1.06	35.19	1.06	28.44	1.06	22.56
Per capita exp. group	4.01	172.94	3.66	110.38	4.29	129.56

Independent variables	Model 1: Total		Model 2: Rural		Model 3: Urban	
	Odds Ratio	z	Odds Ratio	z	Odds Ratio	z
Rural (RC)	1.00					
Urban	3.89	140.14				
East (RC)	1.00		1.00		1.00	
North	1.14	10.15	1.01	0.49	1.38	16.46
South	0.99	-0.41	1.09	4.63	0.96	-2.06
Observations	445111		298904		146207	
LR χ^2	77204.19	0.00	29869.26	0.00	32497.06	0.00
Pseudo R ²	0.20		0.15		0.20	

Note: Per capita expenditure groups have been preferred to absolute levels of per capita expenditure as expenditure is a time variant variable. We require expenditure when the respondent was a student, while NSS reports *current* expenditure levels. Our assumption is expenditure may change since the respondent was a student, but the family remains within the broad expenditure group as before.

RC indicates reference category for dummy variable.

Regional variations in probability of completing school are somewhat surprising. At the all-India level, a child from Northern India has a higher probability of completing school than counterparts from the South or East. Differences in the probability of completing school education between a Southern and Eastern child is statistically insignificant. In rural areas, however, a child from North and East India are equally likely to complete schooling, while a child from Southern India has an advantage over both. In urban areas, on the other hand, North Indian children are better off than East Indian children, while South Indian children have a lowest probability of completing school. It should be emphasized that we are not referring to gender disparity, but to *all* children.

Among other important results are: children from urban areas, and those belonging to younger age cohorts are more likely to complete school. Children from Forward Caste Hindus, Scheduled Tribe Hindus, OBC Hindus and All Others are more advantaged than Muslim children. These are expected results (refer citations in Section 2.1). However, the absence of difference between Muslim and Scheduled Tribe children (in urban areas) is surprising.

4.2 Variations in disparity across regions

While Models 1-3 indicated the presence of gender disparity, our research hypothesis was that this disparity is higher in northern states. This was not addressed in Section 4.1. To

test this hypothesis we have to study differences in probabilities for each region. We have therefore estimated Model 1-3 for each region separately, reporting results for rural and urban areas in each region in the Appendix. Table 5 states the regression results using an urban dummy.

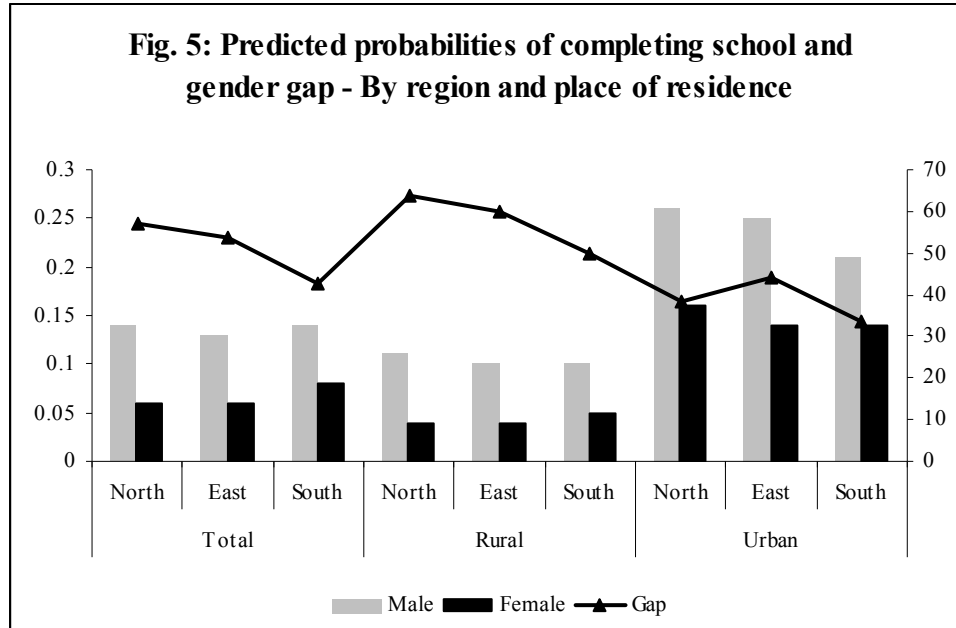
Table 5: Results of Logit Regression Model – Region-wise, Rural + Urban

Independent variables	Model 4: North		Model 5: East		Model 6: South	
	Odds Ratio	z	Odds Ratio	z	Odds Ratio	z
Male (RC)	1		1		1	
Female	0.39	-62.16	0.43	-41.81	0.54	-39.25
Age Cohort	0.72	-67.03	0.78	-37.38	0.67	-76.54
Muslims (RC)	1		1		1	
H-SC	1.45	7.66	1.12	2.03	1.34	5.67
H-ST	1.07	2.03	1.06	1.20	1.36	8.35
H-Others	2.55	35.63	2.44	23.45	2.04	25.62
All Others	1.72	16.03	1.19	4.43	2.95	29.62
Household size	1.06	28.13	1.07	18.75	1.05	15.37
Per capita exp. group	3.84	108.64	3.99	81.56	4.45	108.43
Rural (RC)	1		1		1	
Urban	4.82	101.05	3.52	62.46	3.26	74.44
Observations	182855		109391		152865	
LR χ^2	34267.29	0.00	16187.37	0.00	28562.54	0.00
Pseudo R ²	0.22		0.18		0.21	

Note: Per capita expenditure groups have been preferred to absolute levels of per capita expenditure as expenditure is a time variant variable. We require expenditure when the respondent was a student, while NSS reports *current* expenditure levels. Our assumption is expenditure may change since the respondent was a student, but the family remains within the broad expenditure group as before.

RC indicates reference category for dummy variable.

Once again the gender dummy is significant and less than unity. But what is important is how the value of the odds ratio fluctuates across regions. It can be seen that girls have a 61 percent lower probability of completing schooling in North India, compared to boys. In South and East India this percentage is 46 and 57, respectively. This implies that – after controlling for socio-economic traits – gender disparity is highest in North India, and lowest in South India. This is also observed for rural areas (Appendix A1). In urban areas, however, the gender gap in probability of completing school is higher in Eastern India, compared to North India



This is summarized in Fig. 5, which depicts regional variations in predicted probabilities of completing school education by place of residence, and the gender gaps in these predicted values between boys and girls. It can be seen that ratio of predicted probabilities is lowest in South India, and highest in North India if we consider rural areas. In urban areas, however, it is in Eastern India where the gender gap is highest.

4.3: Estimating residual effects

Now, the above econometric analysis was implicitly based on the assumption that gender of the child leads to a difference in intercept (captured by the intercept dummy, Female), but not in the regression coefficients. But part of it may be attributed to the gender differences in coefficients of explanatory variables (Blinder 1973; Oaxaca 1973). If this assumption is relaxed, the gender difference in outcomes (viz. probability in completing schooling) may be decomposed into two components – explained (difference attributable to differences in socio-economic characteristics) and unexplained (difference attributable to difference in regression coefficients). The latter, residual, component is often taken to be a measure of discrimination.

In this section we decompose the difference in outcomes to estimate the contribution of the residual effects in explaining the difference in outcome. To check robustness of results, we have used extreme weights ($\Omega = 0$ and $\Omega = 1$).

Table 6: Results of Decomposition Analysis

Place of Residence	Region	Difference in outcomes	Discrimination (Percent)	
			$\Omega = 0$	$\Omega = 1$
Total	North	0.10	94.04	95.88
	East	0.08	103.86	102.14
	South	0.08	87.91	88.08
Rural	North	0.10	99.38	100.80
	East	0.07	102.71	102.69
	South	0.07	91.63	91.94
Urban	North	0.10	93.96	96.16
	East	0.10	104.43	100.23
	South	0.08	89.99	89.18

Note: The user-written module in Stata (st0152_1), written by Sinning, Hahn and Bauer (2008), is used.

The results indicate that, although *disparity* may be greater in Eastern India only for the urban population, the *residual (discrimination) effect* is greater in this region for not only the urban population, but also for the total and rural population. In fact, the residual unexplained effect is found to be consistently greater than 100 per cent in Eastern India. This implies that socio-economic characteristics explain “less than nothing” of the gender disparity. The socio-economic context in Eastern India is actually more favourable for girls, and should have lead to girls having higher probabilities in completing school education, vis-à-vis boys. However, the unexplained effect is so strong that it reverses the situation.

5. Conclusion

In conclusion, this paper raises questions about the common perception that gender discrimination is higher in North India. The picture is more complex than the simple claim that “the country can be roughly divided in two by a line that approximates the contours of the Satpura hill range, extending eastward to join the Chota Nagpur hills of southern Bihar” (Dyson and Moore 1983: 38). While it is true that *disparity* levels in

completion of school education are higher in North India (except in urban areas), the level of *discrimination* may be consistently higher in Eastern states. This is somewhat surprising given the prevalence of matrilinear structure and the influence of missionaries in the North-eastern states. In fact we re-estimated residual effects in Eastern India, after dropping the states of Assam, West Bengal and Orissa, but failed to find any major difference in results.

This finding is contrary to common beliefs about the prevalence of gender discrimination in the so called BIMARU states⁹ and in Punjab and in Haryana. The starting hypothesis was based on studies using demographic indicators. These studies establish that North India has low sex ratios and high incidence of sex selection and female infanticides. These are also important determinants of gender discrimination, and it is important to link such demographic indicators with educational decisions. Why regional variations in gender discrimination in education do not match with the observed regional patterns in demographical indicators is an interesting question. Extending the regression model by incorporating district level demographic characteristics like sex ratios, female literacy rates, age of marriage, etc. is a possible area of extension of this work.

⁹ Bihar, Madhya Pradesh, Uttar Pradesh and Rajasthan, before bifurcation of these states. Now, in addition to these states, BIMARU includes Jharkhand, Chattisgarh and Uttaranchal.

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Appendix
Table A1: Results of Logit Regression Model – Region-wise, Rural

Independent variables	Model 7: North		Model 7: East		Model 9: South	
	Odds Ratio	z	Odds Ratio	z	Odds Ratio	z
Male (RC)	1		1		1	
Female	0.29	-57.57	0.37	-36.17	0.46	-33.02
Age Cohort	0.69	-54.89	0.77	-30.05	0.63	-57.85
Muslims (RC)	1		1		1	
H-SC	1.13	2.05	1.01	0.20	1.35	4.24
H-ST	1.19	3.89	1.12	1.79	1.50	6.77
H-Others	2.17	20.68	2.29	17.80	1.97	13.68
All Others	1.30	5.67	1.21	3.96	3.32	19.88
Household size	1.06	23.56	1.07	13.91	1.04	9.94
Per capita exp. group	3.30	68.34	4.06	58.58	4.12	64.91
Observations	129911		79521		89472	
LR χ^2	13505.32	0.00	7162.58	0.00	10354.74	0.00
Pseudo R ²	0.16		0.14		0.17	

Table A2: Results of Logit Regression Model – Region-wise, Urban

Independent variables	Model 10: North		Model 11: East		Model 12: South	
	Odds Ratio	z	Odds Ratio	z	Odds Ratio	z
Male (RC)	1		1		1	
Female	0.53	-27.55	0.52	-21.83	0.62	-22.83
Age Cohort	0.74	-39.74	0.80	-22.34	0.70	-50.61
Muslims (RC)	1		1		1	
H-SC	1.88	7.28	1.52	3.80	1.20	2.15
H-ST	0.80	-4.36	1.00	-0.03	1.20	3.74
H-Others	2.81	28.23	2.70	14.89	2.07	21.41
All Others	2.48	17.87	1.18	2.40	2.65	20.91
Household size	1.07	16.25	1.08	12.61	1.05	11.70
Per capita exp. group	4.33	80.85	3.92	56.02	4.60	85.07
Observations	52944		29870		63393	
LR χ^2	13833.79	0.00	5431.01	0.00	13928.55	0.00
Pseudo R ²	0.22		0.16		0.20	