

Lauridsen, Jørgen; Pradhan, Jalandhar

Article

Socio-economic inequality of immunization coverage in India

Health Economics Review

Provided in Cooperation with:

Springer Nature

Suggested Citation: Lauridsen, Jørgen; Pradhan, Jalandhar (2011) : Socio-economic inequality of immunization coverage in India, Health Economics Review, ISSN 2191-1991, Springer, Heidelberg, Vol. 1, Iss. 11, pp. 1-6,
<https://doi.org/10.1186/2191-1991-1-11>

This Version is available at:

<https://hdl.handle.net/10419/77832>

Standard-Nutzungsbedingungen:

Die Dokumente auf EconStor dürfen zu eigenen wissenschaftlichen Zwecken und zum Privatgebrauch gespeichert und kopiert werden.

Sie dürfen die Dokumente nicht für öffentliche oder kommerzielle Zwecke vervielfältigen, öffentlich ausstellen, öffentlich zugänglich machen, vertreiben oder anderweitig nutzen.

Sofern die Verfasser die Dokumente unter Open-Content-Lizenzen (insbesondere CC-Lizenzen) zur Verfügung gestellt haben sollten, gelten abweichend von diesen Nutzungsbedingungen die in der dort genannten Lizenz gewährten Nutzungsrechte.

Terms of use:

Documents in EconStor may be saved and copied for your personal and scholarly purposes.

You are not to copy documents for public or commercial purposes, to exhibit the documents publicly, to make them publicly available on the internet, or to distribute or otherwise use the documents in public.

If the documents have been made available under an Open Content Licence (especially Creative Commons Licences), you may exercise further usage rights as specified in the indicated licence.



<http://creativecommons.org/licenses/by/2.0/>

RESEARCH

Open Access

Socio-economic inequality of immunization coverage in India

Jørgen Lauridsen^{1*} and Jalandhar Pradhan²

Abstract

To our knowledge, the present study provides a first time assessment of the contributions of socioeconomic determinants of immunization coverage in India using the recent National Family Health Survey data. Measurement of socioeconomic inequalities in health and health care, and understanding the determinants of such inequalities in terms of their contributions, are critical for health intervention strategies and for achieving equity in health care. A decomposition approach is applied to quantify the contributions from socio-demographic factors to inequality in immunization coverage. The results reveal that poor household economic status, mother's illiteracy, per capita state domestic product and proportion of illiterate at the state level is systematically related to 97% of predictable socioeconomic inequalities in full immunization coverage at the national level. These patterns of evidence suggest the need for immunization strategies targeted at different states and towards certain socioeconomic determinants as pointed out above in order to reduce socioeconomic inequalities in immunization coverage.

JEL Classification: I10, I12

Keywords: health inequality, immunization, India, decomposition, socio-economic inequality

Background

The distributive dimension of health or health inequality has become prominent on global health policy agenda, as researchers have come to regard average health status as an inadequate summary of country's health performance [1]. Socioeconomic inequalities in child health are a major concern in developing countries to achieve the Millennium Development Goals set forth by the United Nations [2]. Yet progress towards achieving goals in reducing socioeconomic inequalities in child health may have been stymied by a critical gap in documenting and understanding trends in socioeconomic inequality in child health indicators particularly in less developed countries (endnote a). While many cross sectional studies have been performed, relatively little evidence is available regarding how socioeconomic inequalities in health have changed over time as the development process unfolded and levels of urbanization rose, women's educational attainment improved, infrastructure spread, and income and wealth increased;

however, few studies have shown that socioeconomic disparities in health have in fact increased (endnote b).

In developing countries, gaps in health-related outcomes between the rich and the poor are large [3-7]. These gaps limit poor peoples' potential to contribute to the economy by reducing their capacity to function and live life to the fullest - and even to survive. The study of poor-rich inequalities in health status should not, however, solely aim to quantify their magnitude. Research should also aim to identify which population subgroups are the most disadvantaged. For this purpose, it should be possible to identify the determinants of inequalities, including those associated with age, gender, education, occupation etc. These variables have previously been identified as powerful sources of health inequalities in low and middle income countries [8,9].

A growing number of studies have examined inequalities in immunization coverage by household economic status in developing countries like India [10-14]. Many studies have assessed the level of socioeconomic inequalities in health using concentration indices and concentration curve. Though the values of concentration indices (CIs) show the degree of socio-economic inequality, it does not highlight the pathways through

* Correspondence: jtl@sam.sdu.dk

¹Institute of Public Health - Health Economics, University of Southern Denmark, Denmark

Full list of author information is available at the end of the article

which inequality occurs. Decomposition of inequalities is critical to explore pathways of socioeconomic inequalities in child health.

Moreover the full immunization coverage rate has only increased from 71% in 1992 to 80% in 2006 in India (Figure 1). There is a little progress from wave 2 to wave 3 of the National Family Health Survey i.e. period from 1998-99 to 2005-06. Children not fully immunized have just declined by two percentage point i.e. from 58% to 56%. So, an intensive study is required to assess such disappointing progress in full immunization coverage.

To our knowledge, there has been virtually no study that attempted decomposition of health inequalities in the Indian context to understand such pathways. Moreover, this study also considered state level covariates along with household/individual level variables to examine the degree of contribution to the total socio-economic inequality in full immunization coverage. Given the methodological developments and the policy relevance, an attempt has been made in the present study for the first time to decompose health inequalities in terms of immunization coverage in Indian. The objective of this study is two-fold: first is to use a concentration index to quantify the socioeconomic distribution of child not fully immunized and; second is to decompose these inequalities by quantifying the contribution attributable to both household/individual covariates (i.e. economic status, education of mother, caste, residence, birth order and sex of the child) and state specific variables (i.e. poverty ratio, per-capita state domestic product, Income inequality measured in terms of Gini coefficient, % of public health spending of the total health spending, % of illiterate, % of Scheduled Tribe/Scheduled Caste population).

Methods

Similar to previous studies initiated by Wagstaff et al. [15] we use the concentration index as our measure of

relative socioeconomic inequality in immunization coverage. A concentration curve $L(s)$ plots the cumulative proportion of the population (ranked by socioeconomic status (SES), beginning with lowest SES) against the cumulative proportion of children not being fully immunized. If $L(s)$ coincides with the diagonal everyone is equally off. However, if $L(s)$ lies above the diagonal, then inequality in coverage exists and favors those with high SES. The further $L(s)$ lies from the diagonal, the greater the degree of inequality. The concentration index, C , is defined as twice the area between $L(s)$ and the diagonal and takes a value of 0 when everyone is equally off regardless of SES. The minimum and maximum values of C are -1 and +1, respectively; these occur in the (hypothetical) situation where immunization is concentrated in the hand of the least disadvantaged and the most disadvantaged person, respectively. Thus, the larger negative value of C , the more the absence of full immunization concentrates among low SES groups. A computational formula for C , which allows for application of sample weights was given by Kakwani et al. [16] as

$$C = \frac{2}{N\mu} \sum_{i=1}^N w_i y_i R_i - 1, \text{ where } \mu = \frac{1}{N} \sum_{i=1}^N w_i y_i \text{ is the weighted mean of the sample, i.e. the weighted proportion not fully immunized, } N \text{ the sample size, } y_i \text{ an indicator for not being fully immunized, } w_i \text{ the sample weight of the individual (which sums to } N) \text{ and } R_i \text{ the fractional rank defined according to Kakwani et al. as}$$

$$R_i = \frac{1}{N} \sum_{j=1}^{i-1} w_j + \frac{w_i}{2}, \text{ i.e. the weighted cumulative proportion}$$

of the population up to the midpoint of each individual weight. Following the same authors, C can be conveniently computed as the weighted covariance of y_i and R_i , i.e.

$$C = \frac{2}{\mu} \text{cov}_w(y_i, R_i) = \frac{2}{N\mu} \sum_{i=1}^N w_i (y_i - \mu)(R_i - \frac{1}{2}).$$

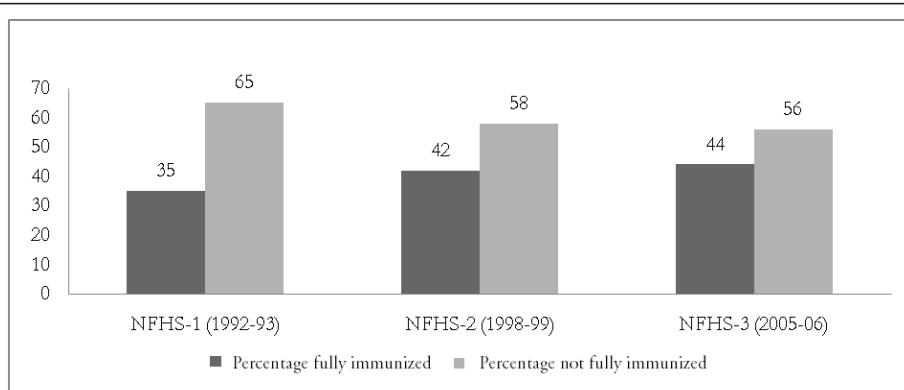


Figure 1 Trend in immunization coverage, India.

A straightforward way of decomposing the predicted degree of inequality into the contributions of explanatory factors was proposed by Wagstaff et al. [17]. Adapting their approach to the present case, where the health indicator is a binary variable and a logit regression specification thus applied, amounts to specifying $l(\hat{p}_i) = \sum_k \beta_k x_{ki}$, where $l(\hat{p}_i)$ is the logit of the predicted probability of not being fully immunized and β_k the logit regression coefficient for the health determinant x_k . Given this linear relationship, the concentration index for $l(\hat{p}_i)$ can be written as $\hat{C} = \sum_k \frac{\beta_k \bar{x}_k}{\hat{\mu}} C_k$, where $\hat{\mu}$ is the mean of $l(\hat{p}_i)$, \bar{x}_k the mean of x_k and C_k the concentration index of x_k (defined analogously to C).

While β_k measures the relationship between the health determinant x_k and the logit $l(\hat{p}_i)$, a more intuitive expression of the relationship between the health determinant and the probability p_i is the marginal effect $m_k = \lambda(\sum_k \beta_k \bar{x}_k) \beta_k$, where $\lambda()$ is the logit density function. Specifically, m_k expresses the average change in the probability of not being fully immunized when the health determinant x_k changes one unit.

In order to assess sampling variability and to obtain standard errors for the estimated quantities, where in particular the concentration indices and the contributions,

i.e. the $\frac{\beta_k \bar{x}_k}{\hat{\mu}} C_k$ parts, cause troubles, we apply a "bootstrap" procedure [18,19] in a five-step manner much similar to van Doorslaer and Koolman [20]: First, sample size is inflated to allow for differences in sampling probability by dividing the sampling weights with the smallest weight and rounding to nearest integer. Second, from this expanded sample a random sub-sample of the size of the original sample is drawn with replacement. Third, the entire set of calculations as specified above are performed on this sample. Fourth, this whole process is repeated 1,000 times, each leading to replicate estimates. Fifth, using the obtained 1,000 replicates, standard deviations and t statistics can be computed.

Data

Data from National Family Health Survey-3, 2005-06 [21] has been used in this study. In addition for state specific covariates data from Census 2001, Central Statistical Organisation and National Sample Survey 61st round on consumer expenditure, 2004-05 [22] are used. The National Family Health Survey-3 collected information on vaccination for all living children born in the five years preceding the survey. Information was collected from mothers for children born since 1 January,

2000 (in states that began fieldwork in 2005) and since 1 January 2001 (in states that began field work in 2006). If a card was available, the interviewer was required to carefully copy the dates on which the child received vaccinations against each disease. For vaccination not recorded on the card, the mother's report that the vaccination was or was not given was recorded. If the mother could not show a vaccination card, she was asked whether the child had received any vaccinations. If any vaccinations had been received, the mother was asked whether the child had received a vaccination against tuberculosis (BCG); diphtheria, whooping cough (pertussis), and tetanus (DPT); poliomyelitis (polio); and measles. For DPT and polio, information was obtained on the number of doses of the vaccine given to the child. Mothers were not asked the dates of vaccinations. To distinguish Polio 0 (polio vaccine given at the time of birth) from Polio 1 (polio vaccine given about six weeks after birth), mothers were also asked whether the first polio vaccine was given just after birth or later.

A binary outcome variable was calculated, namely whether or not each of the live born child aged 12-23 months received all recommended doses of vaccination or not (child fully immunized = 0; child not fully immunized = 1) (endnote c). For the core analysis we considered child not fully immunized as a dependent variable to standardize the interpretation. Two sets of independent variables (household/individual and state specific) are considered for decomposition analysis. The household/individual covariates consist of economic status (poor/non poor), education of mother (illiterate/literate), caste (scheduled caste/tribe (SC/ST)/non scheduled caste/tribe), residence (rural/urban), sex of the child (male/female), and birth order (birth order < 3, birth order 3 or more).

The state specific variables for decomposition analysis included: poverty ratio, per-capita state domestic product, income inequality measured in terms of Gini coefficient, % of public health spending of the total health spending, % of illiterate, and % of scheduled tribe/scheduled caste population.

In the National Family Health Survey-3, an index of economic status (wealth quintile) for each household was constructed using principal components analysis based on data from 109041 households. The wealth quintiles distribution was generated by applying principal components analysis on 33 household assets (endnote d). The wealth quintile distribution was used to determine poor-rich household for subsequent modelling.

For the decomposition analysis, quintiles 1 and 2, and quintiles 3, 4, and 5 were grouped together. This produced a binary variable labelled 'poor economic status',

including households in the bottom 40% of economic status. Mother's education was a categorical variable with the following four levels: illiterate, primary school, guidance/high school, university. For decomposition analysis, mother's illiteracy-a binary variable- was used.

Finally, the decomposition analysis is confined to twelve possible socio-economic determinants including both household/individual and state specific variables that could explain the maximum dimension of socioeconomic inequality particularly in developing countries like India. The predictor variables of interest are i) poor economic status, ii) mother is illiterate, iii) residence in rural area, iv) sex of the child (male), v) birth order of the child (birth order 3 or more) and vi) belong to scheduled caste/scheduled tribe vii) poverty ratio, viii) per-capita state domestic product, ix) income inequality measured in terms of Gini coefficient, x) % of public health spending of the total health spending, xi) % of illiterate, and xii) % of scheduled tribe/scheduled caste population.

To take care of the non-equal probabilities of selection in different domains, a design weight was applied. The national level weight for women is calculated as,

$$W_{wi} = \frac{W_{Di}}{R_{Hi} * R_{Wi}}$$

where W_{Di} is the household design weight for the i^{th} domain is the inverse of the sampling fraction for the i^{th} domain ($f_i = n_i/N_i$); R_{Hi} is the response rate of the household interviewed; R_{Wi} is the response rate of the women interviewed. After adjustment for non response, the weights are normalized so that total number of weighted cases is equal to the total number of un-weighted cases.

Results

Table 1 presents mean values and concentration indices of the variables selected for the study together with regression coefficients and percentage contributions to inequality in immunization of the covariates. From the column of means, it is seen that about 56 percent of the children aged 12-23 months are not fully immunized in India. Furthermore, 47 percent of the children belong to poor household economic status, and a similar proportion of children have mothers who are illiterate. A majority of the children come from rural area (74 percent).

The second column of Table 1 presents concentration indices for both dependent and predictive variables, which provide insights on the poor-rich distributions of immunization and the socio-economic determinants. Thus, the CI value for a child not fully immunized is -0.15021 at the national level which indicates that immunization practice favors children from relatively wealthier families. Furthermore, it is seen that illiteracy of mothers, living in rural areas, belonging to scheduled cast or tribe and high birth order concentrates among the poor.

Estimated marginal effects from the regression analysis are presented in the third column of Table 1. The marginal effects indicate the association between the determinants and child health outcome indicator. The relationship between wealth and immunization coverage is evident, as children from families with poor economic status have a 59 percent higher risk of not being fully immunized. Likewise, being a child of an illiterate mother increases the risk of not being fully immunized

Table 1 Means, concentration indices, marginal effects and contributions of covariates to inequality in immunization (N = 9582)

Variable	Mean ^a	Conc. index	Marginal effect	% Contribution
Child not fully immunized	0.564581	-0.15021***	(dep. var.)	(dep. var.)
Male child	0.532337	0.02277***	-0.13244***	0.4188***
Poor economic status	0.470733	-0.52949***	0.58956***	38.3175***
Mother is illiterate	0.477565	-0.32651***	0.85115***	34.6133***
Rural areas	0.738685	-0.17232***	0.08682***	2.8811***
Belong to Scheduled caste/tribe	0.302047	-0.24674***	0.08780***	1.7066***
Birth order 3 or more	0.407287	-0.20537***	0.35583***	7.7622***
Poverty Ratio	28.96984	-0.06000***	-0.01329***	-6.0251***
Per capita state domestic product	16433.14	0.47684***	-0.0001***	14.3303***
Income inequality (Gini Coeff)	0.329272	0.01622***	0.26994	-0.3760
% of public health spending of the total health spending	16.45547	0.04674***	0.01283***	-2.5741***
% of illiterate	36.66706	-0.05095***	0.02086***	10.1626***
% of Scheduled Caste/Scheduled Tribe population	24.43076	-0.02818***	-0.00678***	-1.2171***

Notes:

^a: Means are weighted with population weights For concentration indices, regression coefficients and contributions, figures significantly different from zero is marked with *** (1 percent level), ** (5 percent level) and * (10 percent level).

The % contribution expresses the contribution in percentage of C.

with 85 percent, while the risks are 8 percent higher for children in rural areas, 35 percent higher for children of birth order 3 or more. Furthermore, percentage of public health spending of total health spending and percentage of illiterate population at the state level are positively related with the child health outcome indicator.

Finally, the last column of Table 1 presents the decomposition analysis of socio-economic inequalities in full immunization coverage. It is seen that the poor household economic status contributes about 38 percent of the total socioeconomic inequalities in child immunization. A major contributor is mother's illiteracy which contributes almost 34 percent to the inequality of immunization. Other important contributors are per capita state domestic product and % of illiterate at the state level which contribute with 14 and close to 10 percent respectively. The result furthermore indicates that public health spending, income inequality and % of scheduled caste and scheduled tribe at the state level play less important role in determining the scale of health inequality in terms of child immunization.

To summarize, most predictable socioeconomic inequalities seem to arise from four socio-economic predictors: poverty itself, illiteracy of mothers, per-capita state domestic product and % of illiterate person at the state level.

Discussion and conclusions

The study presents - to our knowledge - first time evidence on the composition of socioeconomic inequality in child health care in India in terms of children not being fully immunized. Decomposition results reveal that poor household economic status, mother's illiteracy, state domestic product and level of illiteracy at the state level contribute with about 97 percent of the total socioeconomic inequalities in full immunization coverage at the national level. Of these determinants, mother's illiteracy stands out with a contribution of about 34 percent. Furthermore, decomposition analysis of the determinants of health inequalities based on state level data, shows that neither income inequality nor the public share of health spending are significant determinants of health inequalities but per-capita state domestic product and % of illiterate population explains about 24% of the total health inequalities in full immunization coverage.

Policy implications of these results may be that health intervention strategies aiming at reducing socioeconomic inequality in immunization coverage could helpfully benefit from being supplemented with strategies aiming at reducing poverty and illiteracy in particular. Finally, intensive community level analysis is required to

understand the pathways of health inequalities in full immunization coverage at the state level.

Endnotes

a. Numerous studies have examined the effects of socioeconomic status on child health or mortality using cross-sectional data. However, few of them have extended their findings to characterize levels of inequality, using either rate ratios or, especially, more sophisticated measures of inequality. Additional complications of extracting information on trends in socioeconomic inequalities in health from cross sectional studies are that the specific measures of socioeconomic status often differ across studies, as do the number and type of other variables that are held constant [10,5,23].

b. Cleland et al. [24] found that disparities in child survival by socioeconomic status and maternal education did not narrow from the 1970s to the 1980s in a dozen of developing countries. Wagstaff's [6] reanalysis of the result from the number of studies showed that inequality in under-five mortality increased in Bolivia, from 1994 to 1998, in Vietnam from 1993 to 1998 [25], and in Uganda from 1988 to 1995 [26].

c. Fully Immunized involves received BCG, three doses of DPT and Polio, and measles vaccines.

d. The 33 household asset variables are household electrification; type of windows; drinking water source; type of toilet facility; type of flooring; material of exterior walls; type of roofing; cooking fuel; house ownership; number of household members per sleeping room; ownership of a bank or post-office account; and ownership of a mattress, a pressure cooker, a chair, a cot/bed, a table, an electric fan, a radio/transistor, a black and white television, a colour television, a sewing machine, a mobile telephone, any other telephone, a computer, a refrigerator, a watch or clock, a bicycle, a motorcycle or scooter, an animal-drawn cart, a car, a water pump, a thresher, and a tractor.

Author details

¹Institute of Public Health - Health Economics, University of Southern

Denmark, Denmark ²Department of Humanities and Social Sciences, National Institute of Technology, Rourkela, Orissa, India

Authors' contributions

JP carried out the data collection, drafted the study, wrote the background section and contributed to the results section. JL did the statistical analysis, wrote the methods section and contributed to the results section. Both authors read and approved the manuscript.

Competing interests

The authors declare that they have no competing interests.

Received: 14 March 2011 Accepted: 5 August 2011

Published: 5 August 2011

References

1. World Health Organisation: **The World Health Report 2000: Health Systems: Improving Performances.** Geneva: WHO; 2000.
2. Kruk ME, Prescott MR, Pinho H, Galea S: **Equity and the child health Millennium Development Goal: the role of pro-poor health policies.** *J Epidemiol Community Health* 2011, **65**:327-333.
3. Baker JL, van der Gaag J: **Equity in health care and health care financing: evidence from five developing countries.** In *Equity in the Finance and Delivery of Health Care: An International Perspective.* Edited by: van Doorslaer E, Wagstaff A and Rutten F. NY: Oxford University Press; 1993:.
4. Gwatkin DR: **Health inequalities and the health of the poor: What do we know? What can we do?** *Bull the World Health Organ* 2000, **78**:3-17.
5. Leon DA, Walt G: **Poverty, inequality and health: an international perspective.** Oxford: Oxford University Press; 2001.
6. Wagstaff A: **Inequalities in health in developing countries: swimming against the tide?** Washington DC: World Bank. Policy Research Working Paper 2795; 2002.
7. Wagstaff A: **Poverty and health sector inequalities.** *Bull the World Health Organ* 2002, **80**:97-105.
8. Braveman P: **Monitoring Equity in Health: A Policy Oriented Approach in Low and Middle Income Countries.** *Equity Initiative, Paper No. 3 (WHO/CHS/HSS/98.1)* Geneva: WHO; 1998.
9. Starfield B: **State of the art in research on equity in health.** *J Health Polit Policy Law* 2006, **31**:11-32.
10. Wagstaff A: **Socioeconomic inequalities in child mortality: comparisons across nine developing countries.** *Bull World Health Organ* 2000, **78**:19-29.
11. Wagstaff A, Watanabe N: **Socioeconomic inequalities in child malnutrition in the developing world.** *Policy Research Working Paper 2434, Development Research Group* Washington DC: World Bank; 2000.
12. Chandra R, Srivastava VK, Nirupam S: **Impact of urban basic services on immunization coverage in a slum area of northern India.** *Asia Pac J Public Health* 1993, **6**:153-155.
13. Suresh K, Saxena D: **Trends and determinants of immunization coverage in India.** *J Indian Med Assoc* 2000, **98**:10-14.
14. Sekhar C, Jayachandran V: **Immunization coverage in India, 1991-2001: multiple indicator surveys vis-à-vis focused surveys.** In *Human Rights and Social Policies for Children and Women: The Multiple Indicator Cluster Surveys (MICS) in Practice.* Edited by: Alberto M, Enrique D, Marina K. NY: The New School University; 2005:.
15. Wagstaff A, Paci P, van Doorslaer E: **On the measurement of inequalities in health.** *Soc Sci Med* 1991, **33**:545-557.
16. Kakwani NC, Wagstaff A, van Doorslaer E: **Socioeconomic inequalities in health: measurement, computation and statistical inference.** *J Econom* 1997, **77**:87-104.
17. Wagstaff A, van Doorslaer E, Watanabe N: **On decomposing the causes of health sector inequalities, with an application to malnutrition inequalities in Vietnam.** *J Econom* 2003, **112**:219-227.
18. Efron B, Tibshirani RJ: **An Introduction to the Bootstrap.** London: Chapman & Hall; 1993.
19. Deaton A: **The Analysis of Household Surveys: A Microeconometric Approach to Development Policy.** Baltimore: John Hopkins University Press; 1997.
20. Doorslaer E, Koolman X: **Explaining the differences in income related health inequalities across European countries.** *Health Econ* 2004, **13**:609-628.
21. National Family Health Survey: **National Family Health Survey (NFHS-3).** Maryland: International Institute for Population Sciences and ORC Macro; 2005.
22. National Sample Survey Organisation: **Consumer expenditure survey 61st round.** New Delhi: Central Statistical Organisation; 2007.
23. Yiengprugsawan V, Lim LLY, Carmichael GA, Sidorenko A, Sleigh AC: **Measuring and decomposing inequity in self-reported morbidity and self-assessed health in Thailand.** *Int J Equity Health* 2007, **6**:23(17 pages).
24. Cleland J, Bicego G, Fegan G: **Socioeconomic inequalities in childhood mortality: the 1970s to the 1980s.** *Health Transit Rev* 1992, **2**:1-18.
25. Wagstaff A, Nguyen NN: **Poverty and survival prospects of Vietnamese children under Doi Moi.** In *Economic Growth, Poverty and Household Welfare: Policy Lessons from Vietnam.* Edited by: Glewwe P, Agrawal N, Dollar D. Washington DC: World Bank; 2003:.
26. Stecklov G, Bommier A, Boerma T: **Trends in equity in child survival in developing countries: An illustrative example using Ugandan data.**

Conference paper New York: Population Association of America Annual Meeting; 1999.

doi:10.1186/2191-1991-1-11

Cite this article as: Lauridsen and Pradhan: Socio-economic inequality of immunization coverage in India. *Health Economics Review* 2011 **1**:11.

Submit your manuscript to a SpringerOpen® journal and benefit from:

- Convenient online submission
- Rigorous peer review
- Immediate publication on acceptance
- Open access: articles freely available online
- High visibility within the field
- Retaining the copyright to your article

Submit your next manuscript at ► springeropen.com