# Understanding Associations Across Deprivation Indicators in MP 

## Research in-progress

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## Why Joint Distribution Matters?

Example : India NFHS data 2005-6 (sub-sample)
Raw headcount of mortality


Are they mostly the same people? Less than one-third of the time. What implications does this have for a multidimensional measure?

## Multidimensionality \& Association

## Debate:

Low association: to avoid redundancy

- HDI Debates

High association: to create stability

- Composite indicators
- Strong political message
- Techniques vary with data: PCA, MCA, FA, reliability, MD Scaling, Cluster, item response theory

Our practice to date

## This Paper

The aim of this paper is to:

Consider, which techniques to use to assess similarity (strength) and association (strength and direction) of potential variables for inclusion in a multidimensional poverty index.

Clarify how to interpret them in the context of deprivation indicators (dichotomous variables) for a counting index.

Many techniques are surveyed and assessed which do not appear in this presentation.

## 1. Sources of information

Dichotomised deprivation scores, 0 or 1 .

Raw headcounts $\rightarrow$ all deprivations

Censored headcounts $\rightarrow$ deprivations of the poor

## The Contingency Table

Formally:

Child mortality

| Years of Schooling | Non MD poor $=0$ | MD poor $=1$ | Total |
| :---: | :---: | :---: | :---: |
| Non MD poor $=0$ | $n_{00}$ | $n_{01}$ | $n_{0+}$ |
| MD Poor $=1$ | $n_{10}$ | $n_{11}$ | $n{ }^{1+}$ |
| Total | $n_{+0}$ | $n+1$ | $n$ |
| $n_{i j} \quad$ are | cell count freque |  | $\sum_{i=1}^{I} \sum_{j=1}^{J} r$ |

$n_{i+}, n_{+j}$ are the row, and column marginal totals

## 2. Traditional Measures of Association

Association (affinity) between two (or more) nominal (dichotomous) variables refers to a "coefficient" that measures the strength and direction(sign) of the relationship between the two variables.

Most coefficients of association define absence of association ("null" relationship) as independence.

Independence is based on the laws of probability: i.e. two variables are independent if their joint distribution equals the product of marginals. This is tested through the $\chi 2$ statistic.

Most coefficients of association for nominal variables like, Phi, Contingency, Cramer's V, Tschuprovw's T, Lambda, and Uncertainty rely on the $\chi 2$ statistic..

## 2.A Cramer's $V$ - Coefficient of Association

Cramer's $V$ : popular because of its norming range for $0-1$ variables
In the $2 \times 2$ case, $V$ ranges from 0 to $\pm 1$, and take the extreme values under (statistical) independence and "complete association".

$$
V=\frac{n_{00} n_{11-} n_{01} n_{10}}{\left(n_{0+} n_{1+} n_{+0} n_{+1}\right)^{1 / 2}}, \in[-1,1]
$$

## Meaning and interpretability of $\boldsymbol{V}$

$\mathrm{V}^{2}$ is the mean square canonical correlation between two variables.
Hence, V could be viewed as the percentage of the maximum possible variation between two variables.

Reported in many tables in papers in this workshop

## 2.A Cramer's $V$

## Sources of information used by V

Strength of the relationship is defined as the product of matches minus product of mismatches adjusting for the marginal distribution of the variables.

$$
V=\frac{\overbrace{n_{00} n_{11}}^{\text {matches }}-\overbrace{n_{01} n_{10}}^{\text {mismatches }}}{\underbrace{\left.n_{0+} n_{1+} n_{+0} n_{+1}\right)}_{\text {marginal distributions }}} 1 / 2, \in[-1,1]
$$

This is, V uses "entire cross-tab"
What are the implications for MD poverty analysis?

## Examples: Cramer V

## Case I

| Safe water (I) |
| :--- |
| Non MD poor $=0$ |
| MD Poor $=1$ |
| Total |

Child mortality (J)

| Non MD poor $=0$ | MD poor $=1$ | Total |
| :---: | :---: | :---: |
| 4 | 2 | 6 |
| $40 \%$ | $20 \%$ | $60 \%$ |
| 1 | 3 | 4 |
| $10 \%$ | $30 \%$ | $40 \%$ |
| 5 | 5 | 10 |
| $50 \%$ | $50 \%$ |  |

$$
V=\frac{n_{00} n_{11-} n_{01} n_{10}}{\left(n_{0+} n_{1+} n_{+0} n_{+1}\right)^{1 / 2}}=\frac{4 * 3-1 * 2}{(5 * 6 * 5 * 4)^{1 / 2}}=\bigcirc 0.41
$$

Note the + value of V - both indicators move in the same direction Ch Mort: 50\%-50\% (constant) ; Saf wat. 60\%-40\% (decrease) How sensitive V is to changes in the joint distribution?

## Examples: Cramer V

## Case II

Child mortality (J)

| Safe water (I) | Non MD poor $=0$ | MD poor = 1 | Total |
| :---: | :---: | :---: | :---: |
| Non MD poor $=0$ | 1 | 3 | 4 |
|  | 10\% | 30\% | 40\% |
| MD Poor $=1$ | 4 | 2 | 6 |
|  | 40\% | 20\% | 60\% |
| Total | 5 | 5 | 10 |
|  | 50\% | 50\% |  |
| $V=\frac{n_{00} n}{\left(n_{0+} n_{1}\right.}$ | $\frac{-n_{01} n_{10}}{\left.\imath_{+0} n_{+1}\right)^{1 / 2}}=\frac{}{(5}$ | $\frac{1 * 2-4 * 3}{5 * 4 * 5 * 6)}$ | $\frac{12}{12}=$ |

Note the - value of V - both indicators move in opposite directions
Ch Mort: 50\%-50\% (still constant) ; Saf wat. 40\%-60\% (now increase)
$\mathbf{V}$ does not reflect the change in 'poor-poor' cell

## Examples: Cramer V

Case III: Absence of poverty (both indicators)
Child mortality (J)


Non-overlap leads to a $\mathrm{CV}=-0.53$

## Examples: Cramer V

Case IV: Absence of Non poverty (both indicators)
Child mortality (J)

| Safe water (I) | Non MD poor $=0$ | MD poor $=1$ | Total |
| :---: | :---: | :---: | :---: |
| Non MD poor $=0$ | 0 | 3 | 3 |
|  | 0\% | 30\% | 30\% |
| MD Poor $=1$ | 4 | 3 | 7 |
|  | 40\% | 30\% | 70\% |
| Total | 4 | 6 | 10 |
|  | 40\% | 60\% |  |
| $n_{00} n_{11-} n_{01} n_{10}$ |  | $0 * 3-4 * 3$ |  |

Greater poor-poor leads to the same $\mathrm{CV}=-0.53$
Conclusion: Insufficient for our purposes

## 2. Similarity Coefficients

There is an extensive list of binary similarity coefficients.
Hubalek (1982) surveys 43 similarity coefficients for binary/dichotomous data

Two simple and very intuitive ones are:
a) The Simple Matching Coefficient - $S M$ Sokal \& Sneath, (1963)
b) The Jaccard Coefficient - $J$
Jaccard, (1901); Sneath, (1957)

## 2. Jaccard Similarity Coefficient

## Meaning and interpretability

Counts the number of observations (households/individuals) which have the same status (only poor) in both variables

Strength of the relationship is defined as the proportion of
"matches" in poverty only
Sources of information used by SM: Entire cross-tab
$n_{00} \quad$ number of people who are not MD poor
$n_{11}$ number of people who are MD poor in both indicators
$n \quad$ joint distribution of matches and mismatches

$$
J=\frac{n_{11}}{n-n_{o o}}, \in[0,1]
$$

What are the implications for MD poverty analysis?

## Examples: J

## Case I

Child mortality (J)

| Safe water (I) | Non MD poor $=0$ | MD poor $=1$ | Total |
| :---: | :---: | :---: | :---: |
| Non MD poor $=0$ | 4 | 2 | 6 |
|  | 40\% | 20\% | 60\% |
| MD Poor $=1$ | 1 | 3 | 4 |
|  | 10\% | 30\% | 40\% |
| Total | 5 | 5 | 10 |
|  | 50\% | 50\% |  |
|  | $=\frac{n_{11}}{n-n_{o o}}=\frac{}{10}$ | $-4=0.5$ |  |

How sensitive these are to changes in the joint distribution?

## Examples: J

Case III: Absence of poverty (both indicators)
Child mortality (J)

| Safe water (I) | Non MD poor $=0$ | MD poor $=1$ | Total |
| :---: | :---: | :---: | :---: |
| Non MD poor $=0$ | 3 | 3 | 6 |
|  | 30\% | 30\% | 60\% |
| MD Poor $=1$ | 4 | 0 | 4 |
|  | 40\% | 0\% | 40\% |
| Total | 7 | 3 | 10 |
|  | 70\% | 30\% |  |
|  | $J=\frac{n_{11}}{n-n_{o o}}=\frac{}{1}$ | $\frac{0}{0-3}=0$ |  |

Note the levels of poverty: 30\% in Ch. Mort; $40 \%$ in Safe water

## Examples: J

Case IV: Absence of Non poverty (both indicators)
Child mortality (J)

| Safe water (I) | Non MD poor $=0$ | MD poor $=1$ | Total |
| :---: | :---: | :---: | :---: |
| Non MD poor$=0$ | 0 | 3 | 3 |
|  | 0\% | 30\% | 30\% |
| MD Poor $=1$ | 4 | 3 | 7 |
|  | 40\% | 30\% | 70\% |
| Total | 4 | 6 | 10 |
|  | 40\% | 60\% |  |
|  | $\frac{n_{11}}{n-n_{o o}}=\frac{3}{10-0}$ | - $=0.3$ |  |

Full non poverty leads to different J
What about the "levels"? These have increased, but J is not sensitive.

$$
\begin{aligned}
& A: J=(2 /(10-6))=50 \% \\
& B: J=(1 /(10-8))=50 \%
\end{aligned}
$$

- Not sensitive to level;
- Not sensitive to overlap

A

## B

Child mortality (J)

| Safe water (I) | Non MD poor $=0$ | $\begin{aligned} & \mathrm{MD} \text { poor } \\ & =1 \end{aligned}$ | Total | Safe water <br> (I) | Non MD poor $=0$ | $\begin{aligned} & \text { MD poor } \\ & =1 \end{aligned}$ | Total |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| $\begin{aligned} & \text { Non MD poor } \\ & =0 \end{aligned}$ | $\begin{gathered} \mathbf{6} \\ 60 \% \end{gathered}$ | $\begin{gathered} 1 \\ 10 \% \end{gathered}$ | $\begin{gathered} 7 \\ 70 \% \end{gathered}$ | $\begin{aligned} & \text { Non MD } \\ & \text { poor }=0 \end{aligned}$ | $\begin{gathered} 8 \\ 0 \% \end{gathered}$ | $\begin{gathered} 0 \\ 0 \% \end{gathered}$ | $\begin{gathered} 8 \\ 80 \% \end{gathered}$ |
| MD Poor $=1$ | $\begin{gathered} 1 \\ 10 \% \end{gathered}$ | $\begin{gathered} \hline 2 \\ 20 \% \end{gathered}$ | $\begin{gathered} 3 \\ 30 \% \end{gathered}$ | $\text { MD Poor }=1$ | $\begin{gathered} 1 \\ 10 \% \end{gathered}$ | $\begin{gathered} 1 \\ 10 \% \end{gathered}$ | $\begin{gathered} 2 \\ 30 \% \end{gathered}$ |
| Total | $\begin{gathered} 7 \\ 70 \% \end{gathered}$ | $\begin{gathered} 3 \\ 30 \% \end{gathered}$ | 10 | Total | $\begin{gathered} 9 \\ 90 \% \end{gathered}$ | $\begin{gathered} 1 \\ 10 \% \end{gathered}$ | 10 |

## An Alternative Measure "P"

If two deprivation/poverty indicators are not independent, and if at least one of the marginal distributions $n_{1+}, n_{+1}$ is different from zero $P$ is defined as:

$$
P=\frac{n_{11}}{\min \left[n_{1+}, n_{+1}\right]}, \in[0,1]
$$

## Meaning and interpretability

Counts the number of observations (households/individuals) which have the same status (both poor or both deprived) in both variables, adjusted by the "level" of poverty

Strength of the relationship is defined as the proportion of "poverty matches" in the lowest level of poverty

## Sources of information used by $\mathbf{P}$ :

$n_{11} \quad$ number of people who are MD poor in both indicators $\rightarrow$ Joint
$n_{1+}, n_{+1}$ censored headcount ratios ("levels") $\rightarrow$ Marginals

## Examples: P

## Case I

Child mortality (J)

| Safe water (I) | Non MD poor $=0$ | MD poor $=1$ | Total |
| :---: | :---: | :---: | :---: |
| Non MD poor $=0$ | 4 | 2 | 6 |
|  | 40\% | 20\% | 60\% |
| MD Poor $=1$ | 1 | 3 | 4 |
|  | 10\% | 30\% | 40\% |
| Total | 5 | 5 | 10 |
|  | 50\% | 50\% |  |
| $P=\frac{r}{\min [n}$ | $\frac{1}{\left.+, n_{+1}\right]}=\frac{3}{\min [5}$ | $\frac{, 4]}{}=\frac{3}{4}=0 .$ |  |

$50 \%$ of people are poor in Ch.Mort, $40 \%$ in safe water, $30 \%$ both $75 \%$ of poor people in Safe water are poor in both

How sensitive these are to changes in the joint distribution?

## Examples: P

## Case V

Child mortality (J)

| Safe water (I) | Non MD poor $=0$ | MD poor $=1$ | Total |
| :---: | :---: | :---: | :---: |
| Non MD poor $=0$ | 4 | 3 | 7 |
|  | 40\% | 30\% | 70\% |
| MD Poor $=1$ | 1 | 2 | 3 |
|  | 10\% | 20\% | 30\% |
| Total | 5 | 5 | 10 |
|  | 50\% | 50\% |  |
| $P=\frac{n}{\min [n}$ | $\frac{11}{\left.+, n_{+1}\right]}=\frac{2}{\min [5,}$ | $\frac{}{, 3]}=\frac{2}{3}=0 .$ |  |

Decrease in the level of poverty
$50 \%$ of people are poor in Ch.Mort, $30 \%$ in safe water, $20 \%$ both $66 \%$ of poor people in Safe water are poor in both

## Examples: P

## Case IV

Child mortality (J)

$60 \%$ of people are poor in Ch.Mort, $70 \%$ in safe water, $30 \%$ both $50 \%$ of poor people in Ch.Mortality are poor in both

## 3. Illustration of "P" - Countries

| Country | DHS <br> Year |  | Country | DHS <br> Year |
| :--- | :---: | :--- | :--- | :---: |
| Bolivia | 2008 |  |  |  |
| Ethiopia | 2005 |  | Namibia | 2007 |
| Gabon | 2000 | Nepal | 2006 |  |
| Ghana | 2008 | Nigeria | 2008 |  |
| Haiti | 2006 | Rwanda | 2005 |  |
| Kenya | 2009 | Swaziland | 2007 |  |
| Malawi | 2004 | Uganda | 2006 |  |
| Mali | 2006 | Zimbabwe | 2006 |  |

Criteria of selection:
Information on all 10 censored headcount indicators
Variability across indicators

## 3. Censored Headcount Ratios



# 3. "P" Coefficient - Average over 15 countries "P" Coefficient (\%) 

|  |  | Sch. Enrol. Ch.Mort. Nut. |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  | Schooling <br> Enrolment <br> Ch.Mortality <br> Nutrition |  | 35 | 31 | 28 |
|  |  | 45 |  | 45 | 41 |
|  |  | 51 | 54 |  | 46 |
| Indicator with the lowest |  | 39 | 37 | 53 |  |
|  | Coefficient of Variation of "P" |  |  |  |  |
| Censored <br> Headcount |  | Sch. Enrol |  | Ch.Mort. Nut. |  |
|  | Schooling <br> Enrolment <br> Ch.Mortality <br> Nutrition |  | 0.49 | 0.38 | 0.61 |
|  |  | 0.43 |  | 0.28 | 0.44 |
|  |  | 0.35 | 0.42 |  | 0.29 |
|  |  | 0.45 | 0.49 | 0.19 |  |

## 3. What about Living Standard Indicators?

| Let's look at Fuel: | Fuel |  |  |  |
| :---: | :--- | :---: | :---: | :---: | :---: |
|  |  | Average <br> $\mathbf{P}$ | Number <br> of | Coefficient <br> Variation |
|  |  | (\%) | Countries | of P |

Very high values of P across 15 countries, very small C.V Redundancy?

## 4. Concluding Remarks

## Redundancy?

This still needs to be verified for a larger number of countries
This illustration considers countries with very similar profiles of deprivation/ poverty

Our hypothesis:
If high values of P are found, we might need to:
Consider a restrained version of "acute poverty", and alternative weighs.

## Thank you

