



Health Status and Calorie Inequalities Linkage among Major Indian States: An Empirical Exploration

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Health Status and Calorie Inequalities Linkage among Major Indian States: An Empirical Exploration[#]

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Abstract

This paper aims at making an assessment of temporal shifts in relative positioning of the major Indian states, jointly through the chief concomitants of health, as also through the measures of calorie inequalities. The task was accomplished through *factor analysis* (with *promax oblique rotation*), duly coupled with *canonical correlation analysis*, on the information on 16 indicators of health during *three rounds* for *seventeen* major Indian states. As per the findings, major determinants of health status have undergone voluminous reshuffling during the study span. Through *composite index*, states like Tamil Nadu, Maharashtra and Kerala were observed to have undergone perceptible temporal improvements in health status, whereas the so-called better-off states like Punjab and Gujarat have slipped fairly sharply in their relative rankings. As per FGT(2) index (due to Foster *et al.*, 1984, 2010; measuring *relative deprivation*), the averaged inequalities, within each of rural and urban regions, were highly significantly different among the states as also among the rounds. Temporally, the inequalities portrayed an inverted-U pattern. Gravity of the situation on calorie inequalities in south Indian states (like Tamil Nadu, Karnataka and Kerala) was alarming whereas, on the other extreme, the same in certain north-Indian hilly states (like Himachal Pradesh and Jammu & Kashmir) was manageable. Further, through *panel data estimation*, association between the composite index of health status and FGT(2) measure of the inequalities was indirect and statistically significant. There was a feeble indication of an indirect association between the measure of inequalities and per capita Income. However, association between the composite index of health and per capita income was direct and very robust. Thus, as a policy measure, there is a dire need for shifting priorities in favour of investment on both physical and social health infrastructure, particularly in laggard states and in those states which have undergone a rapid slippage in their ranking. *Public-private-partnership model* in this important social activity (which otherwise has remained a soft target by the governments), may prove beneficial.

Key Words: Calorie Inequalities, Time Series Factor Analysis, Promax Oblique Rotation, Composite Index, Canonical Correlation Analysis, FGT Index.

JEL Codes: C33, C38, I14, I15, R11, R58.

1. Introduction

As per WHO (1946), health of an individual is defined as a state of complete physical, mental and social well being, and not merely absence of disease or infirmity. A community would be healthy if a large majority of the individuals constituting the community are so. The basic necessities for anyone to be healthy are availability of adequate amount of food (qualitative

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enough so as to meet the bare minimal nutritional requirements), shelter and clothing which, however, could be met only if there is a sufficient scope of income generation.

Health status of a cohort of people is characterized by a multiplicity of factors – demographic, physical as well as social health infrastructure. Relative significance of these factors may differ temporally as well as spatially. Furthermore, nutritional requirements necessary for keeping good health are also known to vary across individuals and over time due to a multiplicity of reasons, generally not known to us (Ravallion, 1992). In fact, an adequate nutritional intake by the people is accepted as a meaningful yardstick for the success story of development policies of a nation.

Among Indian states, there are known to exist wide inequalities in the level of development. Although hunger and malnutrition are common, yet the extent and pattern differ from state-to-state. Nutritional status of the population depends not merely on food production and availability but more so on quantity and quality of food consumption. Uptil mid-sixties, nutritional problem was considered as the one associated with protein deficiency, especially in respect of quality protein from animal food. In line with the experience of a large majority of the developing economies, nutritional levels of the people are inadequate.

This paper aims at making an assessment of temporal shifts in relative positioning of the major Indian states, jointly through the chief concomitants of health, as also through the measures of calorie inequalities. An evaluation has also made on the relative positioning of the states, jointly on the basis of the so-identified determinants, so as to enable policy-makers to adopt policies conducive for balanced growth.

The paper has been organized into six sections in all, including the current one. The second section deals with a brief review of the related literature, so as to assist us in having a feel of the research gaps, if any, which possibly could be filled through the present study. A brief mention of the data base has been made in the third section, whereas analytical techniques adopted in the study have been mentioned in the fourth section. The fifth section is devoted to a brief discussion on the results obtained, first on health issues and then on calories inequalities. And, finally, concluding remarks and policy implications have been given in the sixth section.

2. Literature Reviewed

In the Indian context and elsewhere, a large number of empirical studies have been carried out on temporal and interregional status of health and nutrition, some of which have been reviewed in this section. As per Sukhatme (1970), calorie deficiency was of grave concern; in almost all classes of people, there was no group that was protein deficient, but no group that was not calorie deficient. His findings were duly supported by Gopalan (1970), Gopalan *et al.*, (1971) for India, and by Ghassemi (1972) for Iran. Subsequently, malnutrition caused by calorie deficiency assumed the central concern, especially among economists who took it as a basis for their measurement of poverty. In their pioneering study, Dandekar and Rath (1971) estimated poverty line on the basis of minimum calorie requirement, and magnitude of the proportion of people living below this level was considered as measurement of poverty. In order to examine the trends in inter-regional and inter-sectoral disparities in India, Rao (1971) constructed a composite index, and observed that less developed states were moving towards national composite rate and that the inter-state disparities were reducing. However, Rao (1973) opined that disparities had not come down at the pace as these should have been in the course of 15 years of planning. Bardhan (1974a, 1974b) and Srinivasan (1974) have contributed significantly towards distributional aspects of income and poverty in India. Chatterjee and Bhattacharya (1974) found that a glaring 60 percent of the real population was below the all India average intake in respect of calorie and all other nutrients, except vitamin C for which the proportion was 40 percent. Expressing his views on 'Some Nutritional Puzzles', Rao (1981) emphasised on balanced diet that would secure both the energy intake and the nutrients needed for good health. He suggested that the first step required for formulating a nutritional policy on food and nutrition is to formulate calorie intake norms that could be unambiguously used for estimating food requirements and, therefore, the deficits that have to be made up by increasing overall supplies. Dasgupta (1983) examined the intake of different nutrients for people of different age-sex-activity group. However, the study made no mention of the extent of inequality in nutritional intake. Dasgupta (1984) further studied the average intake of calories and proteins in India. Making use of the inequality indices G1, G2, and index of under-nutrition U, he found that although nutritional inequality was more evident in rural areas, the problem was not only of distribution but also of availability itself. In his analysis on inter-linkages between regional imbalances and plan outlays for 15 Indian states, Sarkar (1994) observed that Punjab scored the

highest and Bihar the lowest rank on the level of development. Through the construction of a composite index, Pathak and Gaur (1997) tried to find out in relative terms, the health status of different states in India. As per their findings, life expectancy at birth is an indicator of improvement, whereas each of infant mortality rate and crude birth rate are the indicators of deterioration in the health status of the Indian states. Emphasizing on the need to focus on the expansion of human capabilities, Das (1999) observed that the factors like education, availability of food, minimum purchasing power, facilities like safe drinking water, health infrastructure, *etc.*, played an important role in development process. Rao (2000) measured the effects of major forces affecting consumption of pulses during two periods of time, *viz.*, 1961 and 1971, in rural areas of two districts of Delhi. The author found that the income effect of pulses changed from positive to negative, indicating a rise in the general standard of living. Rao also observed a shift from consumption of energy food to consumption of body building and protective foods. Au *et al.* (2001) examined regional variations in the physical and mental health of patients receiving primary care in the largest integrated health care system in the United States. As per their findings, substantial differences in the health of patients enrolled in different VA primary clinics could be attributed to socio-demographic and co-morbid factors. Sethi (2003) emphasized that social sector strengthening (through increased government expenditure on health and education) would expectedly reduce the incidence of poverty *via* providing more employment opportunities, bringing down the death rate, and increasing the literacy rate. He stressed that reducing disparities and structural imbalances of the economy could be a long-lasting solution to bringing down the incidence of poverty. Roy *et al.* (2004) made an assessment of the extent of inequalities in health care and nutritional status across the Indian states with a focus on caste and tribe, and concluded that health services did not reach the disadvantaged sections. Kathuria and Sankar (2005) analyzed the performance of the rural public health systems of 16 major states in India, using stochastic production frontier techniques and panel data for the period 1986–97. The authors found that the health outcomes of Indian states in rural areas were positively related to the level of health infrastructure in terms of access to facilities and availability of skilled professionals, such as doctors. Their results further showed that states differed, not only in capacity building in terms of health infrastructure created but also in efficiency in using these inputs. Nair (2007) analyzed inter-state differentials in malnourishment among children in India on the basis of National Family Health Surveys: 1992–93, 1998–99, and 2005–6. The study

brought out the prevalence of widespread disparities and indicated that differentials were increasing over time. Such differentials, according to the author, did not always vary with the extent of poverty gaps among the states. Gupta (2009) made an inter-state comparison with respect to economic well-being, health, education, human development index, status of women, and social opportunities, and observed the presence of a strong relationship between these variables. Kumar (2011) noticed that wide regional disparities in women's status were present across states, which have persisted over time with little change in the development ranking of the Indian states. Punjab, Haryana, Kerala, Gujarat and Tamil Nadu have continued to occupy higher ranks in the index of economic development over time, while states like Bihar, Madhya Pradesh, Rajasthan and Uttar Pradesh have lagged behind. Kumari (2011) measured inter-district disparity in education and health attainment in Uttar Pradesh state at two points in time: 1990-91 and 2007-08. Through principal component analysis, her results revealed that apart from the presence of wide disparities, there were some regions/ districts that have performed well in educational attainment, but are poorly placed in terms of health attainment and *vice-versa*. Certain other studies (such as, due to Datta & Ganguly, 2002; Randhawa & Chahal, 2005; Giri, 2006; Musebe & Kumar, 2006; Nasurudeen *et al.* 2006; Chadha, 2007; Mittal, 2007; Kumar *et al.*, 2007; *etc.*) have analyzed the dynamics of per capita expenditure on various food groups across different income groups and arrived at the general conclusion that the share of non-cereal items in monthly per capita expenditure has consistently increased over time in both rural and urban regions. Sethi and Pandhi (2011a, 2011b) observed that a major chunk of consumption expenditure in India was incurred on items like Milk & Milk Products and Cereals, whereas the least expenditure was allocated to Fruits & Nuts and Spices. Through the application of Fisher's linear discriminant analysis, the authors detected the prevalence of voluminous interstate divergences in per capita consumption expenditure on major food items. Sethi and Pandhi (2012a, 2013) estimated the extent of inequalities in calorie intake among the Indian states/ union territories (UTs), and also identified the chief determinants of the inequalities. In another empirical study, Sethi and Pandhi (2012b) observed (through the applications of MANOVA and clustering techniques) the presence of high-profile gaps among states and UTs with respect to per capita per diem intake of calories, protein, and fat, separately for rural and urban regions. Further, Sethi and Pandhi (2014) examined the extent of interstate divergences (through a more robust general classificatory analysis) with respect to consumption expenditure on major food

items in India, and also to identify the clusters of the states at a similar level of the expenditure. The present study was a step further, wherein we have made an attempt to seek knowledge on the extent of interlinkage between health and nutritional aspects among the Indian states which, in turn, is expected to provide with a useful input for appropriate policy formulation at the regional level.

3. Data

For the purpose of identification of the major determinants of health, numerical information was compiled on as many as 16 indicators of health at three points in time: 1999-00, 2004-05, and 2011-12. The points had a close proximity with three Rounds of National Sample Surveys Organisation (NSSO) on Nutritional Intake in India: 55th (July, 1999 - June, 2000), 61st (July, 2004 - June, 2005) and 68th (July, 2011 - June, 2012). The indicators considered in the study were a mixture of demographic (or endogenous) variables [*viz.*, Birth Rate (BRRT, per 1000 population p.a.), Death Rate (DTRT, per 1000 population p.a.), Infant Mortality Rate (IMRT, per 1000 live births p.a.), and Life Expectancy at Birth (LEBR, in years)]; exogenous variables on physical and social infrastructure of health [*viz.*, Number of Hospitals per 100 Sq Km (NHPK), Number of Hospital Beds per Lakh of Population (NBPL), Number of Sub-Centers per 100 Sq Km (SBPK), Number of Primary Health Centers per 100 Sq Km (PHPK), Number of Community Health Centers per 100 Sq Km (CHPK), Number of Doctors per Lakh of Population (DCPL), Number of Pharmacists per Lakh of Population (PRPL), Number of Auxiliary Nursing Midwives per Lakh of Population (ANPL), Number of Lady Health Visitors per Lakh of Population (LVPL), Number of Nurses per Doctor (NRPD), and Number of Assistants per Doctor (NAPD)]; and level of living [*viz.* Per Capita Income (PCIN, in Rs'000)]. The major *seventeen* Indian states considered in the study were: Andhra Pradesh (ANP), Assam (ASM), Bihar (BHR), Gujarat (GUJ), Haryana (HAR), Himachal Pradesh (HMP), Jammu and Kashmir (JNK), Kerala (KRL), Karnataka (KTK), Madhya Pradesh (MDP), Maharashtra (MHR), Orissa (ORS), Punjab (PNB), Rajasthan (RAJ), Tamil Nadu (TND), Uttar Pradesh (UTP), West Bengal (WBN).

For measuring calorie inequalities among the states, data on the distribution of households by calorie intake level for different MPCE (*i.e.*, Monthly Per Capita Expenditure) classes of each of the states under study (separately for rural and urban regions), were culled out from the Reports

of 55th, 61st and 68th Rounds of NSSO on Nutritional Intake in India. As has been mentioned in Sethi and Pandhi (2012a), the data provides the two-way distribution of persons by calorie intake level. For constructing these tables, NSSO has used the information on per 1000 distribution of households by MPCE classes and by class intervals of actual calorie intake level as percent of normative level of 2700 kcal, for both rural and urban areas.

4. Analytical Techniques

For accomplishing the task, we have made use of analytical techniques, like *canonical correlation analysis* and *exploratory factor analysis*, duly followed by *FGT(2) measure* of nutritional inequalities. As per requirements of the former two multivariate analytical techniques, each of the manifest variables were considered in comparable terms (either in terms of per unit population or in terms of per unit area) and were suitably re-expressed, if required, in such a manner that higher value of each of the variables indicated towards better health status of a state. For instance, Death Rate was transformed into 1000/DTRT. Similar was the treatment in respect of Birth Rate and Infant Mortality Rate. Before subjecting the transformed data to the analysis, each of the variables were duly standardized for their mean (μ) and standard deviation (σ).

4.1. Canonical Correlation Analysis

Canonical correlation analysis, introduced originally by Hotelling way back in 1935, is known to be a very versatile multivariate technique for identifying relationships between two groups of variables. The analysis aims primarily at studying overall association between linear composites (called *canonical variates*) between two multivariate data sets (Akbas and Takma, 2005; Menderes *et al.*, 2005). Intrinsically, the analysis aims at identifying the *optimum structure* or *dimensionality* of each variable set that maximizes the relationships between the two sets of variables. An added advantage of the canonical correlation analysis is that it places the minimal restrictions on the types of data on which it operates.

Let there be p variates in Group-1 (say, endogenous variables of health) and q variates in Group-2 (say, endogenous variables of health), so that $p \leq q$. Let the variates in the two groups be X_1, X_2, \dots, X_p and Y_1, Y_2, \dots, Y_q , respectively. Suppose we write down the matrix of inter-correlation coefficients between the variables as

$$\mathbf{R} = \begin{pmatrix} \mathbf{r}_{x_1, x_1} & \cdots & \mathbf{r}_{x_1, x_p} & \mathbf{r}_{x_1, y_1} & \cdots & \mathbf{r}_{x_1, y_q} \\ \vdots & & \vdots & \vdots & & \vdots \\ \mathbf{r}_{x_1, x_1} & \cdots & \mathbf{r}_{x_1, x_p} & \mathbf{r}_{x_1, y_1} & \cdots & \mathbf{r}_{x_1, y_q} \\ \mathbf{r}_{y_1, x_1} & \cdots & \mathbf{r}_{y_1, x_p} & \mathbf{r}_{y_1, y_1} & \cdots & \mathbf{r}_{y_1, y_q} \\ \vdots & & \vdots & \vdots & & \vdots \\ \mathbf{r}_{y_q, x_1} & \cdots & \mathbf{r}_{y_q, x_p} & \mathbf{r}_{y_q, y_1} & \cdots & \mathbf{r}_{y_q, y_q} \end{pmatrix} = \begin{pmatrix} \mathbf{R}_{xx} & \mathbf{R}_{xy} \\ \mathbf{R}_{yx} & \mathbf{R}_{yy} \end{pmatrix} \quad \dots (1)$$

where

$$\mathbf{R}_{xx} = \begin{pmatrix} \mathbf{r}_{x_1, x_1} & \cdots & \mathbf{r}_{x_1, x_p} \\ \vdots & & \vdots \\ \mathbf{r}_{x_p, x_1} & \cdots & \mathbf{r}_{x_p, x_p} \end{pmatrix}; \mathbf{R}_{yy} = \begin{pmatrix} \mathbf{r}_{y_1, y_1} & \cdots & \mathbf{r}_{y_1, y_q} \\ \vdots & & \vdots \\ \mathbf{r}_{y_q, y_1} & \cdots & \mathbf{r}_{y_q, y_q} \end{pmatrix}; \mathbf{R}_{xy} = \begin{pmatrix} \mathbf{r}_{x_1, y_1} & \cdots & \mathbf{r}_{x_1, y_q} \\ \vdots & & \vdots \\ \mathbf{r}_{x_p, y_1} & \cdots & \mathbf{r}_{x_p, y_q} \end{pmatrix}; \text{ and } \mathbf{R}_{yx} = \mathbf{R}_{xy}^T$$

Consider the linear combinations among the variables (each expressed in terms of deviations from the corresponding mean values):

$$z_x = \sum_{i=1}^p u_i x_i \quad \text{and} \quad z_y = \sum_{j=1}^q v_j y_j \quad \dots (2)$$

Conceptually speaking, *canonical correlation* is the *maximum correlation that exists between the variates z_x and z_y* . Following Lindemann *et al.* (1980) and Tabachnick & Fidell (1989), the product moment correlation coefficient between the composite variables z_x and z_y is expressible as

$$r_{z_x, z_y} = \frac{\sum_{t=1}^n z_x z_y}{\sqrt{\sum_{t=1}^n z_x^2 \sum_{t=1}^n z_y^2}} = \frac{\mathbf{c}' \mathbf{R}_{xy} \mathbf{d}}{\sqrt{(\mathbf{c}' \mathbf{R}_{xx} \mathbf{c})(\mathbf{d}' \mathbf{R}_{yy} \mathbf{d})}} \quad \dots (3)$$

The vectors of weights \mathbf{c} and \mathbf{d} were chosen (with the help of *eigen roots* and *eigen vectors*) so as to ensure maximization of r_{z_x, z_y} .

Next, by following Lindemann *et al.* (1980), the overall *redundancy measure* (due to Stewart and Love, 1968) in the X-set, given that the entire set of canonical variates based on the Y-set is

where F_1, F_2, \dots, F_m (with $m < p$) are the *factors*; λ_{ij} are *loadings* (of the i^{th} observed measure on j^{th} factor); and $\varepsilon_1, \varepsilon_2, \dots, \varepsilon_p$ are *error* or *uniqueness terms* ($i = 1, 2, \dots, p; j = 1, 2, \dots, m$).

The loadings λ_{ij} might be viewed as the extent to which the observed variable x_i is associated with the factor F_j . A salient loading is the one which is significantly high to assume that a relationship exists between the variable and the factor. In addition, it means that the relationship is high enough, so that the variable can aid in interpreting the factor, and *vice versa* (Gorsuch, 1974).

As per the standard methodology (like the one adopted in *principal component analysis*), factors F_i and F_j would turn out to be orthogonal in the sense that manifest variables constituting a factor would be of the similar nature (or are complimentary) with respect to the phenomenon under study, but the variables constituting different factors would be independent.

The above system of equations could also be written as

$$\underline{x} = \underline{\Lambda}\underline{F} + \underline{\varepsilon} \quad \dots (5)$$

where \underline{x} is $p \times 1$ vector of observed variables, $\underline{\Lambda} = (\lambda_{ij})$ is $p \times m$ matrix of factor loadings, \underline{F} is $p \times 1$ vector of factors, and $\underline{\varepsilon}$ is $p \times 1$ vector of error (or, uniqueness) terms (with zero mean), assumed to be unrelated among themselves and with the factors in \underline{F} . Due to orthogonality of factors, variance of the observed variables is expressible as

$$V(x_i) = h_i^2 + \psi_i \quad (i = 1, 2, \dots, p) \quad \dots (6)$$

where $h_i^2 (= \sum_{j=1}^m \lambda_{ij}^2)$ stands for *communality*, which denotes the extent of variance (=1) in the given observed measure x_i , which stands explained by the common factors F_1, F_2, \dots, F_m , and may be conceptually viewed as something like *coefficient of multiple determination* (R^2) if x_i were regressed upon the m factors extracted. The remaining extent of variance ($=\psi_i$) in x_i is *uniqueness term*. In order to enhance the extent of variance explained by the factors in the observed variables, as also to come out with a conspicuous extraction of the *factors*, we have made use of *promax oblique rotation of the axes*. The number ‘ m ’ of factors extracted (through

the *pca* option) was decided through *eigen values* of the components, duly coupled with the rationale of *steepest descent in scree plot*.

Seeking the help of OECD (2008), and making use of the matrix of Λ of loadings, *composite index* for each of the states was constructed through the following steps:

(i) For each of m factors extracted, the proportion of variance explained (say, pve_j) in the data set was computed as

$$pve_j = \frac{\sum_{i=1}^p \lambda_{ij}^2}{\text{trace}(\text{icm})} \quad \dots (7)$$

(ii) For the i^{th} observed variable, let the maximum loading (say, λ_i^*) is realized on a particular factor having a proportion of variance explained = pve^* (say)

(iii) For this observed variable, weight (W_i) was computed as

$$W_i = \lambda_i^* \times pve^* \quad \dots (8)$$

(iv) And, finally, composite index (CMP_t) for the t^{th} district was computed as

$$CMP_t = \frac{W_i x_{ti}}{\sum_{i=1}^p W_i} \quad \dots (9)$$

where x_{ti} refers to the standardized value of the i^{th} observed variable in respect of t^{th} state. Computed values of the composite index formed the basis for gauging relative positioning of the states, jointly on the basis of the study variables, with respect health status.

4.3. FGT Measure of Calorie Inequalities

For the measurement of calorie inequalities, we have made use of the well-known FGT index of nutritional inequalities, as proposed by Foster, *et al.* (1984, 2010):

$$FGT(\alpha) = \frac{1}{n} \sum_{i=1}^k f_i \left[\frac{z - y_i}{z} \right]^\alpha ; \alpha \geq 0 \quad \dots (10)$$

where, n stands for the size of the population; k for the number of classes below the minimum level of calorie requirement z ; f_i for frequency of the i^{th} class below the nutritional level z ; and y_i for average calorie intake of people in the i^{th} class below the level z . Evidently, FGT measure is a weighted sum of the poverty gap ratios $\left[\frac{z - y_i}{z} \right]$ of the poor.

As has already been indicated in Sethi and Pandhi (2012), the FGT measure's *axiomatic* properties, like *versatility*, *additive decomposability*, *sub-group consistency*, and *distributive sensitivity*, make it to be a useful measure for undernourishment inequalities. With $\alpha = 2$, the index exhaustively takes into account three aspects, *viz.*, the number of undernourished in the population, depth of their undernourishment and their relative deprivation (Osberg and Xu, 2008). Consequently, in the present paper, we have made use of FGT (2) version of inequalities in calorie intake from the compiled information, separately in respect of rural and urban regions for each of the states under consideration.

For carrying out different types of analyses, suitably adapted codes in R-language (solely by the senior author of this paper) were made use of. For factor analysis, in particular, we have sought the help of 'FAiR' package (due to Goodrich, 2012) and 'tsfa' package (due to Gilbert and Meijer, 2012) for applicability in balanced panel data framework.

5. Results and Discussion

Results obtained from the study have been discussed in brief under the following sub-heads:

5.1. Canonical Correlation Analysis

This part of the analysis aimed primarily at examining whether inclusion of the (four) demographic variables in conjunction with the (eleven) variables of physical & social health infrastructure was justified or not. As per our computations (Table 5.1.1), the first (*i.e.*, the highest) ordered canonical correlation coefficient (= 0.8837, associated with 140 d.f.) between the two groups of variables was tested (through *Wilks' λ*) to be highly significant (p-value ≈ 0). However, the other lower-ordered canonical correlates failed to show statistical significance. The redundancy measure (= 0.322) of the first group of variables, given the information on the second group, implied that even in the presence of the latter group, more than two-third of the information contained in the former group remained unsqueezed, thereby providing a strong

Table 5.1.1. Canonical Correlation Coefficients (R_c) of Different Orders between the Two Groups of Variables, and Testing for their Statistical Significance

| Dimension | Computed Value of R _c | Wilk's Λ for R _c | F-Value for Λ | DF ₁ | DF ₂ | p-value for F |
|-----------|----------------------------------|-------------------------------------|-----------------------|-----------------|-----------------|---------------|
| 1 | 0.8837 ^{***} | 0.0882 | 2.820 | 44 | 140 | <0.0001 |
| 2 | 0.6450 ^{NS} | 0.4025 | 1.321 | 30 | 109 | 0.1515 |
| 3 | 0.4657 ^{NS} | 0.6892 | 0.864 | 18 | 76 | 0.6217 |
| 4 | 0.3463 ^{NS} | 0.8801 | 0.664 | 8 | 39 | 0.7193 |

Table 5.1.2. Redundancy Measures for the X-Set (i.e., Endogenous) and Y-Set (i.e., Exogenous) of Variables

| Order (i) | R _{dx_j} | R _{dy_j} |
|--------------|--------------------------------|--------------------------------|
| 1 | 0.4204 | 0.2006 |
| 2 | 0.0465 | 0.0770 |
| 3 | 0.0303 | 0.0338 |
| 4 | 0.0252 | 0.0105 |
| Total | R_{dx} = 0.5224 | R_{dy} = 0.3220 |

evidence in the favour of including demographic variables alongwith the other infrastructural variables of health, as in the next section.

5.2. Chief Determinants of Health Status among Major Indian States

At each of the three points in time under study, examination of the chief determinants of health status was made separately through the usual factor analysis approach (based on the codes of 'FAiR' package), whereas a pooled examination over all the three points (in panel-data framework) was made through time-series factor analysis approach (based on the codes of 'tsfa' package), as briefly discussed below:

5.2.1. Point in time: 1999-00 (≡ 55th Round of NSSO)

For the year 1999-00, the compiled information on 16 variables for each of the 17 states has been given in Table 5.2.1.1. These data were then subjected to the factor analysis approach. The *scree plot* (showing a graphical behavior between component number and eigen values of the components) provided an indication that optimum number of factors which need be extracted is *six*, because of the simple reason that the eigen values associated with rest of the components were less than (the standard norm of) unity. Therefore, the subsequent analysis was carried out by taking the number of factors to be extracted equaling six.

The computed values of *loadings* on the factors extracted (with *promax oblique* rotation of the axes) have been presented in the Table 5.2.1.2. In fact, these loadings are measures of association

Table 5.2.1.1. Information on Different Indicators of Health Status among Major States of India: 1999-00

| State | Variable | | | | | | | | | | | | | | | |
|-------|----------|-------|-------|-------|-------|--------|-------|-------|-------|--------|-------|-------|-------|-------|-------|-------|
| | BR RT | DT RT | IM RT | LE BR | NH PK | NB PL | SB PK | PH PK | CH PK | DC PL | PR PL | AN PL | LV PL | NR PD | NA PD | PC IN |
| ANP | 44.8 | 12.0 | 15.4 | 63.0 | 1.14 | 92.82 | 0.30 | 0.04 | 0.01 | 38.86 | 1.79 | 24.22 | 3.78 | 0.04 | 0.09 | 18.53 |
| ASM | 37.17 | 10.4 | 13.3 | 57.6 | 0.34 | 48.33 | 0.56 | 0.17 | 0.04 | 54.41 | 2.71 | 23.05 | 2.24 | 0.03 | 0.04 | 13.63 |
| BHR | 31.3 | 113.6 | 16.1 | 60.0 | 0.35 | 35.94 | 1.94 | 0.22 | 0.02 | 39.82 | 1.54 | 14.58 | 3.10 | 0.03 | 0.05 | 7.19 |
| GUJ | 39.7 | 133.3 | 16.1 | 63.1 | 1.29 | 127.74 | 1.13 | 0.13 | 0.03 | 64.81 | 1.74 | 21.96 | 3.03 | 0.75 | 0.79 | 21.05 |
| HAR | 37.2 | 133.3 | 14.9 | 64.8 | 0.18 | 35.16 | 2.78 | 0.77 | 0.14 | 5.16 | 4.16 | 21.4 | 1.66 | 0.01 | 0.01 | 26.94 |
| HMP | 45.2 | 13.8 | 16.7 | 65.7 | 0.15 | 102.95 | 0.18 | 0.03 | 0.01 | 3.46 | 4.88 | 64.33 | 11.47 | 0.33 | 0.64 | 25.09 |
| JNK | 51.0 | 161.3 | 19.2 | 65.7 | 0.02 | 21.01 | 0.24 | 0.04 | 0.01 | 64.20 | 0.73 | 10.86 | 3.88 | 0.04 | 0.07 | 14.08 |
| KRL | 55.9 | 15.6 | 71.4 | 73.4 | 0.54 | 314.79 | 0.65 | 0.16 | 0.01 | 92.18 | 3.21 | 28.35 | 5.40 | 0.08 | 0.07 | 22.69 |
| KTK | 45.5 | 12.8 | 17.5 | 64.2 | 0.15 | 73.99 | 0.99 | 0.19 | 0.03 | 110.50 | 1.90 | 26.31 | 4.39 | 0.47 | 0.36 | 19.75 |
| MDP | 32.1 | 97.1 | 11.4 | 56.5 | 0.12 | 30.72 | 1.61 | 0.22 | 0.06 | 29.50 | 4.60 | 40.65 | 3.22 | 0.02 | 0.03 | 12.63 |
| MHR | 47.8 | 133.3 | 20.8 | 66.0 | 1.12 | 104.37 | 0.61 | 0.10 | 0.02 | 81.42 | 2.31 | 18.37 | 4.04 | 0.07 | 0.11 | 25.56 |
| ORS | 41.2 | 95.2 | 10.4 | 57.9 | 0.20 | 33.04 | 1.37 | 0.34 | 0.04 | 38.64 | 4.79 | 20.25 | 3.22 | 0.03 | 0.09 | 11.66 |
| PNB | 46.5 | 135.1 | 19.2 | 68.2 | 0.44 | 62.41 | 2.88 | 0.92 | 0.18 | 131.52 | 2.68 | 27.21 | 6.10 | 0.02 | 0.03 | 29.19 |
| RAJ | 32.1 | 117.6 | 12.7 | 60.6 | 0.03 | 31.59 | 0.33 | 0.06 | 0.01 | 35.83 | 4.27 | 28.39 | 3.69 | 0.66 | 0.15 | 14.71 |
| TND | 52.1 | 12.6 | 19.6 | 64.9 | 0.31 | 79.04 | 0.04 | 0.04 | 0.01 | 102.78 | 2.23 | 23.25 | 7.59 | 0.01 | 0.24 | 23.02 |
| UTP | 30.5 | 97.1 | 12.0 | 58.5 | 0.31 | 29.11 | 0.57 | 0.08 | 0.04 | 26.14 | 0.57 | 19.84 | 5.64 | 0.01 | 0.14 | 11.02 |
| WBN | 48.5 | 14.2 | 19.6 | 63.8 | 0.45 | 68.84 | 0.83 | 0.23 | 0.03 | 62.45 | 1.56 | 21.16 | 3.98 | 0.07 | 0.07 | 17.82 |

Table 5.1.1.2. Determinants of Health Status in India: Results of Factor Analysis: 1999-00

| Variable | Loadings | | | | | |
|----------|----------|----------|----------|----------|----------|----------|
| | Factor-1 | Factor-2 | Factor-3 | Factor-4 | Factor-5 | Factor-6 |

| | | | | | | |
|---|--------------|--------------|--------------|--------------|--------------|--------------|
| BRRT | 0.769 | -0.158 | 0.082 | -0.222 | 0.085 | 0.036 |
| DTRT | 1.018 | -0.022 | -0.174 | 0.190 | 0.003 | -0.198 |
| LEBR | 0.898 | 0.121 | 0.009 | 0.043 | 0.242 | -0.055 |
| DCPL | 0.504 | 0.154 | -0.271 | 0.093 | 0.110 | 0.018 |
| PCIN | 0.626 | 0.422 | 0.340 | 0.015 | -0.094 | 0.249 |
| SBPK | -0.109 | 0.928 | -0.120 | 0.009 | 0.036 | 0.018 |
| PHPK | 0.118 | 0.977 | -0.023 | -0.063 | -0.036 | -0.038 |
| CHPK | 0.122 | 0.962 | 0.008 | -0.046 | -0.135 | -0.013 |
| PRPL | -0.323 | 0.307 | 0.629 | 0.073 | 0.238 | -0.192 |
| ANPL | -0.127 | -0.034 | 0.927 | 0.026 | 0.074 | -0.069 |
| LVPL | 0.391 | -0.268 | 0.742 | -0.164 | -0.136 | -0.124 |
| NRPD | 0.092 | -0.048 | -0.095 | 1.045 | 0.011 | -0.107 |
| NAPD | 0.059 | -0.167 | 0.333 | 0.578 | -0.110 | 0.217 |
| IMRT | 0.340 | -0.074 | -0.034 | -0.050 | 0.833 | -0.094 |
| NBPL | 0.179 | -0.103 | 0.138 | 0.031 | 0.798 | 0.241 |
| NHPK | -0.155 | -0.031 | -0.185 | -0.056 | 0.085 | 1.042 |
| Proportion of Variance Explained | 0.224 | 0.200 | 0.139 | 0.098 | 0.096 | 0.086 |
| Cumulative Variance Explained | 0.224 | 0.423 | 0.562 | 0.660 | 0.756 | 0.842 |

of the study variables on the factors extracted. A glance at the table clearly indicates that the 6 factors taken together were capable of explaining as high as 84.2 percent of the variance in the available data set. In this percentage, the first factor, which was constituted by five variables (*viz.*, birth rate, death rate, life expectancy at birth, number of doctors per lakh of population, and per capita income), was capable of explaining variance to the tune of 22.4 percent. Thus, the most important factor during the year 1999-00 was dominated by the usual demographic (endogenous) variables. The next important (*i.e.*, the second) factor consisted of the number of sub-centers per 100 sq. km, number of primary health centers per 100 sq. km, and number of community health centers per 100 sq. km. Thus, broadly speaking, the factor was composed of the variables pertaining to physical infrastructure of health. Further, health manpower variables (such as, number of pharmacists per lakh of population, number of auxiliary nursing midwives per lakh of population, and number of lady health visitors per lakh of population) constituted the third factor, whereas the variables like number of nurses per doctor and number of health assistants per doctor, constituted the fourth factor. And, finally, the fifth and the sixth factors were found to be consisting of a mixture of demographic (*viz.*, infant mortality rate) and physical infrastructure of health variables (*viz.*, number of hospital beds per lakh of population and number of hospitals per lakh of population).

5.2.2. Point in time: 2004-05 (≡ 61st Round of NSSO)

For the second point in time (*i.e.*, 2004-05, comparable with the time-frame of the 61st Round of the NSSO), basic data on the study variables are given in Table 5.2.2.1. Here again, the scree plot indicated the optimum number of factors to be extracted to be six. Results from the corresponding factor analysis have been given in Table 5.2.2.2. As per the table, the most significant factor (explaining 22.4 percent, out of the total explained variance of 89.0 percent) happened to be constituted by six variables, *viz.*, number of hospital beds per lakh of population; number of auxiliary nursing midwives per lakh of population; number of lady health visitors per lakh of population; number of nurses per doctor; number of assistants per doctor; and per capita income. The second factor consisted of number of sub-centers per 100 sq.km; number of primary health centers per 100 sq. km; and number of community health centers per 100 sq. km. The factor next in importance was constituted by the demographic variables, *viz.*, birth rate; death rate; and life expectancy at birth. Thus, during this point in time, the variables pertaining to health manpower, physical health infrastructure, and demographic traits turned out to be the significant determinants of health status. Rests of the variables were found to have played relatively less significant role towards explaining the extent of variance in the data set.

Table 5.2.2.1 Information on Different Indicators of Health Status among Major States of India: 2004-05

| State | Variable | | | | | | | | | | | | | | | |
|-------|----------|-------|-------|-------|-------|--------|-------|-------|-------|--------|-------|-------|-------|-------|-------|-------|
| | BR RT | DT RT | IM RT | LE BR | NH PK | NB PL | SB PK | PH PK | CH PK | DC PL | PR PL | AN PL | LV PL | NR PD | NA PD | PC IN |
| ANP | 52.4 | 137.0 | 17.5 | 63.5 | 0.13 | 43.19 | 0.37 | 0.04 | 0.01 | 43.58 | 42.70 | 25.25 | 4.25 | 2.43 | 0.10 | 24.46 |
| ASM | 40.0 | 114.9 | 14.7 | 58.1 | 0.13 | 10.57 | 1.07 | 0.17 | 0.04 | 56.72 | 8.56 | 21.28 | 2.07 | 0.64 | 0.04 | 15.96 |
| BHR | 42.0 | 123.5 | 16.4 | 60.7 | 0.11 | 3.34 | 2.39 | 0.21 | 0.04 | 39.09 | 4.59 | 14.12 | 3.01 | 0.25 | 0.08 | 7.41 |
| GUJ | 47.6 | 140.8 | 18.5 | 63.6 | 0.26 | 64.50 | 1.13 | 0.17 | 0.04 | 70.62 | 38.54 | 16.63 | 2.88 | 2.24 | 0.63 | 31.95 |
| HAR | 47.6 | 149.3 | 16.7 | 65.3 | 0.32 | 31.40 | 3.21 | 0.73 | 0.15 | 6.05 | 8.19 | 19.97 | 1.94 | 0.41 | 0.01 | 36.47 |
| HMP | 75.2 | 144.9 | 20.4 | 66.2 | 0.25 | 119.72 | 0.18 | 0.06 | 0.02 | 8.89 | 44.18 | 48.23 | 9.86 | 5.72 | 0.45 | 32.79 |
| JNK | 69.9 | 181.8 | 20.0 | 66.2 | 0.03 | 29.83 | 0.10 | 0.03 | 0.02 | 77.05 | 10.15 | 17.79 | 3.59 | 0.77 | 0.07 | 15.84 |
| KR | 67.6 | 157.1 | 21.7 | 73.7 | 0.66 | 87.0 | 0.70 | 0.20 | 0.00 | 115.23 | 23.29 | 29.4 | 1.9 | 0.0 | 0.0 | 32.0 |

| | | | | | | | | | | | | | | | | |
|----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|
| L | 6 | 6.3 | 4 | 5 | 4 | 50 | 7 | 2 | 5 | .76 | 21 | 60 | 97 | 1 | 3 | 08 |
| KT | 55. | 14 | 20. | 64. | 0.4 | 76. | 0. | 0.1 | 0.0 | 122 | 128 | 21. | 3.5 | 6.4 | 0.2 | 25. |
| K | 9 | 0.8 | 0 | 7 | 5 | 03 | 87 | 5 | 2 | .27 | .06 | 11 | 8 | 4 | 4 | 42 |
| M | 45. | 111 | 13. | 57. | 0.1 | 27. | 1.8 | 0.1 | 0.0 | 31. | 2.1 | 25. | 3.2 | 1.3 | 0.0 | 14. |
| DP | 5 | .1 | 2 | 1 | 1 | 24 | 3 | 8 | 5 | 61 | 25 | 45 | 8 | 5 | 4 | 45 |
| M | 54. | 14 | 27. | 66. | 0.2 | 44. | 0.5 | 0.1 | 0.0 | 93. | 96. | 16. | 4. | 1.6 | 0.0 | 33. |
| H | 9 | 9.3 | 8 | 4 | 2 | 72 | 8 | 0 | 2 | 44 | 90 | 51 | 45 | 8 | 9 | 77 |
| R | | | | | | | | | | | | | | | | |
| OR | 61. | 10 | 13. | 58. | 0.3 | 34. | 1.6 | 0.3 | 0.1 | 38. | 31. | 26. | 3. | 2.2 | 0.0 | 15. |
| S | 3 | 5.3 | 3 | 7 | 0 | 89 | 9 | 1 | 1 | 75 | 35 | 20 | 01 | 7 | 6 | 88 |
| PN | 58. | 14 | 22. | 68. | 0.4 | 43. | 4. | 0.8 | 0.2 | 135 | 137 | 21. | 5. | 0.8 | 0.0 | 33. |
| B | 8 | 9.3 | 7 | 6 | 9 | 02 | 03 | 6 | 0 | .41 | .61 | 33 | 69 | 7 | 3 | 19 |
| RA | 42. | 14 | 14. | 61. | 0.1 | 34. | 0.3 | 0.0 | 0.0 | 37. | 29. | 23. | 3. | 2.3 | 0.1 | 18. |
| J | 0 | 2.9 | 7 | 3 | 3 | 90 | 3 | 5 | 1 | 56 | 82 | 81 | 03 | 6 | 2 | 06 |
| TN | 62. | 13 | 27. | 65. | 0.6 | 116 | 0. | 0.0 | 0.0 | 110 | 153 | 20. | 6. | 6.9 | 0.1 | 28. |
| D | 5 | 5.1 | 0 | 4 | 9 | .01 | 04 | 2 | 1 | .34 | .09 | 86 | 29 | 5 | 8 | 44 |
| UT | 37. | 114 | 13. | 59. | 0.3 | 18. | 0. | 0.1 | 0.0 | 26. | 16. | 1.4 | 4.1 | 0.2 | 0.1 | 12. |
| P | 7 | .9 | 7 | 2 | 8 | 15 | 06 | 0 | 1 | 60 | 93 | 1 | 0 | 4 | 0 | 39 |
| W | 79. | 15 | 26. | 64. | 0.4 | 64. | 0. | 0.2 | 0.0 | 63. | 106 | 17. | 2.1 | 0.9 | 0.0 | 22. |
| BN | 4 | 6.3 | 4 | 1 | 6 | 48 | 83 | 3 | 2 | 28 | .11 | 37 | 0 | 4 | 4 | 22 |

Table 5.2.2.2. Determinants of Health Status in India: Results of Factor Analysis: 2004-05

| Variable | Loadings | | | | | |
|----------------------------------|--------------|--------------|--------------|--------------|--------------|--------------|
| | Factor-1 | Factor-2 | Factor-3 | Factor-4 | Factor-5 | Factor-6 |
| NBPL | 0.671 | -0.161 | 0.041 | 0.413 | -0.044 | 0.086 |
| ANPL | 1.121 | 0.175 | -0.140 | -0.388 | -0.118 | 0.188 |
| LVPL | 0.764 | -0.025 | -0.062 | 0.036 | -0.012 | 0.046 |
| NRPD | 0.718 | -0.222 | -0.222 | 0.165 | 0.257 | -0.166 |
| NAPD | 0.630 | -0.145 | 0.029 | -0.176 | 0.023 | -0.278 |
| PCIN | 0.438 | 0.371 | 0.354 | 0.133 | -0.019 | 0.031 |
| SBPK | 0.005 | 0.973 | -0.136 | -0.104 | 0.115 | -0.077 |
| PHPK | -0.135 | 0.983 | 0.079 | 0.211 | -0.063 | -0.115 |
| CHPK | 0.035 | 0.998 | -0.075 | 0.041 | 0.032 | -0.037 |
| BRRT | 0.291 | -0.069 | 0.512 | 0.123 | -0.077 | 0.049 |
| DTRT | -0.244 | -0.111 | 1.241 | -0.288 | 0.022 | -0.090 |
| LEBR | 0.106 | 0.078 | 0.618 | 0.068 | 0.121 | 0.368 |
| NHPK | -0.057 | 0.117 | -0.245 | 1.077 | 0.010 | 0.288 |
| DCPL | -0.052 | 0.009 | 0.044 | -0.165 | 1.043 | 0.290 |
| PRPL | 0.062 | 0.112 | 0.091 | 0.290 | 0.609 | -0.283 |
| IMRT | 0.091 | -0.183 | -0.001 | 0.290 | 0.079 | 0.920 |
| Proportion of Variance Explained | 0.224 | 0.204 | 0.155 | 0.120 | 0.099 | 0.088 |
| Cumulative Variance Explained | 0.224 | 0.428 | 0.583 | 0.703 | 0.802 | 0.890 |

5.2.3. Point in time: 2011-12 (≡ 68th Round of NSSO)

The basic information on different indicators of health status for the third point in time (*i.e.*, 2011-12, which had a close comparability with the 68th Round of NSSO) is presented in Table 5.2.3.1. Results from factor analysis as applied to these data have been given in Table 5.2.3.2.

Table 5.2.3.1 Information on Different Indicators of Health Status among Major States of India: 2011-12

| State | Variable | | | | | | | | | | | | | | | |
|-------|----------|-------|-------|-------|-------|--------|-------|-------|-------|--------|--------|--------|-------|-------|-------|-------|
| | BR RT | DT RT | IM RT | LE BR | NH PK | NB PL | SB PK | PH PK | CH PK | DC PL | PR PL | AN PL | LV PL | NR PD | NA PD | PC IN |
| ANP | 54.6 | 133.3 | 20.4 | 64.2 | 0.17 | 45.42 | 4.55 | 0.57 | 0.06 | 74.42 | 52.47 | 33.74 | 4.16 | 2.19 | 0.06 | 29.57 |
| ASM | 42.4 | 116.3 | 16.4 | 59.0 | 0.20 | 24.84 | 5.85 | 1.08 | 0.14 | 62.30 | 7.92 | 30.09 | 1.94 | 0.77 | 0.03 | 18.77 |
| BHR | 35.1 | 137.0 | 19.2 | 61.3 | 1.82 | 22.16 | 9.41 | 1.89 | 0.07 | 36.68 | 4.10 | 10.05 | 1.10 | 0.25 | 0.03 | 11.02 |
| GUJ | 44.8 | 145.0 | 20.8 | 64.1 | 0.19 | 48.81 | 3.71 | 0.55 | 0.14 | 78.27 | 35.31 | 19.07 | 2.75 | 1.90 | 0.58 | 40.91 |
| HAR | 44.1 | 144.9 | 19.6 | 66.1 | 0.35 | 31.20 | 16.45 | 0.99 | 0.21 | 16.36 | 28.71 | 25.98 | 2.24 | 0.38 | 0.01 | 48.64 |
| HMP | 58.1 | 135.1 | 22.2 | 66.9 | 0.26 | 117.52 | 3.72 | 0.81 | 0.13 | 18.22 | 41.60 | 43.84 | 2.23 | 2.07 | 0.04 | 42.37 |
| JNK | 53.8 | 172.4 | 22.2 | 66.9 | 0.04 | 32.11 | 0.86 | 0.17 | 0.04 | 91.50 | 17.64 | 19.68 | 1.14 | 7.29 | 0.17 | 17.77 |
| KRL | 68.0 | 151.5 | 83.3 | 73.9 | 0.99 | 94.15 | 11.77 | 1.79 | 0.59 | 144.82 | 53.07 | 24.00 | 3.73 | 1.78 | 0.40 | 43.71 |
| KT | 51.3 | 135.1 | 24.4 | 65.4 | 0.48 | 105.80 | 4.25 | 1.14 | 0.17 | 144.94 | 131.97 | 19.57 | 2.60 | 1.56 | 0.14 | 32.72 |
| MDP | 36.1 | 116.3 | 14.9 | 58.0 | 0.15 | 40.04 | 2.88 | 0.37 | 0.11 | 37.31 | 1.93 | 23.61 | 0.67 | 1.11 | 0.01 | 14.65 |
| MHR | 56.8 | 151.5 | 32.3 | 67.2 | 0.58 | 45.16 | 3.44 | 0.59 | 0.12 | 124.48 | 95.94 | 26.04 | 6.14 | 1.65 | 0.12 | 39.37 |
| ORS | 47.6 | 111.1 | 15.4 | 59.6 | 1.26 | 38.36 | 4.94 | 0.94 | 0.17 | 40.54 | 34.57 | 20.44 | 2.24 | 2.46 | 0.04 | 22.08 |
| PNB | 58.8 | 138.9 | 26.3 | 69.4 | 0.46 | 38.83 | 5.86 | 0.78 | 0.26 | 140.53 | 129.03 | 25.361 | 2.98 | 0.77 | 0.01 | 41.03 |
| RAJ | 36.8 | 147.1 | 16.9 | 61.9 | 0.14 | 47.65 | 3.20 | 0.44 | 0.11 | 42.37 | 27.06 | 27.52 | 3.08 | 2.24 | 0.12 | 22.45 |
| TND | 61.3 | 135.1 | 35.7 | 66.2 | 0.45 | 66.38 | 6.69 | 0.98 | 0.20 | 122.11 | 213.74 | 18.05 | 2.56 | 6.56 | 0.06 | 34.66 |
| UTP | 34.8 | 119.1 | 15.9 | 59.9 | 0.36 | 28.77 | 8.52 | 1.53 | 0.21 | 29.57 | 15.45 | 1.14 | 0.21 | 0.24 | 0.09 | 14.74 |
| WBN | 58.1 | 161.3 | 30.3 | 65.0 | 0.33 | 60.73 | 11.67 | 1.04 | 0.38 | 65.29 | 99.41 | 19.14 | 1.24 | 0.84 | 0.02 | 27.99 |

Table 5.2.3.2. Determinants of Health Status in India: Results of Factor Analysis: 2011-12

| Variable | Loadings | | | | | |
|---|--------------|--------------|--------------|--------------|--------------|--------------|
| | Factor-1 | Factor-2 | Factor-3 | Factor-4 | Factor-5 | Factor-6 |
| BRRT | 0.563 | 0.181 | 0.444 | -0.233 | 0.130 | -0.017 |
| NBPL | 0.595 | 0.076 | 0.157 | 0.048 | -0.186 | -0.035 |
| ANPL | 1.275 | -0.233 | -0.430 | -0.327 | -0.083 | 0.006 |
| LVPL | 0.569 | -0.154 | 0.178 | 0.198 | -0.092 | 0.076 |
| LEBR | 0.413 | 0.225 | 0.221 | 0.092 | 0.389 | 0.205 |
| IMRT | 0.331 | 0.727 | -0.112 | 0.173 | 0.089 | -0.133 |
| NHPK | -0.263 | 0.545 | 0.025 | -0.039 | -0.167 | -0.061 |
| SBPK | -0.009 | 0.839 | -0.070 | -0.243 | 0.233 | 0.707 |
| PHPK | -0.256 | 0.876 | 0.024 | -0.046 | -0.214 | 0.144 |
| CHPK | 0.183 | 0.841 | -0.005 | 0.068 | -0.003 | 0.178 |
| DCPL | -0.085 | 0.021 | 0.742 | 0.160 | -0.040 | -0.215 |
| PRPL | -0.090 | -0.082 | 1.252 | -0.298 | -0.123 | 0.166 |
| NRPD | -0.114 | -0.337 | 0.328 | -0.092 | 0.299 | -0.424 |
| NAPD | -0.197 | -0.060 | -0.286 | 1.143 | 0.044 | 0.009 |
| DTRT | -0.072 | -0.081 | -0.139 | 0.044 | 1.118 | 0.089 |
| PCIN | 0.645 | -0.016 | 0.311 | 0.283 | -0.022 | 0.695 |
| Proportion of Variance Explained | 0.222 | 0.206 | 0.183 | 0.114 | 0.107 | 0.085 |
| Cumulative Variance Explained | 0.222 | 0.429 | 0.612 | 0.726 | 0.833 | 0.918 |

Scree plot for the analysis has, once again, provided an indication towards the acceptance of six factors for extraction. As per the results (Table 5.1.3.2), nearly one-fourth (22.2 percent) of the total explained variance (of 91.8 percent) was attributable to the most significant (*i.e.*, the first) factor, which happened to be a mixture of usual demographic indicators (like, birth rate and life expectancy at birth), physical health infrastructure (like, number of hospital beds per lakh of population), and health manpower statistics (such as, number of auxiliary nursing midwives per lakh of population and number of lady health visitors per lakh of population). Notably, the next most-significant factor (associated with variance explanation of nearly 21 percent) was broadly made up of physical health infrastructure variables (such as, number of hospitals per 100 sq.km, number of sub-centers per lakh of population, number of primary health centers per lakh of population, and number of community health centers per lakh of population). The remaining four factors were far from being explicit in nature, as these involved a mixture of the types of dimensions.

5.2.4. Pooled Analysis

Next, in order to squeeze out the overall picture on major determinants of health, the information at all the three points in time (in panel data frame work) was subjected to time series factor analysis. The analysis was capable of extracting a totality of *five* factors, which together could explain 67.4 percent of the variance in the data set (Table 5.2.4.1).

Table 5.2.4.1. Determinants of Health Status in India: Pooled Results of Time Series Factor Analysis

| Variable | Loadings | | | | |
|----------------------------------|--------------|--------------|--------------|--------------|--------------|
| | Factor-1 | Factor-2 | Factor-3 | Factor-4 | Factor-5 |
| SBPK | 0.971 | -0.104 | 0.001 | 0.044 | 0.060 |
| PHPK | 0.873 | 0.067 | 0.009 | -0.076 | -0.061 |
| CHPK | 0.841 | 0.143 | 0.073 | 0.056 | -0.045 |
| IMRT | 0.120 | 0.830 | -0.130 | -0.050 | 0.066 |
| LEBR | 0.052 | 0.795 | -0.099 | 0.054 | 0.364 |
| NBPL | -0.192 | 0.638 | -0.096 | 0.234 | -0.081 |
| DCPL | -0.126 | 0.640 | 0.413 | -0.296 | -0.084 |
| PRPL | 0.086 | -0.045 | 1.071 | -0.062 | -0.151 |
| NRPD | -0.121 | -0.183 | 0.658 | 0.070 | 0.094 |
| ANPL | 0.144 | -0.153 | -0.061 | 0.983 | -0.130 |
| LVPL | -0.328 | 0.263 | 0.025 | 0.587 | -0.154 |
| DTRT | -0.016 | 0.241 | -0.132 | -0.209 | 1.001 |
| BRRT | -0.073 | 0.276 | 0.358 | 0.072 | 0.315 |
| NHPK | 0.174 | 0.414 | -0.053 | -0.193 | -0.193 |
| NAPD | -0.093 | 0.017 | -0.097 | 0.320 | 0.097 |
| PCIN | 0.462 | 0.262 | 0.251 | 0.379 | 0.092 |
| Prop. of Variance Explained | 0.181 | 0.167 | 0.126 | 0.113 | 0.087 |
| Cum. Prop. of Variance Explained | 0.181 | 0.348 | 0.474 | 0.587 | 0.674 |

As regards constitution of the factors, the first factor (which could explain 18.1 percent of the variance) was constituted by three variables *viz.*, number of sub-centers per 100 sq km; number of primary health centers per 100 sq km; and number of community health centers per 100 sq km. Clearly, each of the constituent variables of the most important factor referred to the physical infrastructure of health. The next important factor (having accounted for 16.7 percent of the variance explained) was composed of four variables, *viz.*, infant mortality rate; life expectancy at birth; number of hospital beds per lakh of population; and number of doctors per lakh of population. The factor, obviously, happened to be a synthesis of physical & social health infrastructure, as also demographic variables. The third factor (capable of explaining 12.6

percent of the variance) was made up of two variables of social infrastructure: number of pharmacists per lakh of population; and number of nurses per doctor. Similarly, the fourth factor (responsible for explaining 11.3 percent of the variance), too, consisted of another couple of variables of social infrastructure, *viz.*, number of auxiliary nursing midwives per lakh of population; and number of lady health visitors per lakh of population. And, the last significant factor (accounting for 8.7 percent of the variance explained) consisted of the lone demographic variable – death rate. Notably, as per the results of the pooled analysis, rest of the four variables (*viz.*, birth rate; number of hospitals per 100 sq km; number of assistants per doctor; and per capita income), in the presence of the rest of the 12 variables, failed to show their importance while determining the health status of a state.

The analysis has thus provided us with an evidence that relative significance of different parameters of health status among the Indian states has not remained time invariant, but has instead undergone a drastic reshuffling during the study span. As per the pooled analysis, the most important dimension of health among the major Indian states is composed of the parameters of physical health infrastructure, followed by those of social health infrastructure.

5.3. Relative Positioning of the Major Indian States – Construction of the Composite Index

In order to examine the relative positioning of the major Indian states, as also to study temporal shifts, if any, in the positioning at each of the points in time (as also over the entire study period taken together), the composite index of health status was constructed through the methodology as outlined in Section 4.2 above (Table 5.3.1). A glance at the table shows that in the year 1999-00, Punjab (with a value of 4.57 for the index) occupied the top position, followed next by Kerala

Table 5.3.1. Composite Index of Health Status among the Major Indian States

| State | 1999-00 | | 2004-05 | | 2011-12 | | Pooled | |
|-------|-----------------|------|-----------------|------|-----------------|------|-----------------|------|
| | Composite Index | Rank | Composite Index | Rank | Composite Index | Rank | Composite Index | Rank |
| ANP | 3.246 | 11 | 2.334 | 12 | 2.666 | 9 | 2.659 | 9 |
| ASM | 2.942 | 16 | 1.978 | 16 | 2.315 | 15 | 2.064 | 14 |
| BHR | 2.958 | 15 | 2.027 | 15 | 2.340 | 14 | 1.914 | 15 |
| GUJ | 3.714 | 5 | 2.659 | 7 | 2.630 | 10 | 2.827 | 8 |
| HAR | 3.937 | 3 | 2.859 | 6 | 2.739 | 8 | 2.125 | 12 |
| HMP | 3.885 | 4 | 3.366 | 2 | 2.933 | 7 | 2.996 | 7 |
| JNK | 3.340 | 10 | 2.339 | 11 | 2.368 | 12 | 2.612 | 10 |
| KRL | 4.274 | 2 | 3.230 | 3 | 4.076 | 1 | 4.436 | 1 |

| | | | | | | | | |
|------------|-------|----|-------|----|-------|----|-------|----|
| KTK | 3.622 | 7 | 2.861 | 5 | 3.001 | 5 | 3.751 | 3 |
| MDP | 3.042 | 14 | 2.147 | 14 | 1.909 | 17 | 1.892 | 16 |
| MHR | 3.637 | 6 | 2.534 | 9 | 3.029 | 3 | 3.379 | 5 |
| ORS | 3.084 | 12 | 2.444 | 10 | 2.378 | 11 | 2.104 | 13 |
| PNB | 4.572 | 1 | 3.566 | 1 | 3.027 | 4 | 3.728 | 4 |
| RAJ | 3.080 | 13 | 2.181 | 13 | 2.360 | 13 | 2.265 | 11 |
| TND | 3.508 | 9 | 3.000 | 4 | 3.194 | 2 | 4.036 | 2 |
| UTP | 2.662 | 17 | 1.724 | 17 | 2.007 | 16 | 1.766 | 17 |
| WBN | 3.527 | 8 | 2.571 | 8 | 2.994 | 6 | 3.232 | 6 |

(4.27), Haryana (3.94), Himachal Pradesh (3.88), and Gujarat (3.71). On the other extreme, the bottom positions were occupied by Uttar Pradesh (2.66) preceded by Assam (2.94) and Bihar (2.96). In the year 2004-05, too, Punjab (having a value of the index equaling 3.57) continued to maintain its top position, while Himachal Pradesh (3.37) improved its position to occupy the second slot, followed then by Kerala (3.23). Notably, the state of Tamil Nadu (3.00), which happened to be at ninth position during 1999-00 underwent a drastic improvement and occupied the fourth position during 2004-05. While, the bottom slots continued to be occupied by the states like Uttar Pradesh (1.72), closely preceded by Assam (1.99) and Bihar (2.03). However, during 2011-12, the state of Kerala (4.27) was observed to have occupied the top position, whereas Tamil Nadu (3.19) experienced a further improvement in its position to have come at the second slot. Our findings are in fair agreement with those due to Kurian (2000) according to whom, even if a state is at a relatively lower level of per capita income, can yet enjoy a comparatively higher level of social development. Incidentally, the state of Punjab (associated with a value of 3.03 for the composite index) has lately slipped from the first rank to the fourth rank which, indeed, is a matter of concern for the state. On the other hand, Maharashtra, which was as low as at the 9th position (during 2004-05) has undergone a perceptible improvement to have occupied the 3rd position (during 2011-12). However, the states like Uttar Pradesh, Madhya Pradesh, Bihar, Assam and Rajasthan have continued to remain at relatively lower positions in terms of health status in a fairly stable manner.

The overall picture of the relative rankings of the states, as obtained through values of the composite index computed from the output of the time series factor analysis (Table 5.2.4.1) had a close similarity with the picture in respect of the analysis for 2011-12 (Table 5.3.1). On the whole, Kerala was ranked number one state (with a score of 4.44), followed next by Tamil Nadu (4.04), Karnataka (3.75), Punjab (3.73) and then by Maharashtra (3.38). On the other extreme,

the states like Uttar Pradesh (1.77), Madhya Pradesh (1.89), Bihar (1.91), Assam (2.06) and Odisha (2.10) have occupied the bottom positions on health traits.

5.4. Calorie Inequalities among the Major Indian States

For measuring calorie inequalities among the states, we have made use of the data on the distribution of households by calorie intake level for different MPCE (separately for rural and urban regions) of each of the states under study. For the purpose of clarification on the format of data used, we have presented such a distribution for rural regions of India as whole in respect of 55th Round of NSSO (Table 5.4.1).

Table 5.4.1. Per Thousand Distribution of Households by Calorie Intake Level for Each MPCE Class in Rural India - 55th Round

| MPCE Class (Rs) | Calorie Intake Level | | | | | | | | All Classes | SMP HHS |
|-----------------|----------------------|-------|-------|--------|---------|---------|---------|-------|-------------|---------|
| | < 70 | 70-80 | 80-90 | 90-100 | 100-110 | 110-120 | 120-150 | ≥ 150 | | |
| < 225 | 654 | 177 | 98 | 44 | 13 | 7 | 3 | 3 | 1000 | 2547 |
| 225-255 | 416 | 239 | 185 | 93 | 40 | 16 | 9 | 3 | 1000 | 2451 |
| 255-300 | 311 | 231 | 200 | 134 | 66 | 28 | 23 | 7 | 1000 | 5147 |
| 300-340 | 182 | 207 | 231 | 173 | 104 | 48 | 43 | 11 | 1000 | 5588 |
| 340-380 | 145 | 181 | 225 | 179 | 127 | 69 | 61 | 14 | 1000 | 5892 |
| 380-420 | 96 | 153 | 190 | 199 | 149 | 94 | 96 | 23 | 1000 | 5895 |
| 420-470 | 71 | 115 | 170 | 200 | 161 | 110 | 131 | 42 | 1000 | 6783 |
| 470-525 | 53 | 90 | 150 | 176 | 171 | 127 | 173 | 60 | 1000 | 6635 |
| 525-615 | 37 | 72 | 130 | 156 | 161 | 134 | 222 | 88 | 1000 | 8253 |
| 615-775 | 27 | 46 | 88 | 134 | 145 | 141 | 274 | 145 | 1000 | 9383 |
| 775-950 | 19 | 34 | 61 | 110 | 126 | 126 | 295 | 230 | 1000 | 5337 |
| ≥ 950 | 23 | 16 | 44 | 69 | 107 | 107 | 271 | 381 | 1000 | 7474 |
| All classes | 134 | 124 | 151 | 149 | 92 | 92 | 143 | 82 | 1000 | 69206 |

The computed values of calorie inequalities, as estimated through the FGT(2) measure (equation 10) have been exhaustively presented in Table 5.4.2. As per the table, there obviously have been wide inequalities in the calorie intake at different levels: between states, between rounds

Table 5.4.2. Computed Values of FGT(2) Index of Inequality among Major Indian States during the Three Rounds – Rural, Urban and Combined Regions

| State | Rural | | | | Urban | | | | Combined | | | |
|-------|------------------|------------------|------------------|--------|------------------|------------------|------------------|--------|------------------|------------------|------------------|--------|
| | Round | | | Mean | Round | | | Mean | Round | | | Mean |
| | 55 th | 61 st | 68 st | | 55 th | 61 st | 68 st | | 55 th | 61 st | 68 st | |
| ANP | 0.0326 | 0.0328 | 0.0158 | 0.0271 | 0.0369 | 0.0468 | 0.0089 | 0.0309 | 0.0348 | 0.0398 | 0.0124 | 0.0290 |
| ASM | 0.0461 | 0.0253 | 0.0252 | 0.0322 | 0.0298 | 0.0284 | 0.0105 | 0.0229 | 0.0380 | 0.0268 | 0.0178 | 0.0276 |
| BHR | 0.0283 | 0.0249 | 0.0220 | 0.0251 | 0.0263 | 0.0304 | 0.0103 | 0.0223 | 0.0273 | 0.0276 | 0.0161 | 0.0237 |
| GUJ | 0.0337 | 0.0399 | 0.0273 | 0.0336 | 0.0328 | 0.0416 | 0.0110 | 0.0285 | 0.0332 | 0.0407 | 0.0192 | 0.0310 |
| HAR | 0.0158 | 0.0222 | 0.0106 | 0.0162 | 0.0295 | 0.0407 | 0.0075 | 0.0259 | 0.0226 | 0.0314 | 0.0090 | 0.0210 |
| HMP | 0.0085 | 0.0095 | 0.0018 | 0.0066 | 0.0092 | 0.0154 | 0.0028 | 0.0091 | 0.0088 | 0.0124 | 0.0023 | 0.0079 |
| JNK | 0.0088 | 0.0102 | 0.0073 | 0.0088 | 0.0082 | 0.0091 | 0.0035 | 0.0069 | 0.0085 | 0.0096 | 0.0054 | 0.0079 |
| KRL | 0.0399 | 0.0391 | 0.0222 | 0.0337 | 0.0393 | 0.0516 | 0.0126 | 0.0345 | 0.0396 | 0.0454 | 0.0174 | 0.0341 |
| KTK | 0.0418 | 0.0461 | 0.0170 | 0.0350 | 0.0394 | 0.0466 | 0.0111 | 0.0324 | 0.0406 | 0.0464 | 0.0140 | 0.0337 |
| MDP | 0.0351 | 0.0389 | 0.0209 | 0.0316 | 0.0338 | 0.0400 | 0.0112 | 0.0283 | 0.0344 | 0.0394 | 0.0160 | 0.0300 |
| MHR | 0.0337 | 0.0416 | 0.0139 | 0.0297 | 0.0344 | 0.0506 | 0.0094 | 0.0315 | 0.0340 | 0.0461 | 0.0116 | 0.0306 |
| ORS | 0.0232 | 0.0357 | 0.0139 | 0.0243 | 0.0168 | 0.0370 | 0.0081 | 0.0206 | 0.0200 | 0.0364 | 0.0110 | 0.0224 |
| PNB | 0.0157 | 0.0184 | 0.0075 | 0.0139 | 0.0252 | 0.0310 | 0.0084 | 0.0215 | 0.0204 | 0.0247 | 0.0079 | 0.0177 |
| RAJ | 0.0120 | 0.0187 | 0.0101 | 0.0136 | 0.0205 | 0.0285 | 0.0079 | 0.0190 | 0.0162 | 0.0236 | 0.0090 | 0.0163 |
| TND | 0.0550 | 0.0456 | 0.0259 | 0.0422 | 0.0472 | 0.0504 | 0.0124 | 0.0367 | 0.0511 | 0.0480 | 0.0192 | 0.0394 |
| UTP | 0.0197 | 0.0159 | 0.0025 | 0.0127 | 0.0326 | 0.0356 | 0.0129 | 0.0270 | 0.0262 | 0.0258 | 0.0077 | 0.0199 |
| WBN | 0.0284 | 0.0222 | 0.0179 | 0.0228 | 0.0313 | 0.0395 | 0.0113 | 0.0274 | 0.0298 | 0.0308 | 0.0146 | 0.0251 |
| Mean | 0.0281 | 0.0286 | 0.0154 | 0.0241 | 0.0290 | 0.0367 | 0.0094 | 0.0250 | 0.0286 | 0.0326 | 0.0124 | 0.0245 |

between regions. For instance, within rural regions, the extent of inequalities during 55th round was as low as 0.0085 in Himachal Pradesh, but was as high as 0.0550 in Tamil Nadu. Further, within urban regions, the extent of inequalities in Maharashtra was as high as 0.0506 during 61st round, but was a mere 0.0094 during 68th round. Similarly, within Uttar Pradesh state, the extent

of inequalities during 68th round was a mere 0.0025 among rural regions, but was comparatively much higher at 0.0129 among urban regions.

However, in order to draw concrete conclusions about differentials in inequalities in calorie intake among the states, the measurements made through the FGT(2) index (Table 5.4.2) were subjected to *three-way ANOVA* technique (with Rounds, States and Regions as the factors). The analysis revealed that within each of rural and urban regions, the averaged inequalities were highly significantly different among the states (for rural regions, $F_{16, 32 \text{ d.f.}} = 9.894$, with a p-value = 3.01×10^{-8} ; and for urban regions, $F_{16, 32 \text{ d.f.}} = 6.057$, with a p-value = 7.96×10^{-6}) as also among the three rounds (for rural regions, $F_{2, 32 \text{ d.f.}} = 28.998$, with a p-value = 6.53×10^{-8} ; and for urban regions, $F_{2, 32 \text{ d.f.}} = 103.628$, with a p-value = 1.05×10^{-14}). However, on an average, the extents of inequalities among rural and urban regions were comparable ($F_{1, 32 \text{ d.f.}} = 2.839$, with a p-value = 0.102). Temporally, the inequalities portrayed an inverted-U pattern. Gravity of the situation on calorie inequalities in south Indian states (like Tamil Nadu, Kerala and Karnataka) was alarming whereas, on the other extreme, the same in certain north-Indian hilly states (like Himachal Pradesh and Jammu & Kashmir) was quite manageable.

5.5. Examining Interlinkages among Health Status, Calorie Inequalities and Per Capita Income among the Major Indian States

In order to examine the nexus, if any, among health status, calorie inequalities and the level of income among the major Indian states, we have sought the help of simple correlation analysis. In fact, the correlation coefficients were obtained indirectly through the simple linear regression analyses (among the three yardsticks, viz., health status, calorie inequalities and per capita income) in the *panel data framework*, with due application of *Hausman's (1978) test*, so as to make a judicious choice between *fixed effects* and Nerlove's version of *random effects modeling*.

Table 5.5.1. Correlation Analysis among Composite Index of Health (CMIN), Index of Calorie Inequality (FGT2) and Per Capita Income (PCIN)

| Pair of Yardsticks | Hausman's test | | Correlation Coefficient | | |
|--------------------|----------------------|---------|-------------------------|------|----------|
| | χ^2 -statistic | p-value | r | d.f. | p-value |
| CMIN & FGT2 | 0.3731 ^{NS} | 0.5413 | -0.3049 [*] | 49 | 0.0298 |
| CMIN & | 0.6372 ^{NS} | 0.4247 | 0.7720 ^{***} | 49 | < 0.0001 |

| | | | | | |
|------------------------|----------------------|--------|-----------------------|----|--------|
| PCIN | | | | | |
| FGT2 & PCIN | 0.6293 ^{NS} | 0.4276 | -0.2512 ^{NS} | 49 | 0.0754 |

^{NS}: Non-significant; *: Significant at 5% probability level; ***: Significant at 0.1% probability level;

In each of the three pairs, non-significance of the χ^2 -statistic for Hausman's test (Table 5.5.1) suggested that the more versatile *random effects modeling* be preferred. As per the modeling, association between the composite index of health status (CMIN) and FGT(2) measure of the inequalities was observed to be indirect ($r = -0.3049$) and statistically significant ($p = 0.0298$). Further, there was a feeble indication of an indirect association between the measure of inequalities and per capita income ($r = -0.2512$; $p = 0.0754$). However, the association between the composite index of health and per capita income was direct ($r = 0.7720$) and very robust (p -value < 0.0001). Thus the states with better health status have shown a very strong tendency to be associated with higher income level and, that; such states would expectedly have less severe incidence of calorie inequalities.

6. Conclusion and Policy Implications

As per the results, the chief determinants of health status have undergone voluminous reshuffling during the study span, from usual demographic indicators to those of physical and social health infrastructure. Within each of rural and urban regions, the averaged inequalities were seen to be highly significantly different among the states as also among the three rounds of NSSO. However, on an average, the extents of inequalities among rural and urban regions were comparable. Further, the states with better health status showed a very strong tendency to be relatively richer and low in the extent of calorie inequalities. Thus, as a policy measure, there is a dire need for shifting priorities in favour of investment on both physical and social health infrastructure, particularly in laggard states and in those states which have undergone a rapid slippage in their ranking. If the state alone cannot shoulder the burden of increased expenditure on this important economic activity, then *public-private-partnership model* needs be propagated.

Improved health conditions would expectedly enhance incomes of the people which, in turn, might bring down the severity of inequalities in calorie intake.

References

- Akbas, Y. and C. Takma (2005). “Canonical Correlation Analysis for Studying the Relationship between Egg Production Traits and Body Weight, Egg Weight and Age at Sexual Maturity in Layers”, *Czech J. Anim. Sci.*, Vol. 50, No.4: 163-8.
- Au, David H., M. B. McDonnell, D. C. Martin, and S. D. Fihn. 2001. “Regional Variations in Health Status.” *Medical Care* 39(8): 879–88.
- Bardhan P K (1974a), “The Pattern of Income distribution in India: A Review”, in T N Srinivasan and P K Bardhan (Eds.), *Poverty and Income Distribution in India*, Statistical Publishing Society, Calcutta: 103-38.
- (1974b), “On the Incidence of Poverty in Rural India in the Sixties”, in T N Srinivasan and P K Bardhan (Eds.), *Poverty and Income Distribution in India*, Statistical Publishing Society, Calcutta: 264-80.
- Chadha, G.K. (2007), “Changing Structure of Demand for Agricultural Commodities: Preparing for the Future”, *Indian Journal of Agricultural Market*, Vol. 21(1):1-42.
- Chatterjee, G S, and N Bhattacharya (1974), “On Disparities in Per Capita Household Consumption in India” in T R Srinivasan and P K Bardhan (Eds.), *Poverty and Income Distribution in India*, Statistical Publishing Society, Calcutta: 337-68.
- Dandekar, V M and N Rath (1971), “Poverty in India”, Indian School of Political Economy. Pune. (First published in *Economic and Political Weekly*, Vol. 6 (1 & 2).
- Das, A. (1999). “Socio-Economic Development in India: A Regional Analysis”, *Development and Society*, Vol. 28(2):313-45.
- Dasgupta, R (1983), “Estimation of Nutritional Intakes”. *Economic and Political Weekly*, Vol. 18 (28): 1238-45.
- (1984), “Nutritional Situation in India — A Statistical Analysis”, *Economic and Political Weekly*, Vol. 19 (34): 1491-4.
- Datta, T. N., & B. K. Ganguly. (2002). “An analysis of consumer expenditure pattern in Indian states with special reference to milk and milk products.” *Indian Dairyman*, 54(9), 47–56.
- Foster J, J Greer, and E Thorbecke (1984), “Notes and Comments : A Class of Decomposable Poverty Measures” *Econometrica*, Vol. 52(3): 761-6.
- Foster J, J Greer, and E Thorbecke (2010), “The Foster-Greer-Thorbecke Poverty Measures: Twenty Five Years Later” *Institute for International Economic Policy Working Paper Series*, The George Washington University.
- Ghassemi, H. (1972). “Food Intake Study of Pre-school Children”, *University of Tehran*, School of Public Health, Tehran.
- Gilbert, P. and E. Meijer (2012), “Package ‘tsfa’, contributed to R-Project (<http://tsanalysis.r-forge.r-project.org/>)
- Giri, A K (2006), “Cereal Consumption Over Time in the Country and Across the States”, *Indian Journal of Agricultural Economics*, Vol. 61(3): 389-98.
- Goodrich, B. (2012), “The Package ‘FAiR’”, contributed to R-Project (<http://wiki.r-project.org/rwiki/doku.php?id=packages:cran:fair>)

- Gorusch, R.L. (1974), *Factor Analysis* (Philadelphia: Saunders, Ken & Georgie).
- Gopalon, C (1970), "Some Recent Studies in the Nutrition Research Laboratories, Hyderabad", *American Journal of Clinical Nutrition*, Vol. 23(1): 35-51.
- Gopalan C, B V Ramasastri and S C Balasubramanian (1971), "Nutritive Value of Indian Foods — Health and Fitness", National Institute of Nutrition (Indian Council of Medical Research), Hyderabad.
- Gupta, D. (2009), "Disparities in Development, Status of Women and Social Opportunities: Indian Experience", *Journal of Alternative Perspectives in the Social Sciences*, Vol. 1(3): 687-719.
- Hausman, E. (1978). "Specification Tests in Econometrics", *Econometrica*, 46(1): 1251-71.
- Hotelling, H. (1935). "The Most Predictable Criterion," *Journal of Education Psychology*, Vol.26: 139-42.
- Kathuria, V. and D. Sankar (2005), "Inter-State Disparities in Health Outcomes in Rural India: An Analysis Using a Stochastic Production Frontier Approach", *Development Policy Review*, Vol. 23 (2): 145-63.
- Kumar, P., Mruthyunjaya, & M. M. Dey. (2007). "Long-term changes in Indian food basket and nutrition." *Economic and Political Weekly*, 42(35), 3567–3572.
- Kumar, N. P. (2011). "Status of Women and Economic Development: A decadal Analysis Based on Indian States", *Socio-economic Voices*: 1-23.
- Kumari, R. (2011). "Inter-District Disparity in Health Care Facility and Education: A Case of Uttar Pradesh", *Journal of Education and Practice*, Vol. 2(1): 38-56.
- Kurian, N. J. (2000), "Widening Regional Disparities in India – Some Indicators", *Economic and Political Weekly*, 35(7) (Feb. 12-18): 538-50.
- Lindeman, R.H., P.F. Merenda and R.Z. Gold (1980). *Introduction to Bivariate and Multivariate Analysis* (London, Foresman and Co.).
- Mendes, M., A. Karabayir, I.E. Erbil and T. Savas (2005), "The relationship among pre-and post slaughter traits of American Bronze Turkey. *Archiv fur Tierzucht-Archives of Animal Breeding* , Vol. 48: 283-9.
- Mittal, S (2007), "What Affect Changes in Cereal Consumption?" *Economic and Political Weekly*, Vol. 42(5): 444-7.
- Musebe, O. R., & P. Kumar. (2006). "Food expenditure pattern of rural household in Andhra Pradesh." *Indian Journal of Agricultural Marketing*, 20(1), 131–139.
- Nair, K.R.G (2007). "Malnourishment among Children in India: Regional Analysis." *Economic and Political Weekly*, Vol. XLII (37): 3797-804.
- Nasurudeen, P, A Kuruvilla, R Sendhil, and V Chandrasekar (2006), "The Dynamics and Inequality of Nutrient Consumption in India" *Indian Journal of Agricultural Economics* Vol. 61(3): 362-73.
- Nerlove, M. (1971). "Further Evidence on the Estimation of Dynamic Economic Relations from a Time-Series of Cross-Sections", *Econometrica*, Vol. 39: 359–82.
- OECD (2008), *Handbook on Constructing Composite Indicators: Methodology and User Guide* (Ispra, Italy: Joint Research Centre of the European Commission).
- Osberg, L and Kuan, Xu (2008), "How Should We Measure Poverty in a Changing World? Methodological Issues and Chinese Case Study" *Review of Development Economics*, Vol. 12(2): 419-41.

- Pathak, P. and K. K. Gaur (1997). "A Composite Index-based Approach for Analysis of the Health System in the Indian Context", *Indian Institute of Health Management Research, Working Paper No. 2*: 1-26.
- Randhawa, G. S., & S. S. Chahal. (2005). "Consumption pattern of milk and milk products in rural Punjab." *Indian Journal of Agricultural Economics*, 61(3), 141–149.
- Rao S.K. (1971). "Inter-regional Variations in Agricultural Growth, 1952-53 to 1964-65: A Tentative Analysis in Relation to Irrigation", *Economic and Political Weekly*, Vol. 6: 1333-46.
- Rao, S.K. (1973). "A Note on Measuring Economic Distances Between Regions in India", *Economic and Political Weekly*, April 1973.
- Rao, V. K. R. V. 1981. "Some Nutritional Puzzles: A Note." *Economic and Political Weekly* 16(28/29): 1205+1207–8.
- Rao, C. H. (2000). "Declining demand for food grains in rural India: Causes and implications." *Economic and Political Weekly*, 35(4), 201–206.
- Ravallion, M (1992), "Does Undernutrition Respond to Income and Prices? Dominance Tests for Indonesia", *The World Bank Economic Review*, Vol. 6 (1): 109-24.
- Raykov, T. and G. A. Marcoulides (2008). *An Introduction to Applied Multivariate Analysis* (New York: Routledge Taylor & Francis Group).
- Roy, T K, S Kumar and Y Vaidehi (2004), "Social Inequalities in Health and Nutrition in Selected States", *Economic and Political Weekly*, Vol. 34 (7): 677-83.
- Sarkar, P.C. (1994). "Regional Imbalances in Indian Economy", *Economic and Political Weekly*, Vol. 29(11): 621-33.
- Sethi, A.S. (2003a), "Determinants of Poverty at the Global Level (With Special Reference to India)", in the *86th Annual Conference Volume of the Indian Economic Association* : 480-91.
- Sethi, A.S. and Sanjeev Kumar (2013), "Co-operative Banking and Rural Development in Punjab – An Examination of Interdependencies through Canonical Correlation Approach", *The Journal of Income and Wealth* , 35 (2): 10-21.
- Sethi, A.S. and R. Pandhi (2011a), "Differential Inequalities in Calorie Intake among Indian States: Evidence From Different Rounds of NSS", *Proceedings of the Annual Conference of the Indian Association for Research in National Income and Wealth* at the Centre for Economic and Social Studies, Hyderabad (March 11-13): Article No B1.
- Sethi, A.S. and R. Pandhi (2011b), "Nutritional Expenditure among Indian States: An Evidence from Divergence Analysis", *Indian Economic Journal* (Special Issue on *Enhancing Human Resources for Inclusive Growth, Employment and Welfare*, released during the 94th Annual Conference of the Indian Economic Association, held at Pune (Dec 27-29, 2010): 783-92.
- Sethi, A. S. and R. Pandhi. (2012a). "Differential inequalities in calorie intake among Indian states: Evidence from different rounds of NSS." *Journal of Indian Association for Research in National Income and Wealth*, 34(1), 30-46.
- Sethi, A. S. and R. Pandhi. (2012b). "Inter-regional differentials in nutritional intake in India: An evidence through cluster analysis." *Poverty and Public Policy*, 4(3), 3-21. Also in *Proceedings of the International Conference on Public Policy and Governance-2012*, Vol-II, 308-318.
- Sethi, A. S. and R. Pandhi. (2013, February). "Identification of major determinants of calorie inequalities: An analysis for Indian states and union territories." Paper presented at the

- International Conference on Human Development and Knowledge Economy, organized by Centre for South-West Asia Study, Punjabi University, Patiala.
- Sethi, A.S. and R. Pandhi (2014), "Interstate Divergences in Nutritional Expenditure in India: A Cluster Analysis Approach", *Poverty and Public Policy*, (Published by Wiley Periodicals, Inc.), 6(1): 80-97.
- Srinivasan, T N (1974), "Income Distribution: A Survey of Policy Aspects", in T N Srinivasan and P K Bardhan (Eds.), *Poverty and Income Distribution in India*, Statistical Publishing Society, Calcutta: 369-96.
- Stewart, D. and W. Love (1968). "A General Canonical Correlation Index", *Psychological Bulletin*, 70: 160-3.
- Tabachnick, B.G. and L.S. Fidell (1989). *Using Multivariate Statistics* (New York, Harper and Row Publisher).
- Sukhatme, P V (1970), "Incidence of Protein Deficiency in Relation to Different Diets in India", *British Journal of Nutrition*, Vol. 24: 477-87.
- World Health Organization (1946). *WHO Definition of Health*, Preamble to the Constitution of the World Health Organization as adopted by the International Health Conference, New York (June 19-22, 1946).