A Structural Equation Model of Elderly Health Care in the Surveyed Households in Kumbakonam Town, Tamil Nadu, India

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Abstract

Introduction: Aged persons are having health problems like hearing and vision decline, muscle strength diminishes, less flexibility of skin and blood vessels, and an overall decline of physique. Their body's organ performs insufficiently with advancing age. Intelligence were also diminishes with age. The social isolation that often exists among older people can dramatically influence mental attitudes and behaviour. The Indian aged population is currently the second largest in the world. The absolute number of the over 60 population in India will increase from 76 million in 2001 to 137 million by 2021. The functional ability of elderly people is crucial to how well they cope with activities of daily living, which in turn affects their quality of life. Therefore, the purpose of this study to examine the causal relationships among factors determining elderly health.

Objectives: 1) to study the socio-economic profile of the aged people, 2) to identify the health status of elderly person, 3) to examine the causal relationships among factors determining elderly health, and 4) to evaluate whether all the measures fit the recommended value, indicating a good fit of the structural model for the collected data.

Methodology: A systematic random sample of 900 elderly people aged above 58 were interviewed from 45 wards, in each ward 20 persons were selected with health problems. The data on different factors were collected through questionnaire survey using 5 point Likert scale method. The causal relationships were established by structural equation modelling (SEM) method using SPSS and AMOS statistical software.

Conclusion: The SEM fitted to the elderly health care data adequately. The results indicated that, the mental health, caring, family environment, life style, health expenditure had significant effect on physiological status (p<0.05). These factors had significant relationship on elderly health.

Key Words: Structural Equation Model, Healthcare Management, Elderly healthcare

I. INTRODUCTION

Aged persons are having health problems like hearing and vision decline, muscular strength diminishes less flexibility of skin and blood vessels, and an overall decline of physique. The organ performs insufficiently with advancing age. Intelligence were also diminishes with age. The social isolation that often exists among older people can dramatically influence mental attitudes and behaviour. The Indian aged population is currently the second largest in the world. The absolute number of the aged over 60 year's population in India will increase from 76 million in 2001 to 137 million by 2021. The functional ability of elderly people is crucial to how well they cope with activities of daily living, which in turn affects their quality of life.

Structural Equation Modelling (SEM) is a multivariate technique, which estimates a series of interrelated dependence relationships simultaneously. The term Structural Equation Modelling conveys that the causal processes under study are represented by a series of structural (i.e. regression) equations, and that these can be modelled pictorially to enable a clearer conceptualization of the study. The hypothesized model can be tested statistically in a simultaneous analysis of the entire system of variables to determine the extent to which it is consistent with the data. If the goodness-of-fit is adequate, the model argues for the plausibility of postulated relations among the variables. Therefore, the purpose of this study is to examine the causal relationships among factors determining elderly health.

II. REVIEW OF LITERATURE

Choice of Health care facility is a complex issue. There are several intricate factors, governed by diverse socio-physical determinants. Access to healthcare is the capacity to pay, clinic operation times, feeling you have the right, ability or interest to access services (Bernard et al., 2004), to the physical setting of healthcare service sites (Tanser, 2006). Access to the healthcare has also been defined on basis of the travel time considering the availability of own vehicle and public transportation especially for elderly (Izumiyama et al, 2007). Access to health care in general can be thought of in a number of ways, including based on geographic location, wait-list times, availability of needed information and service quality (Torgerson et al., 2006).

The public service providers are inexpensive but are overburdened (Shiva Kumar et al., 2011) leading to extremities in high waiting time and underutilization due to poor quality. The elderly, sometimes with their limited mobility, becomes fully dependent on companions to take them to the healthcare units. The Socio demographic factors, social group structure, knowledge about disease are significantly related to the individual's medical orientation. Wirick, (1966) argued that individual's resource is the most significant for demand of healthcare. Transportation availability, costs and facility characteristics also interact with distance in affecting people's use of healthcare (Joseph et al, 1984). The disaggregated level analysis of the people's behaviour when they fall sick reveals that in a given period of illness, the patient makes healthcare decision in stages (Christianson J B, 1976).

The majority of elders are outside the social safety net, and they face economic, health, and emotional insecurity and inequity that pose a challenge to an already overburdened societal system (Alam M, 2004). The new millennium has seen a concerted global effort to mainstream aging into the development agenda, and countries have agreed to link questions of aging to frameworks for social and economic development and human rights (United Nations Report, 2002).

Depressive disorders are the leading cause of burden of disease in high-income countries such as the United States (WHO, 2008). Providing appropriate care to the rapidly growing number of older adults in the U.S. with neuropsychiatric conditions represents a major public health challenge (Jeste et al, 1999). This is particularly true for older Latino adults, who are not only the fastest growing sub-population of older American adults (Angel and Whitfield, 2007; U.S. Census Bureau, 2003), but also have significantly higher rates of unipolar depressive disorders compared to non-Hispanic Whites (Aranda, Lee, and Wilson, 2007; Brennan, Vega, Garcia, Abad & Friedman, 2005; Sorkin, Pham and Ngo-Metzger, 2009).

The research on older people's health has been mostly focused on specific dimension or disease, studied one at a time. Recently, original investigations have made new concepts and perspectives of global health difficulties in the elderly. The first phenotypes of frail adults (Fried et al. 2001) were criticized to be comprised just of physical functioning (Hogan et al. 2003). The following definitions were expanded to include various domains of health, such as such mobility, psychological, cognitive and sensorial problems (e.g. Pel-Little et al. 2009).

While there is controversy concerning what aspects to consider, there is consensus that having health problems is a concept separated from chronic diseases. With this regard, several researches have showed that measures of frailty were associated with mortality independently of illnesses (Puts et al. 2005a, b, c). Salomon et al. (2003) suggested that not selecting medical condition as a domain of multidimensional health was in line with the spirit of the WHO Constitution (1948) and the advancements of the WHO family of classification systems. That is, the diseases are not equated with health status itself, but conceptualized as a possible cause that makes more difficult achieving specific functions or good level of global health (Salomon et al. 2003).

The need to study whether an individual with present multiple problems has been progressively more emphasized (Rockwood et al. 2000, Bortz 2002, Hogan et al. 2003). However, recent research has demonstrated that older people's health cannot be fully described by one global dimension. In fact, such a simple approach misses to describe all the complexity of its multifaceted structure (Brayne et al. 2001, Meinow et al. 2006). In consequence of that, it has become increasingly clear that studying elderly population needs approaches that allow for multiple measures of health to embrace all its complexity (Lafortune 2009, Hallerod 2009). one's state of health is routinely reported to be a major determinant of happiness, particularly in older age groups (Gerstorf et al., 2008), but it is also well established that happy people live longer than unhappy people, even controlling state of health at baseline (Deeg and van Donneveld, 1989; Headey, Hoehne & Wagner, 2013).

Zhang (2002) used SEM to study the relationships between Quality of Healthy Life (QHL), Quality of Objective Environment (QOE) and Psychological Well-Being (PWB). In Keller, Ware and their fellow researchers' paper (1998), the SEM method was used and a second ordered confirmatory factor analysis model was performed to verify their hypothesis about health status indicators with data from ten countries.

Objectives

The objectives of the present study are as follows:

- 1. To study the socio-economic profile of the aged people,
- 2. To identify the health status of elderly person,
- 3. To examine the causal relationships among factors determining elderly health, and
- 4. To evaluate whether all the measures fit the recommended value, indicating a good fit of the structural model for the collected data.

III. METHODOLOGY

Data Collection and Sample

A systematic random sample of 900 elderly people aged above 58 were interviewed from 45 wards, in each ward 20 persons were selected. The data on different factors were collected through questionnaire survey using 5 point Likert scale method (5 indicating strongly agree and 1 indicating strongly disagree) was used for the present study. The collected information is related to socio-economic, cultural, demographic, psychological characters, health status, health problems and healthcare utilization. However, to find out the physiological status height, weight and body mass index were measured.

IV. DATA ANALYSIS

The information collected through the questionnaire has been transformed into 74 selected variables and entered into SPSS for the application of statistical technique. These variables are assumed to be the vital factor in determining the health status of elderly people. Factor analysis was employed for the present data structure and accordingly 22 out of 74 variables were extracted. Further, with the help of Analysis of Moment Structure (AMOS) software package, Structural Equation Modelling (SEM) was used for data analysis to find out the measures fit of recommended value, indicating a good fit of the structural model for the collected data.

V. FINDINGS AND DISCUSSION

Socio-Economic Backgrounds of Elderly

Of the 900 elderly people 50 per cent male and 50 per cent female, 60.4 per cent are aged 58 to 68 years, 27.0 per cent are aged 69 to 78 years, 12.1 per cent are 79 to 89 years and remaining 0.5 per cent are aged up to the 103 years. The elderly unmarried are (2.0%), married (73.2%), widow (18.0%), widower (6.6%) and divorced (0.2%). 82.4 per cent of the respondent's income were Rs.< 100,000, 14.2 per cent were 100,001 to 200,000, 2.6 per cent were 200,001 to 300,000 and 0.8 per cent were > 300,001. Their previous occupations have been as house wife/own work (35.4%), agriculture (1.4%), agricultural labour (1.0%), labour (41.3%), private job (6.9%), government job (8.3%) and business (5.6%).

At present, the elderly economically depend on their daughter (6.4%), son (49.0%), husband/wife (17.2%), pension (8.8%), business (9.6%), aged pension (8.7%) and granddaughter (0.3%). 89.9 per cent are Hindus, 6.0 per cent are Muslims and 4.1 per cent are Christians. They belong to backward (63.6%), most backward (29.1%), scheduled caste (4.8%), scheduled tribe (0.4%) and other (2.1%) communities. Their educational status was elementary (46.2%), middle school (21.8%), high school (16.7%), higher secondary/diploma (2.6%), bachelor degree (1.1%), master degree (0.6%) and illiterate (11.1%). 72.6 per cent and 27.1 per cent of the aged people were have been living in own house and rented house respectively. Their house are thatched roof with mud walls (4.8%), thatched roof with red bricks (10.1%), tiled house with mud wall (3.4%), tiled house with red bricks (46.3%), thokuppu house (2.9%), single storey buildings (30.3%) multi storeys buildings (1.8%) and temple (0.3%).

| Variable | Name of the Variable | Component | | | | | | |
|----------|--------------------------------------|-----------|-------|---|---|---|---|--|
| Number | | 1 | 2 | 3 | 4 | 5 | 6 | |
| Ру 3 | Feeling no one having affection | 0.849 | | | | | | |
| Py 2 | Feeling no one respect by ageing | 0.794 | | | | | | |
| Ру б | Hesitate to disclose health problems | 0.777 | | | | | | |
| Py 5 | Problem for accompanying to hospital | 0.749 | | | | | | |
| Py 10 | Feels family burden | 0.629 | | | | | | |
| Ру 9 | Reasons for not caring | 0.525 | | | | | | |
| Ph 23 | Can you do your work by | | 0.876 | | | | | |

Table-1: Rotated component matrix^a

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| | yourself? | | | | | |
|-------|--|-------|--------|-------|-------|-------|
| Ph 22 | Can you walk without other helps | 0.855 | | | | |
| Ph 25 | Can you do morning work by yourself | 0.824 | | | | |
| Ph 11 | Working without any physical strain | 0.433 | | | | |
| Ph 10 | Who care about your health? | | 0.712 | | | |
| Se 7 | Family Income From | | 0.708 | | | |
| Py 12 | Feels happy playing grandson/daughter | | 0.617 | | | |
| Se16 | Number of Son | | -0.615 | | | |
| Pb 12 | Habits of consuming Alcohol | | | 0.911 | | |
| Pb 16 | Smoking behaviour | | | 0.903 | | |
| Fb 11 | Eating together with family members | | | | 0.670 | |
| Ph 26 | Walking practise | | | | 0.666 | |
| Ph 4 | Present Health Conditions | | | | 0.562 | |
| Ph 14 | Health expenditure (Monthly) | | | | | 0.673 |
| Ph 12 | Physical problems/Diseases | | | | | 0.666 |
| Ph 1 | Weight | | | | | 0.495 |

Extraction method: Principal component analysis.

Rotation method: Varimax with Kaiser Normalizations

^{*a*}. Rotation converged in 4 iterations.

Exploratory Factor Analysis

Factor analysis is used to discover a smaller number of factors underlying larger number of observed variables. The application of factor analysis for the present study is very useful in separating the major dimensions of elderly health conditions. Six dimensions were extracted and contributing a total variance of 56.167 per cent. An Eigen value of 1.0 is taken as a cut-off point to determine the number of dimensions to be extracted. Correlation matrix revealed the presence of many coefficients of 0.4 and above. The Kaiser-Meyer-Oklin (KMO) measure of sampling adequacy value (MSA) is 0.755, exceeding the recommended value of 0.6 and the Barlett's Test of Sphericity reached statistical significance (p<0.001), supporting the factorability of the correlation matrix. Principal components analysis revealed the presence of six components with Eigen values exceeding 1.0.

The correlations between the variable and factor values are shown as the rotated factor loadings in the following table-1. For a good factor solution, a particular variable should load high on one factor and low on all other factors in the rotated factor matrix (Ajai and Sanjaya, 2006). As per the table-1, it can be inferred that out of 74 of selected variables, 22 items are having more than 0.50 factor loadings. These 22 items were taken for further analysis with six components.

Model fit assessment: Structural equation modelling (SEM):

Structural equation modelling was used to analyze the suitability of the model based upon the selected variables. As recommended by Anderson and Gerbing (1988), measurement model to test the reliability and validity of the survey instrument was analyzed first, and by using AMOS version 20 the structural model was analyzed. The structural equation model (SEM) is the most useful when assessing the causal relationship between variables as well as verifying the compatibility of the model used (Peter, 2011).

Structural equation modelling evaluates whether the data fit a theoretical model. In order to evaluate the model, emphasis was given to Chi-square/degrees of freedom (x2/df), CFI, GFI, AGFI, TLI, IFI, RMSEA and PGFI (Table-2). As per the result, Chi square statistics with P = 0.328 which is greater than 0.05, Therefore, it shows a good fit of the model. Consequently, this model is considered for further interpretation in the goodness of fit measures. Common model-fit measures like chi-square/degree of freedom (x2/df), the comparative fit index (CFI), root mean square error of approximation (RMSEA), the normated fit index (NFI), incremental fit index (IFI), and the Tucker Lewis index (TLI) were used to estimate the measurement model fit. Table-2 shows the estimates of the model fit indices from AMOS structural modelling.

| | Fit statistic | Recommended | Obtained |
|----|---|------------------------------------|----------|
| 1 | Chi Square | - | 3.442 |
| 2 | df | - | 3 |
| 3 | Chi Square significance | p = > 0.05 | 0.328 |
| 4 | Chi Square /df | \leq 5.0 (Hair et al., 1998) | 1.147 |
| 5 | Goodness of Fit Index (GFI) | 0.90 (Hair et al. 2006) | 0.999 |
| 6 | Adjusted Goodness of Fit Index (AGFI) | 0.90 (Daire et al., 2008) | 0.990 |
| 7 | Normated Fit Index (NFI) | \geq 0.90 (Hu and Bentler, 1999) | 0.997 |
| 8 | Relative Fit Index (RFI) | 0.90 Hu and Bentler, 1999 | 0.980 |
| 9 | Comparative Fit index (CFI) | 0.90(Hu and Bentler, 1999) | 1.000 |
| 10 | Tucker Lewis Index (TLI) | \geq 0.90 (Hair et al., 1998) | 0.997 |
| 11 | Root mean square error of approximation | < 0.05 (Hair et al., 2006) | 0.013 |
| | (RMSEA) | | |
| 12 | Root Mean Square Residual (RMR) | < 0.02 (Hu and Bentler, 1990) | 0.032 |
| 13 | Parsimony goodness-of-fit index (PGFI) | Within 0.05(Mulaik et al., 1989) | 0.107 |

Table-2: Fit statistics of the Measurement model

AMOS-Output: Compiled by Author

According to Gerbing and Anderson (1992), the criteria for an acceptable model are as follows: RMSEA of 0.08 or lower; CFI of 0.90 or higher; and NFI of 0.90 or higher. The fit between the data and the proposed measurement model can be tested with a chi-square goodness-to-fit (GFI) test where the probability is greater than or equal to 0.9 indicates a good fit (Hu and Bentler, 1999). The GFI of this study was 0.999 more than the recommended value of 0.90 the other measures fitted satisfactorily; AGFI=0.990, CFI=1.000, TLI=0.997, IFI=1.000, NFI=0.997 with x2/df < 3 at 1.47 and RMSEA=0.013 (Bagozzi and Yi, 1988) and RMR=0.032 indicate a good absolute fit of the model. Goodness of fit indices support the model fit and these emphasized indices indicate the acceptability of this structural model. For the purpose of testing the model fit null hypothesis and alternative hypothesis are framed.

Hypothesis

Null hypothesis (H_0): The hypothesized model has a good fit.

Alternate hypothesis (H_1) : The hypothesized model does not have a good fit.

According to the Table-2, it clearly shows that values of all the items are above the suggested value of 0.5 (Hair et al., 2006). According to Bollen (1989a), the higher the probability associated with Chi-square, the closer the fit between the hypothesized model and the perfect fit. The test of our null hypothesis H_0 , is a six-factor structure as shown in Figure-1and 2, give up a chi-square value of 3.442 with 3 degrees of freedom and the calculated P value is 0.328 which is greater than 0.05 which indicates perfectly fit. It is suggesting that the fit of the data to the hypothesized model is entirely adequate. As per the result, Chi square statistics with P = 0.328 shows a good fit of the model. However, this model is considered for further interpretation in the goodness of fit measures.

According to Barbara (2009), both the sensitivity of the Likelihood ratio test to sample size and its basis on the chi-square distribution, which assume that the population (that is, H_0 is correct), have led to problems of fit are now widely known. According to Joreskog and Sorbom (1993), chi-square statistic equals (N-1) F_{min} , (sample size-1, multiplied by the minimum fit function) this value tends to be substantial when the model does not hold and when sample size is large. Barbara (2009) stated that, researchers have addressed the chi-square limitations by developing goodness-of-fit indices that take a more practical approach to the evaluation process. Hair et al. (1998) suggested the value for the fit statistic minimum discrepancy/degrees of freedom (CMIN/DF), otherwise chi-square/ degrees of freedom as ≤ 5.0 . As per the Table-2, the value for the chi-square/degrees of freedom is 1.147 which is less than the accepted cut off value of ≤ 5.0

| Variables | | | Unstandardized co-efficient | S.E. | Standardized co-efficient | Critical Ratio (t) | P Value |
|-------------------------|---|---------------|--------------------------------|------|---------------------------|-----------------------|-----------|
| Physiological Status | < | Mental Health | .032 | .011 | .074 | 2.946 | < 0.003** |
| Physiological Status | < | Caring | .109 | .007 | .383 | 15.360 | < 0.001** |
| Physiological Status | < | Life style | .103 | .040 | .069 | 2.561 | < 0.010** |

 Table-3: Regression Weights: (Group number 1 - Default model)

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| Physiological Status | < | Health Expenditure | .028 | .004 | .166 | 6.679 | < 0.001** |
|-------------------------|---|-------------------------|------|------|------|--------|-----------|
| Physiological Status | < | Family Environment | .197 | .017 | .299 | 11.755 | < 0.001** |
| Elderly Health | < | Physiological Status | .555 | .056 | .527 | 9.924 | < 0.001** |
| Elderly Health | < | Life style | .150 | .057 | .095 | 2.627 | < 0.009** |

Note: ** denotes significant at 1% level

Significance tests of individual parameters

The above table-3 is demonstrating the unstandardized coefficients and associated test statistics. The amount of change in the dependent or mediating variable for each one unit change in the variable predicting it is symbolized by the unstandardized regression coefficient. The Table-3 shows the unstandardized estimate, its standard error (abbreviated S.E.), and the estimate divided by the standard error (abbreviated C.R. for Critical Ratio is t. value). Under the column P, the probability value associated with the null hypothesis that the test is zero is exhibited.

Level of significance for regression weight

The coefficient of *Mental Health* is 0.032 representing the partial effect of *Mental Health* on elderly *Physiological Status*, holding the other variables as constant. The estimated positive sign implies that such effect is positive that elderly *Physiological Status* would increase by 0.032 for every unit increase in *Mental Health* and this coefficient value is significant at 1 per cent level.

The coefficient of *Caring* is 0.109 representing the partial effect of *Caring* on elder's *Physiological Status*, holding the other variables as constant. The estimated positive sign implies that such effect is positive that elder's *Physiological Status* would increase by 0.109 for every unit increase in *Caring* and this coefficient value is significant at 1 per cent level.

The coefficient of *Life Style* is 0.103 representing the partial effect of *Life Style* on elder's *Physiological Status*, holding the other variables as constant. The estimated positive sign implies that such effect is positive that elder's *Physiological Status* would increase by 0.103 for every unit increase in *Life Style* and this coefficient value is significant at 1 per cent level.

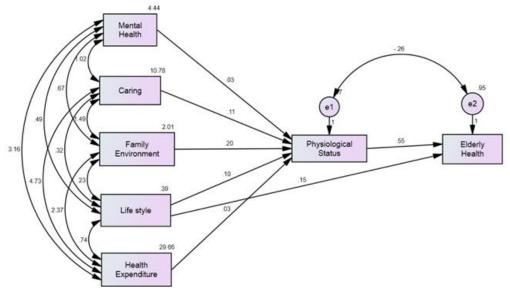


Figure-1: SEM- on Elderly Health: Unstandardized Coefficients

The coefficient of *Health Expenditure* is 0.028 representing the partial effect of *Health Expenditure* on elder's *Physiological Status*, holding the other variables as constant. The estimated positive sign implies that such effect is positive that elder's *Physiological Status* would increase by 0.028 for every unit increase in *Health Expenditure* and this coefficient value is significant at 1 per cent level.

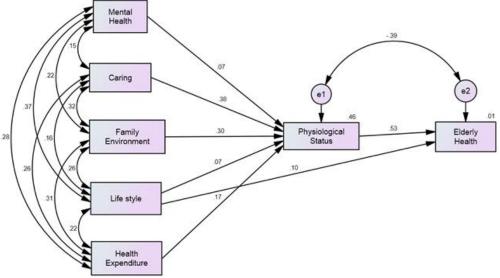


Figure-2: SEM-on Elderly Health: Standardised Coefficients

The coefficient of *Family Environment* is 0.197 representing the partial effect of *Family Environment* on elder's *Physiological Status*, holding the other variables as constant. The estimated positive sign implies that such effect is positive that elder's *Physiological Status* would increase by 0.197 for every unit increase in *Family Environment* and this coefficient value is significant at 1 per cent level.

The coefficient of *Physiological Status* is 0.555 representing the partial effect of *Physiological Status* on *Elderly Health*, holding the other variables as constant. The estimated positive sign implies that such effect is positive that the *Elderly Health* would increase by 0.555 for every unit increase in *Physiological Status* and this coefficient value is significant at 1 per cent level.

The coefficient of *Life Style* is 0.150 representing the partial effect of *Life Style* on *Elderly Health*, holding the other variables as constant. The estimated positive sign implies that such effect is positive that the *Elderly Health* would increase by 0.150 for every unit increase in *Life Style* and this coefficient value is significant at 1 per cent level.

Scalar estimates (group number 1 - default model) Maximum likelihood estimates

Table-3 is also illustrating the standardized estimates for the fitted model. Relative contributions of each predictor variable to each outcome variable can be evaluated by standardized estimates. Figure 1 and 2 show the *Elderly Health Status* structural model. Out of 74 variables, 22 variables were taken for confirmatory factor analysis. As per Figure 1 and 2, it is clear that the elderly health status and its determinant variables.

Confirmatory factor analysis is furthermore known as measurement model. The root mean square error of approximation enlightens us how the model would fit the population covariance matrix (Byrne, 1998). According to Kline (2005), CFI, RMSEA can be utilized along with Chi-Square test to calculate the measurement model fit. As an alternative to Chi-square test, goodness-of-fit statistic (GFI) formed by Joreskog and Sorbom, (1993) is able to calculate the proportion of variance (Tabachnick and Fidell, 2007).

Model can be evaluated with the help of Normated fit index by means of comparing the Chi-square value of the model with Chi-square of the null model (Bentler and Bonnet, 1980). CFI is important in all SEM programs because its measure is least affected by sample size (Fan et al., 1999). According to McDonald and Ho (2002), CFI, GFI, and the NFI are the most frequently used fit indices in structural equation modelling.

VI. CONCLUSION

The aim of this present study is to carry out an experimental analysis of the factors determining the elderly health status. The six factors mental health, caring, family environment, life style, health expenditure and physiological status pertaining to elderly health status model, using a structural equation modelling. This study affirms and develops an instrument of determining factors in the context of elderly health and it examines the relationship among aged people's background. The proposed model is then calibrated using the data collected from aged people in Kumbakonam town of Tamil Nadu. Based on the confirmatory factor analysis, it can be concluded that, the elderly health status scale used in this study adequately fit into the collected data.

As a result, this finding reveals that the hypothesized six-factor model fits the sample data. Based on the feasibility and statistical significance of essential factor estimates; the considerably good fit of the model

shown in Figure 1 and 2 representing an adequate description of elderly health status structure goodness of fit indices support the model fit and these underline indices indicated the adequacy of this structural model. Absolutely, this study will be useful for the population studies, health science, social science and medical researchers to determine the significance of aged people's health factors.

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