

# A Canonical Correlation Analysis of Sectoral **Composition of GDP and Development in Asia**

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# Abstract

This paper identifies the factors that influence percentage contribution of sectors to gross domestic product (GDP) for a group of 32 Asian countries for two cross-section points 1994-96 and 2014-16. Development theories hypothesize that the percentage share of sectors to GDP undergoes transformation with the level of economic development of the country and the degree of competitiveness of its agricultural sector. This paper employed the use of a canonical correlation analysis for 32 Asian countries. This analysis shows that the structural changes in sectoral GDP composition in the selected Asian countries were significantly determined by the factors like employee productivity, employment growth in services sector, rising life expectancy, growth of value added in manufacturing and gross capital formation.

# **Keywords**

GDP, Agriculture, Industry, Services, Development

# **1. Introduction**

Economic development and structural changes in GDP are inter-related. We can see a number of studies related to how the agricultural development is determined by various factors like rural population, life expectancy, foreign direct investment, level of agricultural exports etc. There are studies explaining the growth of service sector in terms of urbanization and per capita income. Similarly, growth of industrial sector is explained in terms of capital formation, foreign direct investment, exports etc. Separate studies on agriculture or industry or services sector are based mainly on multiple regression analysis and excluded the effects of development indicators on the structural changes in the composition of GDP. So this study is based on canonical correlation analysis which is a generalization of multiple regression. In this paper, three response variables are considered simultaneously instead of a single response variable as in multiple regression analysis.

This paper has selected 32 countries which are major countries in terms of population in Asia. Selected countries on the basis of availability of data are Afghanistan, Armenia, Azerbaijan, Bangladesh, Bhutan, Brunei, Cambodia, China, Georgia, Hong Kong, Iran, India, Indonesia, Japan, Kazakhstan, Korea, Rep., Kyrgyz, Malaysia, Maldives, Mongolia, Myamar, Nepal, Pakistan, Philippines, Russia, Singapore, Sri Lanka, Tajikistan, Thailand, Turkey, Turkmenistan, Uzbekistan and Vietnam.

The study has been confined to two cross-section points, 1994-96 and 2014-16. The main reason for this choice is due to missing data for a number of years. There were no data for Afghanistan, Hong Kong, Maldives and Myanmar for the period 1994-96, so these countries are excluded from the analysis of initial period 1994-96. Data for all 32 countries were available for the period 2014-16.

This paper is concerned with an analysis of how the structural changes in sectoral GDP composition are determined by the indicators of development in the context of a group of 32 Asian countries for two cross-section periods 1994-96 and 2014-16. How the variations in development indicators cause variation in sectoral composition of GDP? This question is answered with the help of canonical correlation analysis between the sectoral composition of GDP and the development indicators of 32 Asian countries for two cross-section periods, 1994-96 and 2014-16.

## 2. Literature Review

In the economic literature there are two main schools of thought on how sectoral composition and growth interrelate. The neoclassical view holds that sectoral composition is a relatively unimportant byproduct of growth. However, scholars associated with the world bank, including Kuznets [1], Rostow [2], Chenery and syrquin [3], and Baumol *et al.* [4] posit that growth is brought about by changes in sectoral composition [5].

Lewis theory [6] of economic development explains economic development in terms of structural-change which explains the mechanism of changing structure of underdeveloped economies from subsistence agriculture to more modern and more urbanized. Dual sector theory of Lewis emphasized the importance of agricultural sector in the economy as economic growth progresses. In this theory, industrial sector utilizes the surplus labour in the agricultural sector as its source of growth, along with capital generated by the investment of savings, to expand its production and thus gross output of the economy. As the industrial sector expands in importance, there is a concomitant reduction in the percentage contribution to gross domestic product by the agricultural sector. This growth process thus generally requires the movement of labour from rural areas to the urban areas with a decline of the rural population as a percentage of national population [7] [8].

While recognizing that industrialization is necessary condition for economic

development, there are differing views on sequencing of growth in various sectors. Kaldor, has emphasized that the industrial growth leads to the overall growth. He found a positive correlation between the rates of growth of GDP and the rates of growth of manufacturing output in his study of 12 industrially advanced countries during the period 1953-54 to 1963-64. He observed that the rates of economic growth are almost invariably associated with the fast rate of growth of the secondary sector, mainly, manufacturing [9]. This phenomenon has been so striking to induce some economists to hypothesize that the manufacturing sector is the engine of economic growth, the so-called "engine of growth argument" [9] [10]. Successful industrialization is one aspect of effective development [11].

Wu's study [12] shows that the main determinants of demand for services in India and china are per capita income and urbanization. It is argued that growth of the service sector is determined by several factors such as production specialization, income level and urbanization [13] [14]. These factors are interrelated. As an economy grows, productive activities become more specialized and urbanization accelerates due to the rising level of income. In the meantime, as a result of the increasing specialization of production, firms tend to outsource many service activities such as legal, accounting and security services. Some authors call this process the specialization splintering [15]. It is the main source of demand for services from the producers.

#### 3. Materials and Methods

The main source of data for this study is taken from online statistical database published by United Nations ESCAP and World Bank. In this paper we will use a canonical correlation analysis (CCA) as a technique for determining if there is a relationship between two sets of variables, one measuring sectoral GDP composition and the other measuring development. CCA is a multivariate analysis of correlation between two sets of variables. In CCA, we study interrelationships between sets of multiple predictor variables and multiple response variables.

In a multiple regression analysis, a single variable Y is related to two or more variables X1, X2, ... Xn to see how Y is related to the X variables. From this point of view, canonical correlation analysis is a generalization of multiple regression in which Y variables are simultaneously related to several X variables [16]. In this paper, CCA is applied because the analysis is carried out not on a single response variable, rather 3 multiple response variables.

Hypothesis of interest is change in development indicators cause change in structure of sectoral GDP composition. The null hypothesis is equivalent to testing the hypothesis that all p canonical variate pairs are uncorrelated, or the hypothesis of interest is: Ho:  $\rho * 1 = \rho * 2 = \dots = \rho * p = 0$ ; Ha: Not all pi equal zero.

Response variables (set 2) representing structure of sectoral GDP composition are: 1) agri\_vad, value added from agriculture as % of gdp, 2) ind\_vad, value added from industry as % of gdp and 3) serv\_vad, value added from services, etc., as % of gdp. Three main components of GDP structure are value added from agriculture, value added from industry and value added from services.

Predictor variables (set 1) representing development are: 1) exp, exports of goods and services as % of gdp, 2) fdi, foreign direct investment, net inflows as % of gdp, 3) gcf, gross capital formation as % of gdp, 4) man\_vad, value added from manufacturing as % of gdp, 5) emp\_prod, gdp per person employed (constant 2011 PPP \$), 6) life\_ex, life expectancy at birth, total, in years, 7) upp, urban population as % of total, 8) hepu, public health expenditure as % of gdp, 9) ind\_emp, employment in industry as % of total employment and 10) serv\_emp, employment in services as % of total employment.

Multivariate normality for data sets was conducted using Mardia's and Royston's multivariate normality test. In order to achieve multivariate normality for the response variables, different data transformations had been performed. However, logarithmic transformation was suitable in this case. Variable names are prefixed with l indicating it is in logarithm.

## 4. Empirical Results

Summary Statistics for a group of 29 Asian countries for the period 1994-96 is reported in **Table 1**. Average share of agriculture in GDP was 23.17%, with a minimum 0.17% and maximum 50.72%. Average share of industry in GDP was 33.92%, with a minimum 15.06% and maximum 59.37%. Average share of services in GDP was 42.91%, with a minimum 19.13% and maximum 66.19%. The highest coefficient of variation is observed for the share of agriculture in GDP (61.68%) followed by the share of industry (29.37%) and the least for the services (25.53%) for 1994-96.

Summary Statistics for a group of 32 Asian countries for the period 2014-16 is reported in **Table 1**. Average share of agriculture in GDP was 12.73%, with a minimum 0.04% and maximum 33.26%. Average share of industry in GDP was 32.65%, with a minimum 7.2% and maximum 62.12%. Average share of services in GDP was 54.62%, with a minimum 28.54% and maximum 92.73%. The highest

	Agric	ulture	Indu	ıstry	Serv	ices
-	1994-96	2014-16	1994-96	2014-16	1994-96	2014-16
Mean	23.17	12.73	33.92	32.65	42.91	54.62
Median	21.51	10.40	33.63	32.14	42.07	54.39
Maximum	50.72	33.26	59.37	62.12	66.19	92.73
Minimum	0.17	0.04	15.06	7.20	19.13	28.54
Std. Dev.	14.29	8.99	9.96	11.84	10.96	13.34
Skewness	0.19	0.51	0.44	0.41	0.05	0.74
Kurtosis	-0.75	2.40	0.85	3.93	-0.13	4.00
C.V.	61.68	70.60	29.37	36.30	25.53	24.40

Table 1. Summary statistics of percentage sectoral share of GDP in Asia.

coefficient of variation is observed for the share of agriculture in GDP (70.60%) followed by the share of industry (29.37%) and the least for the services (24.40%).

It is notable that the average share of agriculture in GDP has declined from 23.17% in 1994-96 to 12.73% in 2014-16. The average share of industry has decreased slightly from 33.92% in 1994-96 to 32.65%. On the other hand, the average share of services increased from 42.91% in 1994-95 to 54.62% in 2014-16 (Table 1).

The percentage share of agriculture in GDP is observed higher for Nepal (33.26), Cambodia (28.74), Myanmar (27.59), Tajikistan (25.72), Pakistan (24.85) and Afghanistan (22.26). The percentage share of agriculture in GDP is found to be less than 1% for Singapore and Hong Kong and approximately 1% for Brunei and Japan, less than 3% for Korea Republic. The percentage share of non-agricultural sector in GDP is the lowest for Nepal, Cambodia, Myanmar, Tajikistan, Pakistan and Afghanistan. On the other hand, the share of non-agricultural sector in GDP is above 95% for Singapore, Hong Kong, Brunei, Japan, Korea Republic, Russia and Kazakistan (Figure 1).

The correlation between the variables of sectoral GDP share are moderate, the largest being 0.69 between the share of industrial sector and the share of services sector during 2014-16. Similarly, the correlation between the share of agriculture and the share of services are negative for both periods (Table 2).

Some of the correlations between the development variables are high. High correlations are observed between employee productivity and life expectancy, urban population and employee productivity, urban population and life expectancy, employment in services and employee productivity, and foreign direct investment and the level of exports (**Table 3**). This is true for both periods.

Generally, the correlations between the variables of sectoral share of GDP and development variables are moderate. The share of agriculture has high negative





Year		1994-96			2014-16			
Variable	lagri_vad	lind_vad	lserv_vad	lagri_vad	lind_vad	lserv_vad		
lagri_vad	1.00			1.00				
lind_vad	-0.43	1.00		-0.14	1.00			
lserv_vad	-0.61	-0.04	1.00	-0.35	-0.69	1.00		

Table 2. Correlations among the response variables.

Table 3. Correlations among the development variables.

	lexp	lfdi	lgcf	lman_vad	lemp_prod	llife_ex	lupp	lhepu	lind_emp	lser_emp
1994-96										
lexp	1.00									
lfdi	0.57	1.00								
lgcf	0.29	-0.15	1.00							
lman_vad	0.18	0.10	0.53	1.00						
lemp_prod	0.31	-0.23	0.45	0.29	1.00					
llife_ex	0.21	0.01	0.24	0.46	0.62	1.00				
lupp	0.28	0.02	0.11	0.43	0.74	0.63	1.00			
lhepu	0.03	-0.21	0.29	-0.02	0.30	0.13	0.27	1.00		
lind_emp	0.27	-0.13	0.56	0.71	0.72	0.61	0.68	0.23	1.00	
lser_emp	0.27	-0.10	0.22	0.26	0.85	0.64	0.87	0.32	0.71	1.00
2014-16										
lexp	1.00									
lfdi	0.64	1.00								
lgcf	0.22	0.06	1.00							
lman_vad	0.03	-0.15	0.08	1.00						
lemp_prod	0.44	0.05	0.09	0.24	1.00					
llife_ex	0.47	0.02	0.08	0.15	0.65	1.00				
lupp	0.39	0.09	-0.01	0.24	0.84	0.61	1.00			
lhepu	0.21	-0.12	-0.03	-0.10	0.38	0.58	0.40	1.00		
lind_emp	0.28	0.11	0.09	0.42	0.50	0.35	0.39	0.28	1.00	
lser_emp	0.53	0.25	-0.07	0.17	0.84	0.64	0.80	0.45	0.48	1.00

correlations with employee productivity, life expectancy and urban population for both periods. The share of industry has moderate positive correlation with manufacturing and employee productivity in both periods. However, the share industry had high correlation with gross capita formation in 1994-96. The share of services had moderate positive correlation with life expectancy for both periods (Table 4).

Four multivariate criteria and the F approximations for multivariate test of dimension statistics are presented in **Table 5**. By far the most common method used is Wilk's lamda ( $\lambda$ ) as it tends to have the most general applicability. In our example, the model was statistically significant, with a Wilk's lamda of 0.03, F =

1994-96	lexp	lfdi	lgcf	lman_vad	lemp_prod	llife_ex	lupp	lhepu	lind_emp	lser_emp
lagri_vad	-0.35	0.12	-0.43	-0.31	-0.82	-0.67	-0.59	-0.23	-0.55	-0.69
lind_vad	0.30	-0.04	0.81	0.62	0.55	0.40	0.42	0.27	0.72	0.43
lserv_vad	-0.16	-0.31	0.11	0.01	0.54	0.43	0.26	0.12	0.20	0.46
2014-16										
lagri_vad	-0.50	-0.15	-0.03	-0.17	-0.80	-0.71	-0.68	-0.33	-0.20	-0.74
lind_vad	0.20	0.06	0.36	0.52	0.43	0.06	0.35	-0.14	0.28	0.25
lserv_vad	-0.02	-0.06	-0.37	-0.14	0.15	0.42	0.21	0.43	0.02	0.33

 Table 4. Correlations between response variables and predictor variables.

Table 5. Multivariate statistics and F approximations.

	1994-96				2014-16					
	S = 3 M = 3 N = 6.5					S = 3 M = 3				N = 7
	Value	F Value	Num DF	Den DF	Pr > F	Value	F Value	Num DF	Den DF	Pr > F
Wilks' Lambda	0.03	3.73	30	44.70	< 0.0001	0.03	3.78	30	47.64	< 0.0001
Pillai's Trace	1.85	2.73	30	51	0.0008	1.74	2.49	30	54	0.0017
Hotelling-Lawley Trace	11.49	5.38	30	27.02	< 0.0001	12.62	6.32	30	29.26	< 0.0001
Roy's Greatest Root	9.53	16.19	10	17	< 0.0001	11.18	20.13	10	18	< 0.0001

3.73, df = (30, 44.70) and p < 0.0001 for 1994-96. On the basis of this, we can reject the null hypothesis that there was no relationship between the variable sets and conclude that there probably was a relationship. Using Wilk's lamda,  $1 - \lambda = 1 - 0.03 = 0.97 = r^2$  for the model. All other test statistics are also significant. This means that the model is significant. This is true for both periods.

Now that we have tested the hypothesis of independence and have rejected them, the next step is to obtain estimates of canonical correlation. The estimated canonical correlations are reported in Table 6. In general, the number of canonical dimensions is equal to the number of variables in the smaller set; however, the number of significant dimensions may be even smaller. In this example there are three canonical dimensions of which all of them are not statistically significant. For example for 2014-16, the first test of dimensions tests whether all three dimensions are significant (F = 3.78), the next test tests whether dimensions 2 and 3 are significant (F = 1.34). The last test tests whether dimension 3, by itself, is significant (F = 1.21). These results show that only the first canonical correlation is statistically significant. The last two canonical correlations are not significant. The squared values of the canonical variate pairs, found in the squared canonical correlation column, can be interpreted much in the same way as r<sup>2</sup> values are interpreted. We see that 92% of the variation in  $V_1$  is explained by the variation in W<sub>1</sub>. Only the first canonical correlation is very important. This is also true for 1994-96 where 90% of the variation in  $V_1$  is explained by the variation in  $W_1$  (**Table 6**).

Canonical coefficients are shown in Table 7 and Table 8. The raw canonical

#### Table 6. Canonical correlations.

	Canonical	Adjusted	Approximate	Squared	Eigenvalu	es of Inv€*H =	= CanRsq/(1 –	CanRsq)
	Correlation	lation Canonical Standard Canonical Correlation Error Correlation	Eigenvalue	Difference	Proportion	Cumulative		
				1994-96				
1	0.95	0.93	0.02	0.90	9.53	8.12	0.83	0.83
2	0.76	0.67	0.08	0.58	1.40	0.84	0.12	0.95
3	0.60	0.50	0.12	0.36	0.56		0.05	1.00
				2014-16				
1	0.95	0.94	0.02	0.92	11.18	10.28	0.89	0.88
2	0.69	0.53	0.10	0.47	0.90	0.36	0.07	0.96
3	0.59	0.52	0.12	0.35	0.54		0.04	1.00

(b)

		Test of H0: The canonical co	orrelations in the curren	t row and all that follow ar	e zero
	Likelihood Ratio	Approximate F Value	Num DF	Den DF	Pr > F
			1994-96		
	0.03	3.73	30	44.70	<0.0001
	0.27	1.67	18	32	0.1007
	0.64	1.20	8	17	0.3574
			2014-16		
1	0.03	3.78	30	47.64	<0.0001
2	0.34	1.34	18	34	0.2236
3	0.65	1.21	8	18	0.3458

**Table 7.** Canonical coefficients for the sectoral GDP share variables.

		Raw			Standardized	
1994-96	V1	V2	V3	V1	V2	V3
lagri_vad	-0.29	0.40	1.11	-0.37	0.50	1.4
lind_vad	2.18	2.37	2.13	0.69	0.75	0.67
lserv_vad	0.94	-1.62	4.58	0.26	-0.45	1.28
2014-16						
lagri_vad	-0.47	-0.04	0.84	-0.63	-0.06	1.13
lind_vad	1.94	0.55	4.02	0.72	0.20	1.48
lserv_vad	2.00	-3.87	6.56	0.45	-0.86	1.46

coefficients can be interpreted as same as to interpreting regression coefficients. For example, for 2014-16, for the variable employee productivity, 1 percent increase in employee productivity leads to 0.80 percent increase in the first canonical variate of set 2 when all of the other variables are held constant. Similarly, for 2014-16, for life expectancy, 1 percent increase in life expectancy leads to 2.96 percent increase in the first canonical variate of set 2 when all of the other

	]	Raw		Stan	dardized	
1994-96	W1	W2	W3	W1	W2	W3
lexp	-0.17	-0.09	-1.79	-0.11	-0.06	-1.16
lfdi	0.04	0.10	0.20	0.07	0.20	0.39
lgcf	1.46	2.61	0.76	0.54	0.96	0.28
lman_vad	0.41	-1.81	0.56	0.15	-0.67	0.21
lemp_prod	0.45	-1.11	0.21	0.51	-1.25	0.23
llife_ex	1.88	-3.57	-2.58	0.17	-0.32	-0.23
lupp	0.03	2.06	-1.51	0.02	1.13	-0.83
lhepu	-0.02	-0.23	0.05	-0.01	-0.12	0.02
lind_emp	-0.30	1.96	0.04	-0.16	1.07	0.02
lser_emp	0.36	-1.82	1.76	0.15	-0.74	0.72
2014-16						
lexp	-0.08	0.57	-0.94	-0.06	0.43	-0.71
lfdi	0.13	-0.05	0.11	0.15	-0.06	0.13
lgcf	0.21	1.42	0.49	0.06	0.40	0.14
lman_vad	0.68	0.50	1.165	0.37	0.27	0.63
lemp_prod	0.80	0.75	-0.80	0.76	0.71	-0.76
llife_ex	2.96	-10.99	-4.35	0.19	-0.72	-0.29
lupp	-0.20	0.24	0.70	-0.10	0.11	0.33
lhepu	0.02	-0.39	0.53	0.01	-0.23	0.31
lind_emp	-0.97	0.12	0.86	-0.34	0.04	0.30
lser_emp	0.67	-1.83	1.57	0.24	-0.65	0.56

Table 8. Canonical coefficients for the development measurements.

variables are held constant (**Table 8**). Similarly, high positive raw coefficients are observed for share of manufacturing in gdp (0.68) and employment in services for 2014-16. High negative raw coefficient is also observed for industrial employment (**Table 8**).

However, for 1994-96, gross capital formation and life expectancy emerged to be significant positive determinants of structural change in GDP. Growth of manufacturing, employee productivity and employment in services were also significant determinants of structural change in GDP for 1994-96. For 2014-16, life expectancy, employee productivity, value added in manufacturing, employment in services and gross capital formation were significant positive determinants of structural change in GDP. However, for 2014-16, the impact of growth of industrial employment was negative on structural change in GDP share. This is also evident from standardized coefficients (**Table 8**).

The standardized canonical coefficients are reported in **Table 7** and **Table 8**. The standardized coefficients allow for easier comparisons among the variables when the variables in the model have very different standard deviations. The standardized canonical coefficients are interpreted in a manner as same as to interpreting standardized regression coefficients. For example, consider the variable, the employee productivity, one standard deviation increase in employee productivity leads to 0.76 standard deviation increase in the score on the first canonical variate for set 2 when the other variables in the model are held constant for 2014-16. Similarly, for the variable the share of manufacturing in gdp, one standard deviation increase in the score on the first canonical variate for set 2 when the other share of manufacturing in gdp leads to 0.37 percent increase in the score on the first canonical variate for set 2 when the other variables in the model are held constant (Table 8).

Below are correlations between observed variables and canonical variables which are known as the canonical loadings, which SAS labels as canonical structure. Correlation between the share of agriculture in GDP and their first canonical variable is negative. Similarly, the correlation between the share of industry in GDP and their first canonical variable is positive (**Table 9**) which is true for both periods. Similar picture can also be drawn from correlations between the sectoral GDP share and the canonical variables of the development measurements (**Table 9**).

Correlations between the development measurements and the canonical variables of sectoral GDP share are reported in **Table 10**. Correlation of first canonical variable with employee productivity, urban population, life expectancy, share of employment in service sector, level of export and share of manufacturing in gdp are high and positive for both periods. Similar picture can also be seen in case of correlations between the development measurements and their canonical variables (**Table 10**).

Royston's multivariate normality test for response data sets fulfilled for the period 2014-16, Mardia's and Royston's multivariate normality test fulfilled for the period 1994-96.

Univariate regression analysis has also been carried out to confirm the results from CCA. The percentage share of value added in agriculture was explained negatively by employee productivity and life expectancy for both periods. This model could explain 67 percent variation in the share of agriculture in GDP (Table 11).

Between t The	the Sectora ir Canonica	l GDP Shar al Variables	e and	Between the Sectoral GDP Share and the Canonical Variables of the Development Measurements			
1994-96	V1	V2	V3	W1	W2	W3	
lagri_vad	-0.82	0.46	0.33	-0.78	0.35	0.20	
lind_vad	0.83	0.55	0.02	0.79	0.42	0.01	
lserv_vad	0.46	-0.79	0.40	0.44	-0.60	0.24	
2014-16							
lagri_vad	-0.89	0.22	0.40	-0.85	0.15	0.24	
lind_vad	0.50	0.80	0.32	0.48	0.55	0.20	
lserv_vad	0.18	-0.98	0.05	0.17	-0.68	0.03	

Table 9. Correlations.

	Betv measure Variables	veen the deve ements and th of the Sector	lopment he canonical ral GDP Share	Between the Development Measurements and their Canonical Variables			
1994-96	<b>V</b> 1	V2	V3	W1	W2	W3	
lexp	0.29	0.12	-0.49	0.31	0.16	-0.82	
lfdi	-0.15	0.17	-0.26	-0.16	0.22	-0.44	
lgcf	0.74	0.34	0.08	0.79	0.44	0.133	
lman_vad	0.55	0.31	-0.00	0.57	0.40	-0.00	
lemp_prod	0.83	-0.25	-0.09	0.87	-0.32	-0.15	
llife_ex	0.64	-0.23	-0.12	0.67	-0.30	-0.20	
lupp	0.58	-0.09	-0.21	0.61	-0.13	-0.35	
lhepu	0.30	0.03	0.02	0.31	0.04	0.03	
lind_emp	0.75	0.18	-0.04	0.79	0.22	-0.06	
lser_emp	0.67	-0.23	-0.10	0.71	-0.31	-0.16	
2014-16							
lexp	0.46	0.08	-0.28	0.48	0.12	-0.48	
lfdi	0.11	0.07	-0.17	0.11	0.10	-0.29	
lgcf	0.12	0.40	-0.04	0.12	0.58	-0.08	
lman_vad	0.42	0.23	0.38	0.44	0.34	0.65	
lemp_prod	0.88	0.01	-0.04	0.92	0.01	-0.07	
llife_ex	0.68	-0.31	-0.09	0.71	-0.45	-0.16	
lupp	0.78	-0.07	0.05	0.80	-0.10	0.09	
lhepu	0.30	-0.38	0.06	0.31	-0.55	0.09	
lind_emp	0.34	0.05	0.23	0.35	0.07	0.39	
lser_emp	0.80	-0.19	0.02	0.83	-0.28	0.03	

#### Table 10. Correlations.

 Table 11. Regression estimates for percentage share of value added in agriculture.

			(a)				
		1994-96			2014-16		
Variable	DF	Parameter Estimate	t Value	Pr >  t	Parameter Estimate	t Value	Pr >  t
Intercept	1	27.73	3.48	0.00	42.17	3.29	0.0031
lexp	1	-0.19	-0.83	0.41	-0.22	-1.01	0.3220
lgcf	1	-0.09	-0.20	0.84	0.38	0.74	0.4639
lemp_prod	1	-0.80	-3.60	0.00	-0.84	-2.84	0.0091
llife_ex	1	-4.23	-1.98	0.05	-7.67	-2.36	0.0265
lupp	1	0.35	0.79	0.43	0.16	0.29	0.7738
lhepu	1	0.01	0.04	0.97	0.28	0.96	0.3463

(b)		
	1994-96	2014-16
Root MSE	0.72	0.75
Dependent Mean	2.74	2.13
Coeff Var	26.25	35.50
<b>R-Square</b>	0.73	0.73
Adj R-Sq	0.66	0.67
F Value	10.35	10.82
Pr > F	<0.00	<0.00

Regression results for percentage share of value added in industry shows that gross capital formation was a significant determinant for the change in the share of industry for both periods. However, the growth of manufacturing also found to be significant for 2014-16. The model could explain 50% variation in the share of industry (Table 12).

Regression results for the percentage share of services shows that improvement in life expectancy was a significant positive determinant of expansion of services sector for 2014-16. The model could explain 24% variation in the percentage share of services. However, no variables were found to be significant for 1994-96 (Table 13).

## **5.** Conclusions

For 1994-96, gross capital formation and life expectancy emerged to be significant positive determinants of structural change in GDP. Growth of manufacturing, employee productivity and employment in services were also significant determinants of structural changes in GDP for 1994-96. For 2014-16, improvement in life expectancy, growth of employee productivity, growth of value added in manufacturing, growth of employment in services and gross capital formation were significant positive determinants of structural change in GDP. However, for 2014-16, the impact of growth of industrial employment was negative on the structural change in GDP share.

Univariate regression results show that the most important factors responsible for the transformation of agriculture are the growth in employee productivity

Table 12. Regression estimates for percentage share of value added in industry.

			(a)					
1994-96					2014-16			
Variable	DF	Parameter Estimate	t Value	Pr >  t	Parameter Estimate	t Value	Pr >  t	
Intercept	1	0.83	1.94	0.06	0.21	0.23	0.82	
lgcf	1	0.65	4.72	0.00	0.43	2.05	0.05	
lman_vad	1	-0.01	-0.10	0.92	0.44	4.75	0.00	
lemp_prod	1	-0.05	-0.88	0.38	0.10	0.84	0.40	
lupp	1	0.20	1.75	0.09	-0.03	-0.12	0.90	
lind_emp	1	0.12	0.96	0.34	-0.11	-0.54	0.59	
			(b)					
				1	994-96	2014	-16	
	Root	MSE			0.16	0.:	31	
Dependent Mean				3.48	3.40			
Coeff Var				4.53	9.27			
R-Square				0.80	0.58			
Adj R-Sq				0.75	0.50			
<b>F</b> Value			17.08	6.92				
$\Pr > F$				<0.00		.00		

		(a)					
1994-96					2014-16		
DF	Parameter Estimate	t Value	Pr >  t	Parameter Estimate	t Value	Pr >  t	
1	-0.55	-0.22	0.83	-5.00	-1.78	0.08	
1	0.12	1.57	0.13	-0.11	-1.32	0.19	
1	0.78	1.15	0.26	2.15	2.92	0.00	
1	0.30	1.06	0.30	0.18	0.97	0.33	
1	-0.32	-1.91	0.07	0.05	0.35	0.73	
		(b)					
			1	994-96	2014-16		
Root MSE			0.23		0.20		
Dependent Mean				3.75	3.97		
Coeff Var				6.21	5.20		
R-Square				0.40	0.34		
Adj R-Sq				0.30	0.24		
F Value				3.93	3.63		
$\mathbf{Pr} > \mathbf{F}$				0.01	0.01		
	DF 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1	Jest       Parameter       Base       1       0.705       1       0.12       1       0.78       1       0.30       1       0.30       1       0.30       1       Value       Value       Root       Value       Adj R-Sq       Adj R-Sq       Fr > F	(a) (a) (a) (a) (b) (a) (b) (c) (c) (c) (c) (c) (c) (c) (c	(a) 1994-96 DF Parameter Estimate t Value Pr >  t  1 -0.55 -0.22 0.83 1 0.12 1.57 0.13 1 0.78 1.15 0.26 1 0.30 1.06 0.30 1 -0.32 -1.91 0.07 (b) 1 -0.32 -1.91 0.07 1 Coeff Var Root MSE Dependent Mean Coeff Var R-Square Adj R-Sq F Value Pr > F	(a)         1994-96       Pr >  t        Parameter Estimate         DF       Parameter Estimate       Pr >  t        Parameter Estimate         1       -0.55       -0.22       0.83       -5.00         1       0.12       1.57       0.13       -0.11         1       0.78       1.15       0.26       2.15         1       0.30       1.06       0.30       0.18         1       -0.32       -1.91       0.07       0.05         1       -0.32       -1.91       0.07       0.05         (b)         IPy+96         Root MSE       0.23         Oceff Var       6.21         Good MSE       0.23         Oceff Var       0.40         Add J R-Sq       0.30         Add J R-Sq       0.30         Pr > F       0.01	(a)         2014-16         DF       Parameter Estimate       Value       Pr >  t        Parameter Estimate       Value         1       -0.55       -0.22       0.83       -5.00       -1.78         1       0.12       1.57       0.13       -0.11       -1.32         1       0.78       1.15       0.26       2.15       2.92         1       0.30       1.06       0.30       0.18       0.97         1       -0.32       -1.91       0.07       0.05       0.35         (b)         1994-96       2014         Root MSE       0.23       0.0         Dependent Mean       3.75       3.2         Gotif Var       6.21       5.         R-Square       0.40       0.       0.         Adj R-Sq       0.30       0.01       0.	

Table 13. Regression estimates for percentage value added in services.

and increase in life expectancy which transferred more workforce from agriculture to services. It is notable that improvement in life expectancy was the most important factor responsible for the growth of services sector. Growth of gross capital formation and growth of manufacturing are the most driving factors for the growth of industrial sector.

In order to reduce the contribution of agriculture to GDP and increase the share of non-agricultural sector to GDP, especially, for countries such as Nepal, Cambodia, Myanmar, Tajikistan, Pakistan and Afghanistan, policies to increase life expectancy, employee productivity, employment in services sector, size of manufacturing and gross capita formation are required.

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