# Macro Level Determinants of Poverty: Investigation Through Poverty Mapping of Districts of Pakistan

FARAH SAID, TAREENA MUSADDIQ, and MAHREEN MAHMUD

#### **1. INTRODUCTION**

Successful intervention for social protection of the vulnerable poor first necessitates the identification of the most deprived areas of the country and then an analysis of the factors underlying the prevalence of poverty. A disaggregated, spatial investigation of poverty shows severe regional disparities in the incidence of poverty in Pakistan and often leads us to question whether the determinants of poverty are region based. If so, it behoves us to question whether certain regions are destined to be chronically poorer or is it possible to influence poverty and inequality through planned interventions.<sup>1</sup> Using potential and actual measures of income and wealth, Jamal (2003) shows how regional poverty and inequality has persisted in Pakistan between 1981and 1998; if anything, the gaps between the provinces have increased.

Sen's (1985) capability approach highlights that poverty is multi-dimensional, and indeed there is increasingly a consensus forming in literature that poverty cannot be reduced to a single index. Although income and consumption based indexes are still used and defended by some,<sup>2</sup> others argue that well-being is not completely dependent on these economic measures. Data on income and expenditure tends to be noisy, is often misreported and the link between income and well-being is not always clear. Welfare functions, based on socio-economic factors such as nutrition, wealth, sanitation, education, infrastructure and access to opportunities, are used commonly to explain the incidence of poverty.<sup>3</sup> For this reason, and also due to paucity of relevant data, conventionally used poverty lines based on consumption expenditure are rejected—any threshold for poverty is normative and vulnerable to inflation and shocks, and recent data on alternative measures such as calorie intake has not been available. Instead, this paper develops and uses two indices to determine spatial poverty: one quantifying wealth and asset ownership and the other being a measure of basic household needs.

Farah Said <farahs@lahoreschool.edu.pk> is Teching and Research Fellow, Centre for Research in Economics and Business, Lahore School of Economics, Lahore. Tareena Musaddiq <tareena@lums.edu.pk> is Teaching Fellow, Lahore University of Management Sciences, Lahore. Mahreen Mahmud <mahreenm @lahoreschool.edu.pk> is Senior Teaching and Research Fellow, Centre for Research in Economics and Business, Lahore School of Economics, Lahore.

<sup>1</sup>Amjad and Kamal (1997).

<sup>2</sup>Glewwe and Gaag (1988), Zaidi and Klaas de Vos (2001).

<sup>3</sup>See Henninger (1998), Ravallion (1996), Bellido, *et al.* (1998), Hayati, *et al.* (2006), Booysen, *et al.* (2008), Esposito and Chiappero-Martinetti (2010), Jamal (2009).

The study of the extent and nature of poverty in Pakistan is not a new one. Studies have used both basic needs and calorie-intake measures. Within these some provincial level studies also concentrate on the rural-urban or male-female dimension of the poverty severity front.<sup>4</sup> Cheema, *et al.* (2008) use district representative data from the Multiple Indicators Cluster Survey 2003-04 for Punjab and find concentration of high poverty regions in the South and West of Punjab. Jamal (2009) does the same for Pakistan using household data from PSLM 2004-05 to show that over half of Pakistan's population belongs to poor households. Both Jamal and Cheema make use of Principal Component Analysis (PCA) to form poverty measures. Jamal's study reflects the exacerbation in poverty and inequality when seen in context of an older analysis by Ghaus-Pasha and Jamal (2001) who use a poverty line measure and are able to demonstrate that 30 percent of the population is poor, with an overwhelming 70 percent of them chronically poor.

The contribution of this paper, apart from using recent data (PSLM 2007-08) for a country-wide analysis, is that it delves into uncovering the determinants of poverty econometrically. These determinants will have important poverty alleviation policy implications.

The organisation of the paper is as follows: Section 2 describes the data employed, Section 3 details the methodology used, while results of poverty mapping and regression analysis are in Section 4. Section 5 concludes the paper with a discussion on possible policy implications of the results.

#### 2. DATA

The data employed for the study is The Pakistan Social and Living Standards Measurement Survey (PSLM) 2008-09. PSLM is the latest household survey for Pakistan which is representative at the district level, covering both rural and urban areas, and is complete for all four provinces. PSLM provides a set of representative, population based estimates of social indicators which help in assessing the wellbeing of the population. Since the PSLM is designed to assess the Millennium Development Goals, it provides a range of health, education and physical environment indicators. For the purpose of this study, these indicators are used in constructing the Asset index and Basic Needs index.

The data for the econometric part of the study is gathered from multiple sources. Since the analysis is conducted at the district level, some of the variables are computed from PSLM 08-09 such as urbanisation and dependency ratios for districts. Employment rates are obtained from the Labour Force Survey 2007-08. Additionally, some variables<sup>5</sup> are obtained from Provincial Development Reports of the four provinces and the Population Census of 1998. (Appendix A details the sources of data for the variables employed in the study).

#### **3. METHODOLOGY**

Previous studies on Pakistan have largely focused on identifying micro level determinants with households as the unit of analysis,<sup>6</sup> with fewer studies focusing on the

<sup>&</sup>lt;sup>4</sup>Jamal (2009), Malik (1996), Ali and Tahir (1999).

<sup>&</sup>lt;sup>5</sup>Number of schools, number of hospitals, number of factories and road density. <sup>6</sup>Jamal (2004).

macroeconomics factors contributing towards varying poverty levels in different regions.<sup>7</sup> We take a different approach by using the household level data to estimate our poverty indices and then attempting to identify the macro level factors that determine these estimates.

#### **3.1.** Constructing Indices

The first part of the study entails ranking the districts in terms of poverty along multiple definitions of poverty. As discussed above, income measures can be noisy due to the shocks or cyclical changes in earnings of individuals. It also tends to be deliberately misreported at times due to concerns with tax authorities. Therefore, we concentrate on the wealth and living status of the households which reflects both aggregate income and smoothed out consumption. This will be carried out through the construction of two indices: Asset index and a Basic Needs index using the Principal Component Analysis (PCA) technique. The factors that are included in the construction of both indices will be averaged at the district level from the household level data to arrive at an estimate for each district. This averaging out will also serve the purpose of dealing with a criticism of the PCA approach i.e. the failure of PCA to properly deal with categorical, hence nonnormal, variables in the construction of indices.<sup>8</sup> In this process, the categorical variables, such as those for asset ownership are converted in to averages and therefore into noncategorical values for the district. These indices will serve as the basis of our analysis in the next part and also allow us to map wellbeing at the district level.

Asset index covers a range of durable assets that the household might own (variables used detailed in Table 1 and Table 2). These include assets contributing to a better living environment such as a fan, assets for transportation purposes (motorbike) and assets for communication purposes (television, telephone). Additionally, house ownership is also considered. Ownership of land, livestock etc., is not considered because such variables bias the index between the rural and urban households, since rural households tend to own such assets for sustenance purposes. The Asset index therefore presents a holistic view of asset ownership of the households.

#### Table 1

Variables	Value
Does the household own the house?	=1 if yes, 0 otherwise
Does the household possess an electric fan?	=1 if yes, 0 otherwise
Does the household possess a radio/cassette player?	=1 if yes, 0 otherwise
Does the household possess a television?	=1 if yes, 0 otherwise
Does the household possess a refrigerator?	=1 if yes, 0 otherwise
Does the household possess a cooler?	=1 if yes, 0 otherwise
Does the household possess an air conditioner?	=1 if yes, 0 otherwise
Does the household possess an iron?	=1 if yes, 0 otherwise
Does the household possess a computer?	=1 if yes, 0 otherwise
Does the household possess a bicycle?	=1 if yes, 0 otherwise
Does the household possess a motorcycle/scooter?	=1 if yes, 0 otherwise

Variables used in the Construction of the Asset Index

<sup>7</sup>Akhtar and Ahmad (2003).

<sup>8</sup>Kolenikov and Angeles (2009).

Т	a	bl	le	2

Variables used in the Construction of the Basic Needs index

Variables	Value
Housing Characteristics/Physical	
Environment	
What type of toilet facility does the	=1 if flush system, 0 otherwise
household have?	(Averaged at district level)
What is the main source of drinking water for	=1 if any other source, =2 if Tanker
the household?	Trunk, water fetcher. $=3$ if river, stream
	or pond. =4 if Open well =5 if covered
	well, =6 if water motor, =7 if hand
	pump, =8 if tap (outside home).=9 if tap
	(inside home)
What is the main source of fuel for cooking?	=1 if electricity, gas or oil, 0 otherwise
······································	(Averaged at district level)
What is the main source of fuel for lighting?	=1 if electricity or gas 0 otherwise
	(Averaged at district level)
Does the household have access to telephone?	=1 if mobile or landline 0 otherwise
2 ses die nousenoid have decess to telephone.	(Averaged at district level)
What is the material used in construction of	=1 if burned bricks/blocks_0 otherwise
the walls of the house?	(Averaged at district level)
What is the material used in construction of	=1 if RCC/BCC or cement. 0 otherwise
the roof of the house?	(Averaged at district level)
Health Indicators	(11) or uged at ansate to (or)
Attended births in the district	Number of births in the last 3 years
	attended by doctor, nurse or trained
	midwife/Total number of births in the
	last 3 years
Immunisation Rate of the district	Number of children aged 6 and below
	immunised/Total number of children
	aged 6 and below
Education Indicators	
Gross Primary enrolment rate of the district	Number of children enrolled in primary
	schools/Total number of children aged
	between 3 and 10 years
Gross Secondary enrolment rate of the	Number of children enrolled in
district	secondary schools/Total number of
	children aged between 9 and 15 years
Adult Literacy Rate (Female) of the district	Number of females aged 17 and above
	who can read and write in any language
	with understanding/Total Number of
	females aged 17 and above
Adult Literacy Rate (Male) of the district	Number of males aged 17 and above
	who can read and write in any language
	with understanding/Total Number of
	males aged 17 and above
	0

The Basic Needs index looks at three broad dimensions. Firstly, the physical environment of the households, which includes variables pertaining to the type of dwelling, water and sanitation and access to utilities like electricity and gas. Secondly, immunisation rates amongst children and proportion of attended births are taken as indicators of health. Lastly, educational levels are estimated both by flow measures i.e., enrolment rates at primary and secondary levels and stock measures i.e., male and female adult literacy rates.

#### 3.2. Regression Analysis

In the second part of the study, OLS regression technique will be employed to identify macroeconomic determinants of poverty at a district level for Pakistan. The following specification will be separately estimated with the two indices calculated above as the dependent variable in each:

$$I_i = \alpha_0 + \sum \beta X + \sum \theta Y + \sum \gamma Z + \delta d_i + \varepsilon_i \qquad \dots \qquad \dots \qquad \dots \qquad (1)$$

Where  $I_i$  is the index value for district *i*, *X* is a vector of social service variables in the district, *Y* is a vector of variables capturing the physical development of the district, *Z* is a vector of demographic factors and *d* are provincial dummies.  $\alpha_0$ ,  $\beta$ ,  $\gamma$  and  $\delta$  are regression parameters while  $\varepsilon$  is the error term of the regression.

Difference across provinces can be an important determinant of varying degrees of acquisition of assets and level of well-being of inhabitants even when other factors are similar. Therefore, provincial dummies with Balochistan as the base category are included in the analysis. It is expected that given the lack of development of the province, the districts of other three provinces will have better indices and hence a positive coefficient.

Social services span indicators related to health and educational facilities available in the district. On the educational side, these have been incorporated by using the number of government schools both at the primary and secondary level and health dimension by the number of government hospitals in the district. We take the average number of people per school and people per hospital and hence expect that there would be a negative relation to the dependent variable. Since health and education provisions are expected to impact the current working population with lag, we employ the 1998 census values in the regression analysis. We are not using conventional measures for health and education such as literacy and immunisation rates of the district because these measures have already been used in the construction of basic needs index.

To capture the demographic profile of the district we factor in the overall employment opportunity in the district as indicated by the employed people as a proportion of the total labour force. High employment rates will reflect in better living standards and asset acquisition capabilities and so should result in higher value of indices for these districts. The urbanisation rate is the number of households living in the urban area in a district as a ratio of the total number of households in the district. On the one hand, it can lead to better standard of living and easier access to assets; while on the other it can cause congestion and result in a larger number of people contesting over a few resources. Hence, the expected sign of the coefficient on the variable is ambiguous. Further, the dependency ratio (we take the conventional definition: number of people below16 and above the age of 60 as a ratio of people between 16 and 60 in each household) is included to ascertain if there is any variation in the indices due to the differing burden on the earning hands in a district. Other things remaining the same, the greater the number of mouths to feed, on average, as compared to the hands contributing to the livelihood, the less likely is the household to have a higher level of standard of living and asset accumulation.

The physical dimension will capture factors like industrial development and road access.<sup>9</sup> We use the number of registered factories in the district to proxy for industrial development—this is expected to have a positive relation with the indices. Finally, the ease of access to and from the district is important to the overall development of the district both in terms of facilitating enterprise and businesses and in guaranteeing ready availability of goods. One way of capturing this is the road density of the district, measured by the kilometres of metalled roads as a ratio of the total area of the district.

#### 4. RESULTS

#### 4.1. Spatial Mapping

The construction of the two indices allows us to identify the deprived districts of Pakistan. Poverty maps—the spatial representation of wellbeing and poverty, represented in this case through our basic needs and asset ownership indices—are powerful tools to identify clusters, trends and patterns [Davis (2002)]. They are especially helpful for development practitioners and policy makers in identifying the regions where intervention is needed most and to then track the impact of the said intervention.

Figures 1 and 2 in Appendix C are the poverty maps representing the Basic Needs and Asset indices, respectively. Most of the districts of Punjab and Khyber Pakhtunkhwa lie in the top two quartiles for the Asset index. Likewise, most of the districts lying in the 3rd and 4th quartiles belong to Balochistan and Sindh. This clustering is further intensified for the Basic Needs index, with most of the relatively well-off districts lying in Punjab only and almost 90 percent of the districts from the bottom quartile belonging to Balochistan. Table 3 details the top and bottom ten districts for both indices.<sup>10</sup> The Federal Capital Territory Islamabad ranks the highest from either angle.<sup>11</sup> Its index value for the Asset index (10.53) is almost twice that of the next district in ranking.

As can be seen, seven out of the top ten districts in the basic needs index are from Punjab, two from KPK and one is from Sindh. Interestingly only two of the provincial capitals—Lahore and Karachi—appear in the top ten districts while Quetta is ranked at 20th and Peshawar at 15th. On the other hand, none of the districts of Sindh or Punjab appear in the bottom ten districts, where nine out of the bottom ten districts are from Baluchistan, the remaining one being from KPK.

<sup>&</sup>lt;sup>9</sup>Rupasingha and Goetz (2007).

<sup>&</sup>lt;sup>10</sup>For complete district wise rankings for both indices, see Appendix B.

<sup>&</sup>lt;sup>11</sup>Islamabad being capital of the country is not reported as a district by the Punjab Government. Additionally, index values for Islamabad were exceptionally high and appeared to be an outlier. It was therefore excluded from the analysis.

Table 3
---------

Basic Needs Index					Asse	et Index	
Тор Т	Ten	Bottom	n Ten	Top	Гen	Bottor	n Ten
District	Index	District	Index	District	Index	District	Index
	Value		Value		Value		Value
Karachi	5.90	Awaran	-4.21	Lahore	6.53	Lasbilla	-3.054
Lahore	5.58	Qillah Sai	-4.57	Karachi	5.69	Thatta	-3.12
Rawalpindi	5.50	Chagi	-4.74	Peshawar	5.61	Barkhan	-3.13
Jehlum	4.76	Bolan	-4.90	Rawalpindi	5.10	Badin	-3.17
Sialkot	4.74	Barkhan	-4.94	Jehlum	3.62	Chagi	-3.19
Chakwal	4.73	Musakhel	-5.54	Sialkot	3.31	Tharparkar	-3.73
Abbottabad	4.65	Jhal Magsi	-5.59	Quetta	3.01	Musakhel	-3.80
Haripur	4.35	Kohistan	-5.81	Gujrat	2.71	Awaran	-4.15
Gujrat	4.35	Dera Bugti	-5.82	Gujranwala	2.63	Kohlu	-4.42
Gujranwala	4.11	Kohlu	-6.87	Sargodha	2.52	Kohistan	-4.52

Top and Bottom Ten Districts by Basic Needs and Asset Index

For the Asset index seven out of the top ten districts are from Punjab, and one each from the remaining three provinces. Unlike the Basic Needs index all the provincial capitals appear in the top ten districts for the Asset index. Six districts of Punjab are the same as the Basic Needs index, with one exception being Sargodha that appears in the top ten for Asset index in place of Sialkot. For the bottom ten districts, the distribution is skewed towards districts of Baluchistan but not as much as the Basic Needs index. Three districts from Sindh appear in the lowest ten compared to none for the case of Basic Needs index. Just one district appears from KPK and the remaining six are from Balochistan.

Overall, 72 districts remain in the same quartile whether viewed by the Asset index or the Basic Needs index. As compared to the Asset index however, 13 districts shift 1 quartile down in the Basic Needs index while 18 move up one quartile. More interestingly, however, is the move of more than one quartile between the two indices for some districts. Mansehra, for example, ranked in the third quartile according to the Asset index moves up to the top quartile for the Basic Needs index. Likewise, Batagram moves from the bottom quartile of the Asset index up to the 2nd quartile of the Basic Needs index. It is interesting to note that both the districts moving up two quartiles in Basic Needs as compared to the Asset index are from the KPK.

Five districts, namely Khuzdar, Pishin, Sibi, Qillah Abdullah and Tank, fare worse by two quartiles in terms of basic needs as compared to the Asset index. As can be seen, four out of these five districts are from Balochistan and one from KPK. None of the districts of Sindh and Punjab present such a picture and there are no districts in Punjab which shift places by more than two quartiles for any of the indices.

#### 4.1.1. Spatial Mapping at Provincial Level

Insight into the spatial mapping of indices at the provincial level would help in providing an overview of the results at a more disaggregated level and recognise areas of concern for respective provincial governments. Table 4 below shows the top and bottom three districts of each province and their overall ranking with respect to the entire country.

Ta	bl	le	4
----	----	----	---

Trovince wise Top and Bonom Districts Tisser maen								
	Punjab	Rank	Sindh	Rank	KPK	Rank	Balochistan	Rank
Тор З	Lahore	2	Karachi	4	Peshawer	4	Quetta	8
	Rawalpindi	5	Hyderabad	14	Bannu	12	Pashin	20
	Jhelum	6	Nowshero Feroze	18	Abbottabad	21	Qilla Abdullah	30
Bottom 3	Jhang	71	Thatta	102	Upper Dir	98	Musakhel	107
	Muzaffargarh	82	Badin	104	Shangla	100	Awaran	108
	Rajanpur	88	Tharparkar	106	Kohistan	110	Kohlu	109

*Province wise Top and Bottom Districts—Asset Index* 

Districts encompassing the provincial capitals ranked at the top for each province.<sup>12</sup> The top three districts of Punjab are amongst the top ten of the country. In fact, with the exception of Qilla Abdullah in Balochistan, the top three districts of all provinces belong to the overall top quartile of the Asset index. With the exception of Jhang and Muzaffargarh in Punjab, the bottom districts of all provinces lie in the country-wide bottom quartile. The following table lists the corresponding values and districts for the Basic Needs index.

Once again all the provincial capitals appear in the top three districts of each province. For Punjab and Sindh the top three districts in terms of basic needs are the same as those under the Asset index, as opposed to KPK and Balochistan where changes are seen. Top three districts for all provinces belong to the country-wide top quartile except Baluchistan, where Gawadar and Ziarat lie in the third quartile.

Ta	hl		5
1 a	υı	C.	2

Province wise Top and Bottom Districts—Basic Needs Index

	Punjab	Rank	Sindh	Rank	KPK	Rank	Balochistan	Rank
Тор 3	Lahore	3	Karachi	2	Abbottabad	8	Quetta	20
	Rawalpindi	4	Hyderabad	15	Haripur	9	Gawadar	68
	Jhelum	5	Nowshero Feroze	18	Peshawar	13	Ziarat	76
Bottom 3	D G Khan	66	Thatta	89	Upper Dir	81	Jhal Magsi	107
	Muzaffargarh	70	Badin	91	Shangla	90	Dera Bugti	109
	Rajanpur	82	Tharparkar	95	Kohistan	108	Kohlu	110

None of the lowest three districts of Punjab belong to the country-wide bottom quartile, while the lowest three districts for each of the other provinces lie in it. In the case of both KPK and Sindh, the bottom three districts from the basic needs perspective are the same as those for Asset accumulation. However, there are differences in the rankings by asset accumulation and basic needs for Punjab and Balochistan.

#### 4.2. Regression Results

Estimation of Equation 1 across the two dimensions of poverty under study reveals that the macro determinants of these appear to be similar (results in Table 6).<sup>13</sup> Therefore, we have a combined discussion on the results.<sup>14</sup>

<sup>12</sup>For provincial level analysis the capital, Islamabad, is not considered.

<sup>13</sup>Macro level variables employed in the regression are not available for Balochistan at the district level for the years relevant to the analysis. To overcome the issue of employing out dated data, we use divisional level data for Baluchistan for the year 2006 and therefore include divisions of Balochistan for the purpose of regression analysis, and not districts. The use of divisions rather than districts is reasonable for the case of Balochistan, given the sparsely populated districts in the province relative to other districts of the country. The divisions employed are in line with those defined and used in Burki (2011).

<sup>14</sup>Provincial dummies came out to be insignificant, indicating that differences in provinces are accounted for by the remaining variables. The final results reported do not include the dummies.

Table 6	5
---------	---

Basic Needs Index Asset Index Industrialisation 0.004\*\*\* 0.004\*\*\* (3.29)(4.33)4.87\*\*\* 2.65\*\* Road Density (3.49)(2.51)4.14\* **Employment Rate** 2.01 (1.97)(1.26)Dependency Ratio -6.15-1.09(-0.58)(-0.14)Urbanisation 0.13 0.25 (0.04)(0.12)People to School 0.0002 0.0001 (0.39)(0.32)People to Hospital Beds 0.0002 0.00005 (0.52)(0.20)Constant -3.02-2.78(-0.54)(-0.66)Ν 65 65 Adjusted R<sup>2</sup> 0.36 0.38

**OLS Regression Results** 

t-statistics in parenthesis, \*\*\* p<0.01, \*\* p<0.05, \* p<0.1.

Road density across both measures of poverty is positive and highly significant. It appears to be a very important variable in explaining the pattern of poverty in Pakistan. Therefore, we also map road density of Pakistan (Figure 3 in Appendix B) to try to ascertain if it follows a similar pattern to the spatial pattern of the poverty indices. This can help shed further light on the significance of this particular variable in explaining poverty patterns in the country.

With the exception of Punjab and most of KPK, majority of Sindh and all of Balochistan fare poorly in the availability of road network (measured against land area). Note how the road densities correspond to the poverty maps drawn for the Asset and Basic Need indices. If anything, the road density map provides a starker picture. Excepting Lahore, none of the other districts containing the provincial capitals appear in the upper tier of road density. This only serves to highlight the almost privileged position Punjab seems to hold in terms of access by a metalled road network, followed closely by KPK. As discussed earlier, metalled road density is a key measure of infrastructure development in any district. At the micro level it ensures individual access to and from potential markets, thereby boosting economic activities both in terms of business activity and labour mobility.<sup>15</sup> Linkage with the rest of the country is of utmost importance both at the input and the output end for any business activity will in turn not just benefit the individual but will be a source of employment for the entire region/area. Thus, road network has significant positive spillover effects in the district.

<sup>15</sup>See UNESCAP Report (2008), Van de Walle (1999), Bryceson, et al. (2006).

For people to work outside their hometowns, commuting with ease is necessary. A good road network will facilitate this and would mean that inhabitants are not just restricted to their native areas in seeking employment. This also becomes an important factor in determining the choice of location for an industry. All inputs (both raw material and labour), as well as the end product, will require, at least, ease of access that a good road network provides.

For an agriculturally biased country like Pakistan, especially for those whose main source of livelihood depends on agricultural produce, good roads also allow the transportation of the produce from one area to the other. Perishable agricultural commodities, unless transported in a timely manner will rot and be wasted leading to high economic costs for the producer. A good transport network would mean that this wastage can be reduced and result in higher incomes for these farmers.

A good road network will in general also mean that products from other areas will reach the district with ease and hence without overhead costs. This results in an increase in the availability and variety of consumer durables, facilitating higher asset accumulation. In general, it also allows easy flow of information and results in an integration of that particular area with the rest of the country which, as discussed, benefits the area in many ways. Hence, it is not surprising that the coefficient on the variable is positive and highly significant in explaining districts with superior Asset and Basic Needs indices.

On the physical development side, industrialisation also comes out to be positive and highly significant in both specifications. It captures the level of industrial development in a district and as discussed earlier will benefit the people of the area both in terms of direct and indirect employment generation. This in turn will have an income enhancing impact which would allow greater acquisition of assets as well as higher spending on well-being of the households.

On the demographic side, only employment comes out to be a significant factor in explaining the variation amongst districts for the basic needs index only. This is not surprising since higher employment will be a result of greater business activity (industrial and/or agriculture) in the region. This seems to be resulting in greater ability to spend on education, health and maintaining better living conditions.

Dependency ratio comes out to be an insignificant factor in explaining the variation in the indices across provinces. This might be attributable to the social structure of the country where children from a very young age in poor households start to work and in turn are no longer a burden on the family. They in fact contribute to the livelihood of the family.

Greater urbanisation can have a dual impact: it can result in easier access to assets but it can also have a detrimental impact on the standard of living due to congestion and higher cost of living. Households living in a more urbanised district might be so hardpressed to fulfil their basic needs that in spite of ready availability of consumer durables as well as schools and hospitals, their ability to avail these services and enjoy consumer good may not be any better than their counterparts in less urbanised areas. For those, who are able to afford these goods and services, it is likely to have a positive impact. It appears that neither of these countervailing effects overwhelms the other and hence, on average, the extent of urbanisation has no effect. Finally, social service provision in a district as proxied by the number of people to a hospital and number of people to a school does not explain any variation across districts. This may be due to the quality of public sector services or the possibility that the contribution of the private sector is more meaningful in these areas. District level data on the private sector both for quantity and quality of these services can help us explore this avenue.

#### 5. CONCLUSION

The objective of this paper was to develop a spatial map of poverty for Pakistan based on micro level asset and basic needs indicators. As per *a priori* expectations there is an obvious bias towards the districts of the north (particularly north east), with the exception of the district that contains the largest city of the country (Karachi). Such stark disparities between the provinces, particularly the favourable position held by Punjab, requires investigation into whether it is the result of historical biases, public policy or a combination of both. This would first necessitate looking into the factors influencing well-being in a district.

Econometric analysis indicates that development of infrastructure is a key contributor towards a particular regions relative ability to thrive. This result is further corroborated by the spatial mapping of road density. Once again whether a result of the initial endowment or deliberate public policy, road densities in Punjab are significantly higher than anywhere else in the country, even the district of Karachi.

Econometric results provide a very interesting insight into what potentially influences poverty in Pakistan. Contrary to popular criticism about potential over spending on building infrastructure, it turns out that rather than public sector education and health provision, roads or the lack thereof seem to be the major factor impacting deprivation in the country. Again, as discussed earlier the role of the private sector in social service provision is important to be accounted for in order to present a more complete picture. In addition, active government policy to provide incentives for industries to set up and enterprise to thrive in these marginalised areas would be a step towards pulling these regions out of their current state.

#### **Appendices**

APPENDIX A					
Variable	Source				
Number of Schools	Population Census 1998				
Number of Hospitals	Population Census 1998				
Urbanisation	PSLM 08-09				
Employment Rate	PSLM 08-09				
Industrialisation	Provincial Development Statistics				
Road Density	Provincial Development Statistics				
Dependency Ratio	PSLM 08-09				

### APPENDIX B

## Fig. 1. Spatial Mapping of Asset Index



Fig. 2. Spatial Mapping of Basic Needs Index



## Fig. 3. Spatial Mapping of Road Density



### APPENDIX C

### Asset Index Values

-

	0.8120
1 Istaliadad 10.55576 56 Kilushab 0.576715 75 Jalisholo	-0.8139
2 Lanole 0.51240 59 Milanwan 0.52307 70 Malisheta	-0.88270
A Dechawar 5 605259 41 Veour 0.25251 77 Maching	-0.88932
4 Feshawai 5.005256 41 Kasu 0.275251 76 Masung 5 Devalaindi 5.00941 42 Veral 0.27529 70 Nasunbad	-0.89089
5 Rawaipindi 5.09841 42 Karak $0.22588$ /9 Nasirabad	-1.04458
o Jneium 3.02432 43 Dadu 0.198943 80 Loran	-1.0821
/ Statkot 3.309069 44 Banawaipur 0.194226 81 Shandadkot	-1.12523
8 Quetta 3.006212 45 Nankana Sanib 0.16253 82 Muzaffargarn	-1.12/86
9 Gujrat 2.711241 46 Sahiwal 0.120562 83 Zhob	-1.14092
10 Gujranwala 2.63034 47 Okara 0.093955 84 Gwadar	-1.4655
11 Sargodha 2.523592 48 Pakpattan 0.084401 85 Batagram	-1.53491
12 Bannu 2.44482 49 Narowal -0.04694 86 Panjgur	-1.55675
13 Hyderabad 2.374197 50 Bahawalnagar -0.11433 87 Kalat	-1.57494
14 Faisalabad 2.346325 51 Nawabshah -0.12264 88 Rajanpur	-1.59099
15 Haripur 2.195988 52 Layyah -0.16483 89 Ketch	-1.6377
16   Chakwal   2.193927   53   Lower Dir   -0.2222   90   Tando Muda	-1.69902
17 Nowshero F 1.945039 54 Khuzdar –0.23433 91 Nushki	-1.8301
18 M Bahuaddin 1.836626 55 Larkana -0.27378 92 Qillah Saifullal	h –2.06641
19 Multan 1.776168 56 Lakki Marwat -0.33435 93 Kharan	-2.24803
20 Pashin 1.743047 57 D.G.Khan -0.34163 94 Bolan	-2.33969
21 Abbottabad 1.413793 58 Vehari -0.40333 95 Jhal Magsi	-2.34866
22 Sheikupura 1.402695 59 Kashmore -0.43022 96 Washuk	-2.37476
23 Attock 1.371606 60 Shikarpur -0.44004 97 Chitral	-2.46526
24 Tank 1.350145 61 Sanghar -0.44522 98 Upper Dir	-2.46611
25 Nowshera 1.277339 62 Khanewal -0.46102 99 Dera Bugti	-2.96835
26 Mardan 1.276713 63 Khairpur -0.46932 100 Shangla	-3.03858
27 Hangu 1.245491 64 Ziarat -0.49268 101 Lasbela	-3.05395
28 Swabi 1.156287 65 Lodhran -0.52071 102 Thatta	-3.11718
29 Swat 1.100324 66 RahimYar Khan -0.52165 103 Barkhan	-3.13219
30 Qillah Abd 1.011304 67 Mirpur Khas -0.54322 104 Badin	-3.17392
31 Charsada 0.967917 68 Bhakhar -0.58027 105 Chagi	-3.18545
32 T.T.Singh 0.954534 69 Bonair -0.59717 107 Musakhel	-3.80096
33 D.I.Khan 0.887271 70 Jaccobabad -0.63845 108 Awaran	-4.14663
34 Sibi 0.878156 71 Jhang -0.70223 109 Kohlu	-4.42047
35 Malakand $0.866824$ 72 Maitari $-0.7194$ 110 Kohistan	-4 51882
36 Kohat 0.847362 73 Ghotki -0.74046	4.51002
37 Sukkur 0.835253 74 Jafarabad -0.80344	

Said, Musaddiq, and Mahmud

Well-being Index Values

1   Islamabad   6.234603   38   Karak   0.9754925   75   Jacobabad   -1.226724     2   Karachi   5.903576   39   Bahawalpur   0.881728   76   Ziarat   -1.230156     3   Lahore   5.57656   40   Khanewal   0.8675861   77   Pashin   -1.308706     4   Rawalpindi   5.49893   41   Layyah   0.784137   79   Shahadakto   -1.599927     6   Sialkot   4.736196   43   Ihang   0.7327576   80   Mastung   -1.6499977     7   Chakwal   4.654023   45   Swabi   0.7081991   82   Rajanpur   -1.852241     9   Haripur   4.33773   46   Pakpattan   0.6573209   84   Ketch   -1.9765689     10   Gujrat   4.347954   47   Mardan   0.5231762   85   Khuzdar   -1.979743     12   Sheikupura   3.205434   51   Bahawalnag   0.2612551   88	The being much runes									
2   Karachi   5.903576   39   Bahawalpur   0.8810728   76   Ziarat   -1.291056     3   Lahore   5.57656   40   Khanewal   0.8675861   77   Pashin   -1.508751     4   Rawalpindi   5.49893   41   Layyah   0.7846873   78   Panjgur   -1.508751     5   Jehlum   4.763879   42   Hangu   0.7846873   78   Panjgur   -1.509927     6   Sialkot   4.72863   44   Batagram   0.7327576   81   Upper Dir   -1.699977     8   Abbottabad   4.654023   45   Swabi   0.7081991   82   Rajanpur   -1.852241     9   Haripur   4.337954   47   Mardan   0.5231762   85   Khuzdar   -1.979743     12   Sheikupura   3.291612   49   Dadu   0.4704441   86   Mirpur Khas   -1.98095     13   Peshawar   3.22519   50   Lodhran   0.2328257   89	1	Islamabad	6.234603	38	Karak	0.9754925	75	Jaccobabad	-1.226724	
3   Lahore   5.57656   40   Khanewal   0.8675861   77   Pashin   -1.308706     4   Rawalpindi   5.49893   41   Layyah   0.7846873   78   Panjgur   -1.508751     5   Jehlum   4.763879   42   Hangu   0.7781137   79   Shahdadkot   -1.509927     6   Sialkot   4.736196   43   Jhang   0.7699276   80   Mastung   -1.624826     7   Chakwal   4.72863   44   Batagram   0.7327576   81   Upper Dir   -1.629927     8   Abbottabad   4.654023   45   Swabi   0.7081991   82   Rajanpur   -1.825241     9   Haripur   4.33773   46   Pakpattan   0.691561   83   Lasbela   -1.858197     10   Gujran wala   4.110255   48   Malkand   0.5231762   85   Khuzdar   -1.990595     12   Sheikawar   3.225019   50   Lodrara   0.2289518   87	2	Karachi	5.903576	39	Bahawalpur	0.8810728	76	Ziarat	-1.291056	
4   Rawalpindi   5.49893   41   Layyah   0.784137   78   Panjgur   -1.508751     5   Jehlum   4.763879   42   Hangu   0.784137   79   Shahdadkot   -1.599927     6   Sialkot   4.73619   43   Jhang   0.7699276   80   Mastung   -1.624826     7   Chakwal   4.72863   44   Batagram   0.7321576   81   Upper Dir   -1.699977     8   Abbottabad   4.654023   45   Swabi   0.7081991   82   Rajanpur   -1.852241     9   Haripur   4.337954   47   Mardan   0.5673209   84   Ketch   -1.956589     11   Gujrat   4.347954   47   Mardan   0.5231762   85   Khuzdar   -1.979743     12   Sheikupura   3.291612   49   Dadu   0.4704441   86   Mirpur Khas   -1.980055     13   Byderabad   3.025786   53   Tando Allah   0.2249512   90	3	Lahore	5.57656	40	Khanewal	0.8675861	77	Pashin	-1.308706	
5 Jehlum 4,73879 42 Hangu 0.784137 79 Shahdadkot -1.639927   6 Sialkot 4,738619 43 Jhang 0.7699276 80 Mastung -1.624826   7 Chakwal 4,72863 44 Batagram 0.7327576 81 Upper Dir -1.699977   8 Abbottabad 4.654023 45 Swabi 0.7081991 82 Rajanpur -1.852241   9 Haripur 4.333773 46 Pakpattan 0.6915661 83 Lasbela -1.858197   10 Gujrat 4.347954 47 Mardan 0.5673209 84 Ketch -1.956589   12 Sheikupura 3.225019 50 Lodhran 0.2889536 87 Jafarabad -2.019936   14 Faisalabad 3.025434 51 Bahawalnag 0.2612551 88 Sibi -2.171111   16 Mahuddin 3.035786 53 Tando Allah 0.2249512 90 Shangla -2.420689   17 Attock 2.92887 54	4	Rawalpindi	5.49893	41	Layyah	0.7846873	78	Panjgur	-1.508751	
6   Sialkot   4.736196   43   Jhang   0.7699276   80   Mastung   -1.624826     7   Chakwal   4.73663   44   Batagram   0.7327576   81   Upper Dir   -1.699977     8   Abbottabad   4.654023   45   Swabi   0.7081991   82   Rajanpur   -1.852241     9   Haripur   4.33773   46   Pakpattan   0.6915661   83   Lashela   -1.858197     10   Gujrat   4.347954   47   Mardan   0.5673209   84   Ketch   -1.956589     11   Gujranwala   4.110255   48   Malakand   0.5231762   85   Khuzdar   -1.979743     12   Sheikupura   3.225019   50   Lodhran   0.2889536   87   Jafarabad   -2.01936     14   Faisalabad   3.205434   51   Bahawalnag   0.2612551   88   Sibi   -2.050518     15   Hyderabad   3.087317   52   Shikarpur   0.2332857 <t< td=""><td>5</td><td>Jehlum</td><td>4.763879</td><td>42</td><td>Hangu</td><td>0.784137</td><td>79</td><td>Shahdadkot</td><td>-1.599927</td></t<>	5	Jehlum	4.763879	42	Hangu	0.784137	79	Shahdadkot	-1.599927	
7 Chakwal 4.72863 44 Batagram 0.7327576 81 Upper Dir -1.699977   8 Abbottabad 4.654023 45 Swabi 0.7081991 82 Rajanpur -1.858197   9 Haripur 4.353773 46 Pakpattan 0.6915661 83 Lasbela -1.858197   10 Gujrat 4.347954 47 Mardan 0.5673209 84 Ketch -1.976743   12 Sheikupura 3.291612 49 Dadu 0.4704441 86 Mirpur Khas -1.980095   13 Peshawar 3.225019 50 Lodhran 0.2889536 87 Jafarabad -2.019936   14 Faisalabad 3.005736 53 Tando Allah 0.2249512 90 Shangla -2.420689   15 Hyderabad 3.035786 53 Tando Allah 0.21332857 89 Thatta -2.173114   16 Bahavadin 3.035786 55 Nawabshah 0.1744347 92 Zhob -2.710144   17 Attock 2.9219385 </td <td>6</td> <td>Sialkot</td> <td>4.736196</td> <td>43</td> <td>Jhang</td> <td>0.7699276</td> <td>80</td> <td>Mastung</td> <td>-1.624826</td>	6	Sialkot	4.736196	43	Jhang	0.7699276	80	Mastung	-1.624826	
8   Abbottabad   4.654023   45   Swabi   0.7081991   82   Rajanpur   -1.852241     9   Haripur   4.333773   46   Pakpattan   0.6915661   83   Lasbela   -1.858197     10   Gujrat   4.347954   47   Mardan   0.5673209   84   Ketch   -1.956589     11   Gujramwala   4.110255   48   Malakand   0.523762   85   Khuzdar   -1.979743     12   Sheikupura   3.291612   49   Dadu   0.4704441   86   Mirpur Khas   -1.980095     13   Peshawar   3.225019   50   Lodhran   0.2889536   87   Jafarabad   -2.019361     14   Faisalabad   3.05736   53   Tando Allah   0.2249512   90   Shangla   -2.420669     17   Attock   2.92887   54   Bhakhar   0.2153661   91   Badin   -2.424653     18   Nowshero F   2.919385   55   Nawabhah   0.174347	7	Chakwal	4.72863	44	Batagram	0.7327576	81	Upper Dir	-1.699977	
9   Haripur   4.353773   46   Pakpattan   0.6915661   83   Lasbela   -1.858197     10   Gujrat   4.347954   47   Mardan   0.5673209   84   Ketch   -1.956589     11   Gujranwala   4.110255   48   Malakand   0.5231762   85   Khuzdar   -1.997743     12   Sheikupura   3.225019   50   Lodhran   0.2889536   87   Jafarabad   -2.019936     14   Faisalabad   3.205434   51   Bahawalnag   0.2612551   88   Sibi   -2.019936     15   Hyderabad   3.087317   52   Shikarpur   0.2332857   89   Thatta   -2.173111     16   M Bahuaddin   3.035786   53   Tando Allah   0.2249512   90   Shargla   -2.420689     17   Attock   2.92887   54   Bhakhar   0.2153661   91   Badin   -2.710144     19   T.T.Singh   2.860409   56   Sanghar   0.1087336	8	Abbottabad	4.654023	45	Swabi	0.7081991	82	Rajanpur	-1.852241	
10   Gujrat   4.347954   47   Mardan   0.5673209   84   Ketch   -1.956589     11   Gujranvala   4.110255   48   Malakand   0.5231762   85   Khuzdar   -1.979743     12   Sheikupura   3.221612   49   Dadu   0.4704441   86   Mirpur Khas   -1.980095     13   Peshawar   3.225019   50   Lodhran   0.2889536   87   Jafarabad   -2.019936     14   Faisalabad   3.025743   51   Bahawalnag   0.2612551   88   Sibi   -2.050518     15   Hyderabad   3.087317   52   Shikarpur   0.2332857   89   Thata   -2.173111     16   M Bahuadin   3.035786   53   Tando Allah   0.2249512   90   Shangla   -2.424653     18   Nowshero F   2.919385   55   Nawabshah   0.1744347   92   Zhob   -2.710144     19   T.T.Singh   2.864092   56   Sanghar   0.108736 <td>9</td> <td>Haripur</td> <td>4.353773</td> <td>46</td> <td>Pakpattan</td> <td>0.6915661</td> <td>83</td> <td>Lasbela</td> <td>-1.858197</td>	9	Haripur	4.353773	46	Pakpattan	0.6915661	83	Lasbela	-1.858197	
11 Gujranwala 4.110255 48 Malakand 0.5231762 85 Khuzdar -1.979743   12 Sheikupura 3.291612 49 Dadu 0.4704441 86 Mirpur Khas -1.980095   13 Peshawar 3.225019 50 Lodhran 0.2889536 87 Jafarabad -2.019936   14 Faisalabad 3.025434 51 Bahawalnag 0.2612551 88 Sibi -2.050518   15 Hyderabad 3.087317 52 Shikarpur 0.2332857 89 Thatta -2.173111   16 M Bahuaddin 3.03736 53 Tando Allah 0.2249512 90 Shangla -2.420689   17 Attock 2.92887 54 Bhakhar 0.2153661 91 Badin -2.4214553   18 Nowshero F 2.919385 55 Nawabshah 0.1744347 92 Zhob -2.710144   19 T.T.Singh 2.864092 56 Sanghar 0.1087336 93 Kalat -2.991286   20 Quetta 2.805496	10	Gujrat	4.347954	47	Mardan	0.5673209	84	Ketch	-1.956589	
12 Sheikupura 3.291612 49 Dadu 0.4704441 86 Mirpur Khas -1.980095   13 Peshawar 3.225019 50 Lodhran 0.2889536 87 Jafarabad -2.019936   14 Faisalabad 3.205434 51 Bahawalnag 0.2612551 88 Sibi -2.050518   15 Hyderabad 3.087317 52 Shikarpur 0.2332857 89 Thatta -2.173111   16 M Bahuaddin 3.035786 53 Tando Allah 0.2249512 90 Shangla -2.420689   17 Attock 2.92887 54 Bhakhar 0.2153661 91 Badin -2.424553   18 Nowshero F 2.919385 55 Nawabshah 0.1744347 92 Zhob -2.710144   19 T.T.Singh 2.864092 56 Sanghar 0.1087336 93 Kalat -2.991286   20 Quetta 2.805496 57 Charsada -0.018679 95 Tharparkar -3.023462   21 Nankana Sahib 2.	11	Gujranwala	4.110255	48	Malakand	0.5231762	85	Khuzdar	-1.979743	
13 Peshawar 3.225019 50 Lodhran 0.2889536 87 Jafarabad -2.019936   14 Faisalabad 3.205434 51 Bahawalnag 0.2612551 88 Sibi -2.050518   15 Hyderabad 3.087317 52 Shikarpur 0.2332857 89 Thatta -2.173111   16 M Bahuaddin 3.035786 53 Tando Allah 0.2249512 90 Shangla -2.420689   17 Attock 2.919385 55 Nawabshah 0.1744347 92 Zhob -2.710144   19 T.T.Singh 2.864092 56 Sanghar 0.1087336 93 Kalat -2.991286   20 Quetta 2.805496 57 Charsada -0.013796 94 Nasirabad -3.023462   21 Nankana Sahib 2.523984 58 Maitari -0.018679 95 Tharparkar -3.040952   22 Narowal 2.471714 59 Lakik Marwat -0.046269 96 Nushki -3.69676   23 Sargodha	12	Sheikupura	3.291612	49	Dadu	0.4704441	86	Mirpur Khas	-1.980095	
14Faisalabad3.20543451Bahawalnag0.261255188Sibi-2.05051815Hyderabad3.08731752Shikarpur0.233285789Thatta-2.17311116M Bahuaddin3.03578653Tando Allah0.224951290Shangla-2.42068917Attock2.9288754Bhakhar0.215366191Badin-2.42455318Nowshero F2.91938555Nawabshah0.174434792Zhob-2.71014419T.T.Singh2.86409256Sanghar0.108733693Kalat-2.99128620Quetta2.80549657Charsada-0.01379694Nasirabad-3.02346221Nankana Sahib2.52398458Maitari-0.01867995Tharparkar-3.00495222Narowal2.47171459Lakki Marwat-0.04626996Nushki-3.05698623Sargodha2.4241760RahimYar Khan-0.08953597Lorali-3.20720724Nowshera2.14323961Khairpur-0.20591998Qillah Abd-3.66967625Mianwali2.03270962Jamshoro-0.27942599Kharan-4.03456226Kasur1.938764Larkana-0.305670101Awaran-4.20944728Manshera1.88079565Kashmore-0.379073102Qillah Saifullah-4.571843	13	Peshawar	3.225019	50	Lodhran	0.2889536	87	Jafarabad	-2.019936	
15 Hyderabad 3.087317 52 Shikarpur 0.2332857 89 Thatta -2.173111   16 M Bahuaddin 3.035786 53 Tando Allah 0.2249512 90 Shangla -2.420689   17 Attock 2.92887 54 Bhakhar 0.2153661 91 Badin -2.424553   18 Nowshero F 2.919385 55 Nawabshah 0.1744347 92 Zhob -2.710144   19 T.T.Singh 2.864092 56 Sanghar 0.1087336 93 Kalat -2.991286   20 Quetta 2.805496 57 Charsada -0.018679 94 Nasirabad -3.02462   21 Nankana Sahib 2.523984 58 Maitari -0.018679 95 Tharparkar -3.040952   22 Narowal 2.471714 59 Lakki Marwat -0.04629 96 Nuskit -3.056986   23 Sargodha 2.44217 60 RahimYar Khan -0.025919 98 Kilath -3.207207   24 Nowshera 2.	14	Faisalabad	3.205434	51	Bahawalnag	0.2612551	88	Sibi	-2.050518	
16 M Bahuaddin 3.035786 53 Tando Allah 0.2249512 90 Shangla -2.420689   17 Attock 2.92887 54 Bhakhar 0.2153661 91 Badin -2.424553   18 Nowshero F 2.919385 55 Nawabshah 0.1744347 92 Zhob -2.710144   19 T.T.Singh 2.864092 56 Sanghar 0.1087336 93 Kalat -2.991286   20 Quetta 2.805496 57 Charsada -0.013796 94 Nasirabad -3.023462   21 Nankana Sahib 2.523984 58 Maitari -0.018679 95 Tharparkar -3.040952   22 Narowal 2.471714 59 Lakki Marwat -0.046269 96 Nushki -3.056986   23 Sargodha 2.42417 60 RahimYar Khan -0.089535 97 Lorali -3.207207   24 Nowshera 2.143239 61 Khairpur -0.205919 98 Qillah Abd -3.669676   25 Mianwali <t< td=""><td>15</td><td>Hyderabad</td><td>3.087317</td><td>52</td><td>Shikarpur</td><td>0.2332857</td><td>89</td><td>Thatta</td><td>-2.173111</td></t<>	15	Hyderabad	3.087317	52	Shikarpur	0.2332857	89	Thatta	-2.173111	
17Attock2.9288754Bhakhar0.215366191Badin-2.42455318Nowshero F2.91938555Nawabshah0.174434792Zhob-2.71014419T.T.Singh2.86409256Sanghar0.108733693Kalat-2.99128620Quetta2.80549657Charsada-0.01379694Nasirabad-3.02346221Nankana Sahib2.52398458Maitari-0.01867995Tharparkar-3.04095222Narowal2.47171459Lakki Marwat-0.04626996Nushki-3.05698623Sargodha2.4241760RahimYar Khan-0.08953597Lorali-3.20720724Nowshera2.14323961Khairpur-0.20591998Qillah Abd-3.66967625Mianwali2.03270962Jamshoro-0.27942599Kharan-4.03456226Kasur1.93982463Ghotki-0.301621100Washuk-4.1188827Multan1.938764Larkana-0.0379073102Qillah Saifullah-4.57184329Sukkur1.8005466D.G.Khan-0.446036103Chagi-4.74387430Khushab1.79437967Bonair-0.750567105Barkhan-4.93788832Hafizabad1.58936969D.I.Khan-0.763493106Musakhel-5.54526 <t< td=""><td>16</td><td>M Bahuaddin</td><td>3.035786</td><td>53</td><td>Tando Allah</td><td>0.2249512</td><td>90</td><td>Shangla</td><td>-2.420689</td></t<>	16	M Bahuaddin	3.035786	53	Tando Allah	0.2249512	90	Shangla	-2.420689	
18Nowshero F2.91938555Nawabshah0.174434792Zhob-2.71014419T.T.Singh2.86409256Sanghar0.108733693Kalat-2.99128620Quetta2.80549657Charsada-0.01379694Nasirabad-3.02346221Nankana Sahib2.52398458Maitari-0.01867995Tharparkar-3.04095222Narowal2.47171459Lakki Marwat-0.04626996Nushki-3.05698623Sargodha2.4241760RahimYar Khan-0.08953597Lorali-3.20720724Nowshera2.14323961Khairpur-0.20591998Qillah Abd-3.66967625Mianwali2.03270962Jamshoro-0.27942599Kharan-4.03456226Kasur1.93982463Ghotki-0.301621100Washuk-4.1188827Multan1.938764Larkana-0.305670101Awaran-4.20944728Manshera1.88079565Kashmore-0.379073102Qillah Saifullah-4.57184329Sukkur1.8005466D.G.Khan-0.446036103Chagi-4.74387430Khushab1.79437967Bonair-0.750567105Barkhan-4.93788832Hafizabad1.58936969D.I.Khan-0.763493106Musakhel-5.54526<	17	Attock	2.92887	54	Bhakhar	0.2153661	91	Badin	-2.424553	
19T.T.Singh2.86409256Sanghar0.108733693Kalat-2.99128620Quetta2.80549657Charsada-0.01379694Nasirabad-3.02346221Nankana Sahib2.52398458Maitari-0.01867995Tharparkar-3.04095222Narowal2.47171459Lakki Marwat-0.04626996Nushki-3.05698623Sargodha2.4241760RahimYar Khan-0.08953597Lorali-3.20720724Nowshera2.14323961Khairpur-0.20591998Qillah Abd-3.66967625Mianwali2.03270962Jamshoro-0.27942599Kharan-4.03456226Kasur1.93982463Ghotki-0.301621100Washuk-4.1188827Multan1.938764Larkana-0.305670101Awaran-4.20944728Manshera1.88079565Kashmore-0.379073102Qillah Saifullah-4.57184329Sukkur1.8005466D.G.Khan-0.446036103Chagi-4.74387430Khushab1.79437967Bonair-0.750567105Barkhan-4.93788832Hafizabad1.58936969D.I.Khan-0.763493106Musakhel-5.58670634Swat1.42638371Chitral-0.903833108Kohistan-5.808076 <tr< td=""><td>18</td><td>Nowshero F</td><td>2.919385</td><td>55</td><td>Nawabshah</td><td>0.1744347</td><td>92</td><td>Zhob</td><td>-2.710144</td></tr<>	18	Nowshero F	2.919385	55	Nawabshah	0.1744347	92	Zhob	-2.710144	
20Quetta2.80549657Charsada-0.01379694Nasirabad-3.02346221Nankana Sahib2.52398458Maitari-0.01867995Tharparkar-3.04095222Narowal2.47171459Lakki Marwat-0.04626996Nushki-3.05698623Sargodha2.4241760RahimYar Khan-0.08953597Lorali-3.20720724Nowshera2.14323961Khairpur-0.20591998Qillah Abd-3.66967625Mianwali2.03270962Jamshoro-0.27942599Kharan-4.03456226Kasur1.93982463Ghotki-0.301621100Washuk-4.1188827Multan1.938764Larkana-0.305670101Awaran-4.20944728Manshera1.88079565Kashmore-0.379073102Qillah Saifullah-4.57184329Sukkur1.8005466D.G.Khan-0.446036103Chagi-4.74387430Khushab1.79437967Bonair-0.750567105Barkhan-4.93788832Hafizabad1.58936969D.I.Khan-0.763493106Musakhel-5.5452633Bannu1.42638371Chitral-0.930833108Kohistan-5.80807634Swat1.42638371Chitral-0.907125109Dera Bugti-5.815793 <t< td=""><td>19</td><td>T.T.Singh</td><td>2.864092</td><td>56</td><td>Sanghar</td><td>0.1087336</td><td>93</td><td>Kalat</td><td>-2.991286</td></t<>	19	T.T.Singh	2.864092	56	Sanghar	0.1087336	93	Kalat	-2.991286	
21Nankana Sahib2.52398458Maitari-0.01867995Tharparkar-3.04095222Narowal2.47171459Lakki Marwat-0.04626996Nushki-3.05698623Sargodha2.4241760RahimYar Khan-0.08953597Lorali-3.20720724Nowshera2.14323961Khairpur-0.20591998Qillah Abd-3.66967625Mianwali2.03270962Jamshoro-0.27942599Kharan-4.03456226Kasur1.93982463Ghotki-0.301621100Washuk-4.1188827Multan1.938764Larkana-0.305670101Awaran-4.20944728Manshera1.88079565Kashmore-0.379073102Qillah Saifullah-4.57184329Sukkur1.8005466D.G.Khan-0.446036103Chagi-4.74387430Khushab1.79437967Bonair-0.750567105Barkhan-4.99741231Okara1.58936969D.I.Khan-0.763493106Musakhel-5.5452633Bannu1.43899370Muzaffargarh-0.871703107Jhal Magsi-5.80807634Swat1.42638371Chitral-0.990833108Kohistan-5.80807635Sahiwal1.37491672Tank-0.997125109Dera Bugti-5.815793 <t< td=""><td>20</td><td>Quetta</td><td>2.805496</td><td>57</td><td>Charsada</td><td>-0.013796</td><td>94</td><td>Nasirabad</td><td>-3.023462</td></t<>	20	Quetta	2.805496	57	Charsada	-0.013796	94	Nasirabad	-3.023462	
22 Narowal 2.471714 59 Lakki Marwat -0.046269 96 Nushki -3.056986   23 Sargodha 2.42417 60 RahimYar Khan -0.089535 97 Lorali -3.207207   24 Nowshera 2.143239 61 Khairpur -0.205919 98 Qillah Abd -3.669676   25 Mianwali 2.032709 62 Jamshoro -0.279425 99 Kharan -4.034562   26 Kasur 1.939824 63 Ghotki -0.301621 100 Washuk -4.11888   27 Multan 1.9387 64 Larkana -0.305670 101 Awaran -4.209447   28 Manshera 1.880795 65 Kashmore -0.379073 102 Qillah Saifullah -4.571843   29 Sukkur 1.80054 66 D.G.Khan -0.446036 103 Chagi -4.743874   30 Khushab 1.794379 67 Bonair -0.750567 105 Barkhan -4.997888   32 Hafizabad 1.589369 </td <td>21</td> <td>Nankana Sahib</td> <td>2.523984</td> <td>58</td> <td>Maitari</td> <td>-0.018679</td> <td>95</td> <td>Tharparkar</td> <td>-3.040952</td>	21	Nankana Sahib	2.523984	58	Maitari	-0.018679	95	Tharparkar	-3.040952	
23 Sargodha 2.42417 60 RahimYar Khan -0.089535 97 Lorali -3.207207   24 Nowshera 2.143239 61 Khairpur -0.205919 98 Qillah Abd -3.669676   25 Mianwali 2.032709 62 Jamshoro -0.279425 99 Kharan -4.034562   26 Kasur 1.939824 63 Ghotki -0.301621 100 Washuk -4.11888   27 Multan 1.9387 64 Larkana -0.305670 101 Awaran -4.209447   28 Manshera 1.880795 65 Kashmore -0.379073 102 Qillah Saifullah -4.571843   29 Sukkur 1.80054 66 D.G.Khan -0.446036 103 Chagi -4.743874   30 Khushab 1.794379 67 Bonair -0.749114 104 Bolan -4.997412   31 Okara 1.589369 69 D.I.Khan -0.763493 106 Musakhel -5.54526   33 Bannu 1.426383	22	Narowal	2.471714	59	Lakki Marwat	-0.046269	96	Nushki	-3.056986	
24Nowshera2.14323961Khairpur-0.20591998Qillah Abd-3.66967625Mianwali2.03270962Jamshoro-0.27942599Kharan-4.03456226Kasur1.93982463Ghotki-0.301621100Washuk-4.1188827Multan1.938764Larkana-0.305670101Awaran-4.20944728Manshera1.88079565Kashmore-0.379073102Qillah Saifullah-4.57184329Sukkur1.8005466D.G.Khan-0.446036103Chagi-4.74387430Khushab1.79437967Bonair-0.749114104Bolan-4.90741231Okara1.71530568Gwadar-0.750567105Barkhan-4.93788832Hafizabad1.58936969D.I.Khan-0.763493106Musakhel-5.5452633Bannu1.42638371Chitral-0.930833108Kohistan-5.8070634Swat1.42638371Chitral-0.907125109Dera Bugti-5.81579336Kohat1.16666873Tando Muda-1.118154110Kohlu-6.86542337Vehari1.07447774Lower Dir-1.183369-1.183369	23	Sargodha	2.42417	60	RahimYar Khan	-0.089535	97	Lorali	-3.207207	
25Mianwali2.03270962Jamshoro-0.27942599Kharan-4.03456226Kasur1.93982463Ghotki-0.301621100Washuk-4.1188827Multan1.938764Larkana-0.305670101Awaran-4.20944728Manshera1.88079565Kashmore-0.379073102Qillah Saifullah-4.57184329Sukkur1.8005466D.G.Khan-0.446036103Chagi-4.74387430Khushab1.79437967Bonair-0.749114104Bolan-4.90741231Okara1.71530568Gwadar-0.750567105Barkhan-4.93788832Hafizabad1.58936969D.I.Khan-0.763493106Musakhel-5.5452633Bannu1.43899370Muzaffargarh-0.871703107Jhal Magsi-5.58670634Swat1.42638371Chitral-0.930833108Kohistan-5.80807635Sahiwal1.37491672Tank-0.997125109Dera Bugti-5.81579336Kohat1.16666873Tando Muda-1.118154110Kohlu-6.86542337Vehari1.07447774Lower Dir-1.183369-1.183369	24	Nowshera	2.143239	61	Khairpur	-0.205919	98	Qillah Abd	-3.669676	
26Kasur1.93982463Ghotki-0.301621100Washuk-4.1188827Multan1.938764Larkana-0.305670101Awaran-4.20944728Manshera1.88079565Kashmore-0.379073102Qillah Saifullah-4.57184329Sukkur1.8005466D.G.Khan-0.446036103Chagi-4.74387430Khushab1.79437967Bonair-0.749114104Bolan-4.90741231Okara1.71530568Gwadar-0.750567105Barkhan-4.93788832Hafizabad1.58936969D.I.Khan-0.763493106Musakhel-5.5452633Bannu1.43899370Muzaffargarh-0.871703107Jhal Magsi-5.58670634Swat1.42638371Chitral-0.930833108Kohistan-5.80807635Sahiwal1.37491672Tank-0.997125109Dera Bugti-5.81579336Kohat1.16666873Tando Muda-1.118154110Kohlu-6.86542337Vehari1.07447774Lower Dir-1.183369-1.183369	25	Mianwali	2.032709	62	Jamshoro	-0.279425	99	Kharan	-4.034562	
27Multan1.938764Larkana-0.305670101Awaran-4.20944728Manshera1.88079565Kashmore-0.379073102Qillah Saifullah-4.57184329Sukkur1.8005466D.G.Khan-0.446036103Chagi-4.74387430Khushab1.79437967Bonair-0.749114104Bolan-4.90741231Okara1.71530568Gwadar-0.750567105Barkhan-4.93788832Hafizabad1.58936969D.I.Khan-0.763493106Musakhel-5.5452633Bannu1.43899370Muzaffargarh-0.871703107Jhal Magsi-5.58670634Swat1.42638371Chitral-0.930833108Kohistan-5.80807635Sahiwal1.37491672Tank-0.997125109Dera Bugti-5.81579336Kohat1.16666873Tando Muda-1.118154110Kohlu-6.86542337Vehari1.07447774Lower Dir-1.183369-1.183369	26	Kasur	1.939824	63	Ghotki	-0.301621	100	Washuk	-4.11888	
28Manshera1.88079565Kashmore-0.379073102Qillah Saifullah-4.57184329Sukkur1.8005466D.G.Khan-0.446036103Chagi-4.74387430Khushab1.79437967Bonair-0.749114104Bolan-4.90741231Okara1.71530568Gwadar-0.750567105Barkhan-4.93788832Hafizabad1.58936969D.I.Khan-0.763493106Musakhel-5.5452633Bannu1.43899370Muzaffargarh-0.871703107Jhal Magsi-5.58670634Swat1.42638371Chitral-0.930833108Kohistan-5.80807635Sahiwal1.37491672Tank-0.997125109Dera Bugti-5.81579336Kohat1.16666873Tando Muda-1.118154110Kohlu-6.86542337Vehari1.07447774Lower Dir-1.183369-1.183369	27	Multan	1.9387	64	Larkana	-0.305670	101	Awaran	-4.209447	
29Sukkur1.8005466D.G.Khan-0.446036103Chagi-4.74387430Khushab1.79437967Bonair-0.749114104Bolan-4.90741231Okara1.71530568Gwadar-0.750567105Barkhan-4.93788832Hafizabad1.58936969D.I.Khan-0.763493106Musakhel-5.5452633Bannu1.43899370Muzaffargarh-0.871703107Jhal Magsi-5.58670634Swat1.42638371Chitral-0.930833108Kohistan-5.80807635Sahiwal1.37491672Tank-0.997125109Dera Bugti-5.81579336Kohat1.16666873Tando Muda-1.118154110Kohlu-6.86542337Vehari1.07447774Lower Dir-1.183369-	28	Manshera	1.880795	65	Kashmore	-0.379073	102	Qillah Saifullah	-4.571843	
30Khushab1.79437967Bonair-0.749114104Bolan-4.90741231Okara1.71530568Gwadar-0.750567105Barkhan-4.93788832Hafizabad1.58936969D.I.Khan-0.763493106Musakhel-5.5452633Bannu1.43899370Muzaffargarh-0.871703107Jhal Magsi-5.58670634Swat1.42638371Chitral-0.930833108Kohistan-5.80807635Sahiwal1.37491672Tank-0.997125109Dera Bugti-5.81579336Kohat1.16666873Tando Muda-1.118154110Kohlu-6.86542337Vehari1.07447774Lower Dir-1.183369-	29	Sukkur	1.80054	66	D.G.Khan	-0.446036	103	Chagi	-4.743874	
31Okara1.71530568Gwadar-0.750567105Barkhan-4.93788832Hafizabad1.58936969D.I.Khan-0.763493106Musakhel-5.5452633Bannu1.43899370Muzaffargarh-0.871703107Jhal Magsi-5.58670634Swat1.42638371Chitral-0.930833108Kohistan-5.80807635Sahiwal1.37491672Tank-0.997125109Dera Bugti-5.81579336Kohat1.16666873Tando Muda-1.118154110Kohlu-6.86542337Vehari1.07447774Lower Dir-1.183369	30	Khushab	1.794379	67	Bonair	-0.749114	104	Bolan	-4.907412	
32Hafizabad1.58936969D.I.Khan-0.763493106Musakhel-5.5452633Bannu1.43899370Muzaffargarh-0.871703107Jhal Magsi-5.58670634Swat1.42638371Chitral-0.930833108Kohistan-5.80807635Sahiwal1.37491672Tank-0.997125109Dera Bugti-5.81579336Kohat1.16666873Tando Muda-1.118154110Kohlu-6.86542337Vehari1.07447774Lower Dir-1.183369	31	Okara	1.715305	68	Gwadar	-0.750567	105	Barkhan	-4.937888	
33 Bannu 1.438993 70 Muzaffargarh -0.871703 107 Jhal Magsi -5.586706   34 Swat 1.426383 71 Chitral -0.930833 108 Kohistan -5.808076   35 Sahiwal 1.374916 72 Tank -0.997125 109 Dera Bugti -5.815793   36 Kohat 1.166668 73 Tando Muda -1.118154 110 Kohlu -6.865423   37 Vehari 1.074477 74 Lower Dir -1.183369 -1.183369	32	Hafizabad	1.589369	69	D.I.Khan	-0.763493	106	Musakhel	-5.54526	
34   Swat   1.426383   71   Chitral   -0.930833   108   Kohistan   -5.808076     35   Sahiwal   1.374916   72   Tank   -0.997125   109   Dera Bugti   -5.815793     36   Kohat   1.166668   73   Tando Muda   -1.118154   110   Kohlu   -6.865423     37   Vehari   1.074477   74   Lower Dir   -1.183369   -1.183369	33	Bannu	1.438993	70	Muzaffargarh	-0.871703	107	Jhal Magsi	-5.586706	
35 Sahiwal 1.374916 72 Tank -0.997125 109 Dera Bugti -5.815793   36 Kohat 1.166668 73 Tando Muda -1.118154 110 Kohlu -6.865423   37 Vehari 1.074477 74 Lower Dir -1.183369 -6.865423	34	Swat	1.426383	71	Chitral	-0.930833	108	Kohistan	-5.808076	
36 Kohat   1.166668   73   Tando Muda   -1.118154   110   Kohlu   -6.865423     37 Vehari   1.074477   74   Lower Dir   -1.183369   -6.865423	35	Sahiwal	1.374916	72	Tank	-0.997125	109	Dera Bugti	-5.815793	
37 Vehari 1.074477 74 Lower Dir -1.183369	36	Kohat	1.166668	73	Tando Muda	-1.118154	110	Kohlu	-6.865423	
	37	Vehari	1.074477	74	Lower Dir	-1.183369				

#### REFERENCES

- Ali, S. S. and S. Tahir (1999) Dynamics of Growth, Poverty and Inequality in Pakistan. *The Pakistan Development Review* 38:4, 837–858.
- Akhtar, S. and M. Ahmad (2003) Modeling Poverty Trends in Pakistan: Some Additional Empirical Evidence. Social Policy Development Centre. (Research Report 27).
- Amjad, R. and A. R. Kemal (1997) Macroeconomic Policies and Their Impact on Poverty Alleviation in Pakistan. *The Pakistan Development Review* 36:1, 39–68.
- Bellido, N. P., M. D. Jano, F. J. López Ortega, M. P. Martín-Guzmán, and M. I. Toledo (1998) The Measurement and Analysis of Poverty and Inequality: An Application to Spanish Conurbations. *International Statistical Review/Revue Internationale de Statistique* 66:1, 115–131.
- Booysen, F., S. Van der Berg, R. Burger, M. von Maltitz, and G. Du Rand (2008) Using an Asset Index to Assess Trends in Poverty in Seven Sub-Saharan African Countries. *World Development* 36:6, 1113–1130.

- Bryceson, D. F., A. Bradbury, and T. Bradbury (2006) Roads to Poverty Reduction? Dissecting rural roads' impact on mobility in Africa and Asia. Paper presented at the Conference on Reducing Poverty and Inequality—Centre for the Study of African Economies, Oxford University, 20-21 March, 2006, Oxford.
- Burki, A. A. (2011) Exploring the Links between Inequality, Polarisation and Poverty: Empirical Evidence from Pakistan. (SANEI Working Paper Series No. 11-04).
- Cheema, A., L. Khalid, and M. Patnam (2008) Does the Labour Market Structure Explain Differences in Poverty in Rural Punjab? In Proceedings of the Fourth Annual Conference on Management of the Pakistan Economy. Lahore School of Economics, 24-25 April, 2008, Lahore.
- Davis, B. (2002) Is It Possible to Avoid a Lemon? Reflections on Choosing a Poverty Mapping Method. Agricultural and Development Economics Division of the Food and Agriculture Organisation of the United Nations. (Working Papers 02-07).
- Esposito, L. and E. Chiappero-Martinetti (2010) Multidimensional Poverty: Restricted and Unrestricted Hierarchy among Poverty Dimensions. *Journal of Applied Economics* 13:2, 181–204.
- Ghaus-Pasha, A. and H. Jamal (2001) Incidence of Income Poverty in Pakistan. (Social Policy and Development Centre Report No. 39).
- Glewwe, P. and J. van der Gaag (1988) Confronting Poverty in Developing Countries. Definitions, Information, and Policies. World Bank Living Standards Measurement Study. (Working Paper No. 48).
- Hayati, D., E. Karami and B. Slee (2006) Combining Qualitative and Quantitative Methods in the Measurement of Rural Poverty: The Case of Iran. *Social Indicators Research* 75:3, 361–394.
- Henninger, N. (1998) Mapping and Geographic Analysis of Poverty and Human Welfare—Review and Assessment. Report prepared for the UNEP/CGIAR Initiative on GIS, World Resources Institute, Washington, D.C.
- Jamal, H. (2003) The Changing Profile of Regional Inequality. (Social Policy and Development Centre Report No. 47).
- Jamal, H. (2004) In Search of Poverty Predictors: The Case of Urban and Rural Pakistan. Social Policy Development Centre. (Research Report Number 59).
- Jamal, H. (2009) Estimation of Multidimensional Poverty in Pakistan. (Social Policy and Development Centre Report No. 79).
- Kolenikov, S. and G. Angeles (2005) Socioeconomic Status Measurement with Discrete Proxy Variables: Is Principal Component Analysis a Reliable Answer? *Review of Income and Wealth* 55:1, 128–165.
- Malik, S. (1996) Determinants of Rural Poverty in Pakistan: A Micro Study. *The Pakistan Development Review* 35:2, 171–187.
- Ravallion, M. (1996) Issues in Measuring and Modeling Poverty. *Economic Journal* 106, 1328–1343.
- Rupasingha, A. and S. J. Goetz (2007) Social and Political Forces as Determinants of Poverty: A Spatial Analysis. *The Journal of Socio-Economics* 36:4, 650–671.
- Sen, Amartya K. (1985) Commodities and Capabilities. Oxford: Oxford University Press.

- United Nations Economic and Social Commission for Asia and the Pacific, United Nations Economics and Social Council. (2008) *Transport and Development:* Assessment of the Contribution of Transport to Economic and Social Development. September 2008.
- Van de Walle, D. (1999) Choosing Rural Road Investments to Help Reduce Poverty. (World Bank Working Paper No. 2458).
- Zaidi, M. A. and K. de Vos (2001) Trends in Consumption-based Poverty and Inequality in the European Union during the 1980s. *Journal of Population Economics* 14, 367–390.