

# An Examination of the Determinants of The Mode of Transport to Primary Health Facilities in A Developing Region

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**Abstract** Access to primary health facilities is a key determinant of the overall well being of the population in an area. In rural regions where distances to public facilities are usually longer compared to urban areas, it is not clear if people are still willing to walk to use these facilities. It is pertinent therefore to clarify such uncertainty since walking distance is a standard measure used to plan such public facilities particularly in rural regions. The objective of this study therefore is to provide a framework to determine the factors that will influence a health care service seeker in a developing region to walk or use other means of transport to a primary health facility. The case study for this research is Ijebu North Local Government Area of Ogun state made up of eleven urban and rural wards. One hundred and fifty households were selected at random for interview. Logit regression was used to describe how some predictor variables were used to explain the likelihood of a particular household walking to a primary health facility. The predicting model in this study was able to classify 80.0% of the cases correctly. This simply shows that the predictors (independent variables) contribute to the predicting power of the logistic regression model. The pseudo R-squares of Cox and Snell's R-square and Nagelkerke's R also show that our logistic model is relevant to predicting whether a household will walk or use a vehicle while attending a health facility. In our study, we noted that settlement status ( $p=0.00$ ) and transport cost to health facility ( $p=0.00$ ) contributed significantly to the prediction. This study also reveals that the odds for household members in an urban area to walk to the health facility often used is 88.1% lower than the odds for a household in a rural area. It was revealed that households that are poor are 49% times more likely to walk to the health facility they frequently used. The knowledge of the factors that will determine whether health care service seekers in a developing region will want to walk or not will assist government in the planning and provision of health facilities.

**Keywords:** Developing Region, Accessibility, Primary Health Facilities, Logistic Modelling, Transport modes

**Abstrak** Akses ke fasilitas kesehatan yang bersifat primer merupakan faktor penentu kesejahteraan dari penduduk di suatu daerah. Pada wilayah pedesaan dengan jarak ke fasilitas umum yang lebih jauh dibandingkan perkotaan, ketersediaan orang-orang berjalan untuk menggunakan fasilitas-fasilitas tersebut menjadi tidak jelas. Hal ini sangat relevan karena untuk mengklarifikasi ketidakpastian tersebut, dimana berjalan kaki menjadi ukuran standar yang digunakan untuk merencanakan fasilitas umum di daerah pedesaan. Penelitian ini bertujuan untuk menyediakan kerangka pemikiran untuk menentukan faktor yang berpengaruh terhadap pencari layanan kesehatan pada wilayah yang sedang berkembang, apakah dengan berjalan atau menggunakan moda transportasi lain untuk memperoleh fasilitas kesehatan primer. Studi kasus untuk penelitian ini adalah Pemerintah Daerah Ijebu Utara pada Provinsi Ogun-Nigeria yang terdiri dari wilayah kecamatan perkotaan dan pedesaan. Seratus lima puluh rumah tangga dipilih secara acak untuk interview. Regresi logit digunakan untuk mendeskripsikan beberapa variabel prediksi yang digunakan untuk kecenderungan rumah tangga terkait untuk berjalan memperoleh fasilitas kesehatan primer. Model prediksi pada daerah penelitian mampu mengklasifikasikan 80,0% kasus dengan benar. Kondisi ini dengan sederhana menunjukkan bahwa variabel prediksi (variabel independen) berkontribusi terhadap kemampuan prediksi dari model regresi logistik. Nilai pseudo R-squares dari Cox dan Snell R-square serta Nagelkerke R juga menunjukkan bahwa model logistik relevan untuk memprediksi bahwa rumah tangga memilih berjalan atau menggunakan kendaraan untuk memperoleh akses pendidikan. Dalam penelitian ini, status permukiman ( $p=0,00$ ) dan biaya transportasi berkontribusi signifikan terhadap prediksi fasilitas kesehatan ( $p=0,00$ ). Studi ini juga menunjukkan bahwa jumlah rumah tangga pada wilayah perkotaan yang bersedia untuk berjalan memperoleh fasilitas kesehatan adalah 88,1%, lebih rendah dibandingkan rumah tangga pada wilayah pedesaan. Penelitian ini juga menunjukkan bahwa rumah tangga miskin 49% kali lipat bersedia untuk berjalan untuk memperoleh fasilitas kesehatan. Pengetahuan tentang faktor-faktor yang akan menentukan apakah pencari pelayanan kesehatan di daerah berkembang akan ingin berjalan atau tidak akan membantu pemerintah dalam perencanaan dan penyediaan fasilitas kesehatan.

**Kata kunci:** Wilayah berkembang, aksesibilitas, fasilitas kesehatan primer, model logistik, moda transportasi

## I. Introduction

Access to primary health facilities is an important determinant of the overall well being of any given population in an area [Obrist et al., 2007; Kumar, 1999; Guagliardo & Mark, 2004; Bagheri, et al., 2005, Jean-Frederic, et al., 2013]. Lack of access can cause ill health

and ill health perpetuates poverty among the populace. It is in recognition of this importance of health care service that the International Conference on Primary Health Care held at Alma-Ala in 1978 declared that health, which is a state of complete physical, mental and social well being, is a fundamental human right. The separating distance is a major determinant of physical access to primary health facilities especially in rural regions where density of population is often low and settlements are far apart. In rural regions distances to public facilities are usually longer compared to urban areas. However it is often assumed in public facilities location studies in developing regions that users of facilities will walk to them. Thus distance standard in facilities planning are usually based on this assumption of the users walking. For example the government in Ogun state in her policy on health [Ogun State Government, 2010] specify an average distance of five kilometers and maximum distance of ten kilometers to primary health centres. These standards are based on the perceived walking capability of health care service seekers. It has been observed by Rushton [1988] that decision makers don't always research into the basis for setting the ideal distance standard in public facility planning.

Health care service seekers often use different modes of transport to overcome the tyranny of distance by walking or use of commercial or private bicycles and vehicles. In modern times it is often more convenient to use a vehicle than to trek and there is limit to the extent people can walk and are willing to walk to use health facilities. In some developing regions there are areas without mechanized means of transport, there are areas where people cannot afford to pay for transport services

and some settlements are remotely located away from primary health facilities. In such situations, will people be willing to walk to use primary health facilities? Following from the question raised, the focus of this study is to identify the factors that will determine whether health care service seekers in a developing region will want to walk to use primary health facility and the extent to which they are willing to walk. Such knowledge will assist the planners in the planning and provision of primary health care facilities. This research paper has been divided into three sections. Section one is the introduction and it discusses the research problem and objective of the study. Also contained in section one is the discussion of the study area, methodology and a review of the literature on access to public facilities. Section two of the paper discusses the research findings and section three is the conclusion.

The study area is discussed here to provide the background information such as the location and the mode of transport available in the area. The area used for this study is Ijebu North local government area of Ogun State, Nigeria. This local government area has been chosen to represent a typical region in the developing world. It consists of three urban centres and the remaining part is rural. This pattern allowed the implication of rural/urban place of residence of an health care service seeker on his readiness to walk to be examined in this study.

The local government area (LGA) has its headquarter in Ijebu Igbo and it is located at the northern end of Ogun state. The local government area is approximately located between latitude 6055' and 70 N and between longitude 3045; and 4005'E. The total land area of the local government area is about

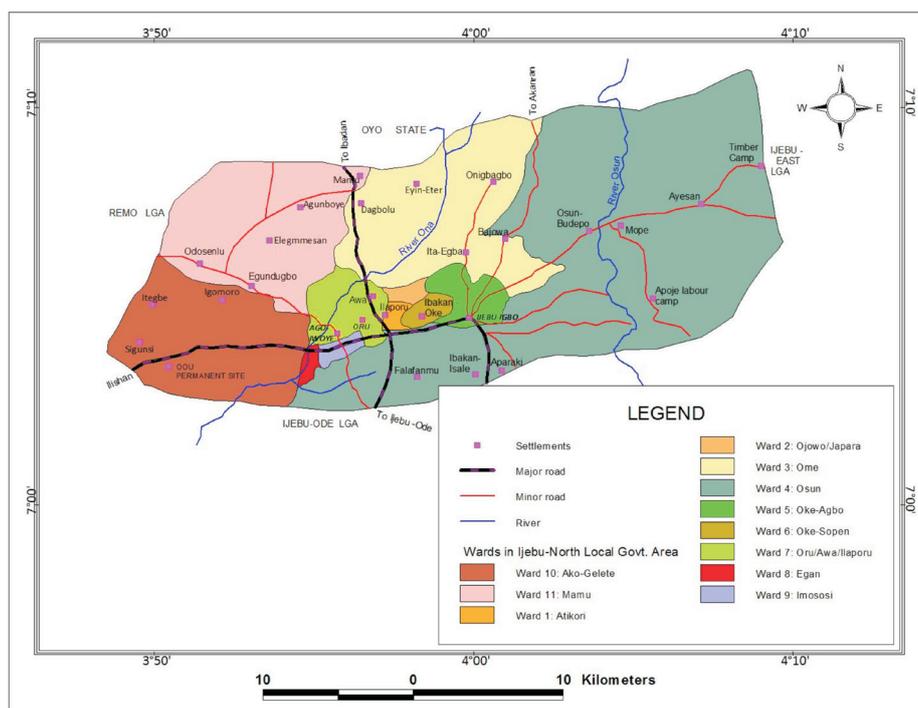


Figure 1. Map of Ijebu North Local Government Area, Ogun State

967 square km. The 2006 population of the LGA is 284,336 (National Population Commission). The LGA is bounded to the north by Lagelu LGA of Oyo state. To the east is Ijebu East and Ijebu Ode LGA is to the south west. To the west of the LGA is Ikenne LGA. Ijebu North LGA is partitioned into eleven (11) wards which are: Atikori, Oke Agbo, Oke Sopen, Omen, Osun, Oru/Awa/Ilaporu, Ojowo/Japara, Egan, Imososi, Mamu and Ako Gelete. As shown in Table 1 some of the wards are urban and others are rural.

The dominant mode of transport available in Ijebu North Local Government Area is road. Roads are used for accessing public facilities in the local government area. Before the penetration of the colonial administrators, bush paths were the forms of transport routes in the study area. Some of these paths were later widen into motorable roads. The motorable roads are classified as either a main road or a secondary road. See Figure 1 for map of the road networks in Ijebu North Local Government Area.

**2. The Methods**

This section discusses the type of data used and all the methods used in this study in collecting and analyzing the collected data to achieve the objective of the study.

The main data for this study is from a household survey in Ijebu North local government area of Ogun state. The sampled households were selected from the eleven wards. Some of these wards form parts of the urban centres in the local government area and others are basically rural. Ward 1, Ward 2, Ward 5 and Ward 6 are the urban areas of Ijebu Igbo. Ward 8 and

Ward 9 are the urban part of Ago-Iwoye and Ward 7 covers Oru, Awa and Ilaporu which are also urban areas. One hundred questionnaires were administered in the urban wards and fifty in the rural wards. The number of questionnaires is for convenience sake given the time and resources available for the study. The questionnaires were administered on house hold basis and in proportion to the projected population of the wards as shown in Table 1. The sampled households were selected at random starting from one end of the ward and moving progressively to the end of the ward.

The characteristics of the interviewed households were presented in frequency tables and the logistic regression model was used to establish the nature of the relationship between the dependent variable and the predictors.

A summary of the literature on physical access to public facilities is presented in this section. Accessibility can be seen as the ease with which the users are able to use the services provided by a public facility. Most studies on access to public facilities often focus on the barriers to the use of facilities. According to Peters et al. [2008] the dimensions of barriers to assessing health services has been categorized into four: (1) geographical accessibility (2) availability, (3) affordability and (4) acceptability.

Public service are form of public goods that the users travel to consume or the service are delivered to the consumer like the case of a fire fighting service [Lea, 1982]. Thus the separating physical distance between a user and the facility serve as a barrier to access the facility. Researches relating to improving physical access to public facilities cover studies on: location and

Table 1. Administration of questionnaires

Ward	Status	1991 actual census*	2013 projected population**	Number of questionnaires
Ward 1 (Atikori)	urban	13432	25117	13
Ward 2 (Ojowo/Japara)	urban	11321	21169	11
Ward 3 (Ome)	rural	1149	2149	5
Ward 4 (Osun)	rural	6524	12199	27
Ward 5 (Oke Agbo)	urban	16626	31089	16
Ward 6 (Oke Sopen)	urban	23538	44014	22
Ward 7 (Oru/Awa/Ilaporu)	urban	11445	21401	11
Ward 8 (Egan)	urban	18795	35145	18
Ward 9 (Imososi)	urban	10833	20257	10
Ward 10 (Ako Gelete)	rural	2132	3987	9
Ward 11 (Mamu)	rural	2172	4061	9
Total				150

Sources:

\* National Population Commission

\*\* Projected population: The Annual growth rate used for 1991 - 2000 was 2.60%. 2.75% was used for 2001 - 2005 while 3.18% was used from 2006 -2013.

allocation modeling of public facilities [Ayeni, 1985; Ayeni, 1986; Rushton, 1988; Owoola, 2002], improving the network links between the users and facilities [Moller-Jensen and Richard Y.K., 2001; Schoeps et al. 2011] and improving the transport cost between the demand points and public facilities. The focus in this study is on the mode of transport used to primary health facilities.

To understand the location of service centres vis-a-vis the demand points the central place theory (CPT) is fundamental. The theory attempts to explain the size, number and spacing of service centres providing services to the surrounding population. The concepts of threshold and range of the CPT are fundamental to explaining location of public facilities. The threshold ensures that there exists a minimum number of people that generate enough demand to keep the service running. Furthermore, there exists a maximum distance above which consumers would find it more profitable to visit a nearer service centre. This maximum distance was called the range [Berry, 1967; Christaller, 1933]. The range is a measure of physical accessibility of a particular service.

The issue of accessibility as a concept is the most commonly articulated and the least understood in locational studies [Ayeni and Rushton, 1986; Owoola, 1996]. On one hand, it is interpreted in terms of physical proximity of locations that permit involvement in activity space of a given region [Burns, 1978; Moseley, 1979; Owens and Shaw, 1972]. Others have interpreted the concept in terms of matching the distribution of services to the distribution of potential population to be served using the logic of central place theory [Gould, 1978; Fisher and Rushton, 1988]. However, of particular interest in this study is physical accessibility defined as the spatial separation of service users from service centres. Thus physical accessibility is the ease of movement to activity locations and it is the distance people are willing to travel to utilize a service [Ikporukpo, 1987; Okafor, 1989; Ayeni, 1989].

Most published measures of spatial accessibility to health have been classified into: distance to nearest provider, average distance to a set providers, and gravitational models of provider influence [Guagliardo, 2004]. Travel distance to nearest provider is typically measured from a patient's residence or from a population centre. Travel distance to nearest provider has been assumed to be a good measure of spatial accessibility for rural areas, where provider choices are very limited and the nearest provider is also the most likely to be used Russell [2008]. Maximal service distance is another measure of travel distance to nearest provider, but here the distance, time or cost of a user most distant from a provider is considered. The maximal service distance is a common measure of articulating location of public facilities in the literature. Another measure of spatial accessibility is the average travel distance to provider [Ulises and Carina, 2012]. For this measure the distance

from any patient or population to all providers within a system is summed and average. Gravity models are a combined indicator of accessibility and availability of health service. Gravity models attempt to represent the potential interaction between any population point and all service points within a reasonable distance, discounting the potential with increasing distance or travel impedance.

The focus of this study is on determinants of mode of transport used by a health care service seeker. Such trips can be made by different modes of travel and the determination of the choice of travel mode by individual is known as modal split [Salter, 1983]. Modal split model therefore is the public transportation version of the general model of human choice that explains how people select between competing alternatives [UCLA, 2011]. Modal split could be analyzed using probabilistic models such as the discriminant analysis, probit and the multinomial or dichotomous logit models [Okoko, 2006]. The focus here is on choice between use of vehicle and walking to health facilities. The discussion above is a summary of some of the researches on physical access and mode of transport to public facilities. Following from this review of the literature, the key findings of this study are presented in the next section.

### 3. Result and Discussion

The findings in this study are discussed in this section. The characteristics of the households interviewed are discussed according to whether they reside in rural or urban area. The result from the application of the regression model framework to determine the willingness of health care seekers to walk or use some other modes of transport is discussed in this section.

The characteristics of the households selected for interview are presented in Table 2. The discussions include the socio-economic and demographic characteristics of the households and the characteristics of transport used while attending primary health care service. The discussion below focused on where the characteristics of households in urban and rural areas is significantly different and this difference has implication on the predictive logistic model in the next section. The households were divided into two categories of those that spend below \$2 a day (categorized as poor) and those that spend above \$2 a day. Table 2 shows that 41% in urban areas and 78% in rural areas spend below \$2 a day. This observed difference and the importance of expenditure on mode of transport used made the variable to be included in the logistic regression model in the next section. This study also revealed that 17% in urban area and 50% of households in rural areas trek to the health facilities they use often. The proportion of people that trek is our dependent variable in the logistic model in the next section. It is shown in Table 2 that 58% of households in urban areas and 38% of households in rural areas

Table 2. Characteristics of the households in the study area

Attributes of Head of Household	Urban		Rural	
	Frequency	Percent	Frequency	Percent
<b>Gender</b>				
Male	49	49.0	22	44.0
Female	51	51.0	28	56.0
<b>Age</b>				
Below 18 years	5	5.0	3	6.0
18 – 40 years	60	60.0	33	66.0
41 – 60 years	32	32.0	13	26.0
Above 60 years	3	3.0	1	2.0
<b>Monthly income</b>				
Less than N18,000	25	25.0	22	44.0
N18,000 – N30,000	33	33.0	12	24.0
N30,001 – N60,000	21	21.0	7	14.0
N60,001 – N100,000	12	12.0	0	0.0
Above N100,000	5	5.0	2	4.0
<b>Amount spend per day</b>				
Below N320	41	41.0	39	78.0
Above N320	59	59.0	11	22.0
<b>Type of health facility often used</b>				
Orthodox	57	57.0	15	30.0
Traditional	13	13.0	15	30.0
Chemist	30	30.0	18	36.0
<b>Ownership of health facility</b>				
Public	52	52.0	24	48.0
Private	39	39.0	17	34.0
<b>Means of transport to health facility often used</b>				
Trek	17	17.0	25	50.0
Cab	20	20.0	2	4.0
Commercial motorcycle/Tricycle	45	45.0	18	36.0
Private motorcycle	10	10.0	4	8.0
Private vehicle	7	7.0	1	2.0
<b>Distance of residence to health facility often use</b>				
Less than 1km	33	33	19	38.0
1 – 2 km	46	46	21	42.0
3 – 4 km	20	20	9	18.0
5 – 6 km	1	1	0	0
<b>Perception of accessibility of health facility</b>				
Very accessible	28	28.0	16	32.0
Accessible	58	58.0	19	38.0
Not accessible	14	14.0	15	30.0
<b>Distance willing to travel to use health facility</b>				
Less than 1 km	21	21.0	9	18.0
Between 1 - 2km	30	30.0	17	34.0
Between 2- 4 km	28	28.0	10	20.0
Between 4 – 7 km	16	16.0	7	14.0
Between 7 – 10 km	5	5.0	10	20.0
Above 10 km	0	0.0	2	4.0

continue Table 2 .....

Reasons for preferring a particular health facility				
Cheapness				
Reliability of services	26	26.0	14	28.0
Safety of the premises	25	25.0	13	26.0
Accessibility	12	12.0	6	12.0
Convenience	1	1.0	0	0.0
Average transport cost from home to health facility				
N20 – N50	26	26.0	14	28.0
N51 – N100	25	25.0	13	26.0
N101 – N150	12	12.0	6	12.0
Above N150	1	1.0	0	0.0
Average movement time from home to health facility				
Less than 5 minutes	20	20	11	22.0
10 mins – 30 mins	55	55	23	46.0
30 mins 1 hour	24	24	16	32.0
Usual problems encountered on your trip to the health facility				
Bad road condition	16	16.0	13	26.0
Long distance covered	15	15.0	9	18.0
Insufficient transport facilities	7	7.0	5	10.0
High transport cost	6	6.0	8	16.0
Poor management of vehicles	1	1.0	3	6.0

perceived the health facilities in their area as accessible. The difference and the importance of perception of users on accessibility made the variable to be included in the logistic regression model.

In this section we are going to examine the factors that determine whether a user of health facility will walk or go by some means of vehicles to the health facility. Thus the dependent variable (i.e. the variable we want to predict) is dichotomous that indicates whether the user will walk or not. Of the independent variables (the predictors), the amount of expenditure per day (to measure level of poverty), the estimated distance to health facility and the user's perception of accessibility of the health facility are of particular interest to the researcher. The binary logistic regression is used to model and examine the relationship above. A logistic regression model allows us to establish a relationship between a binary outcome variable and a group of predictor variables [UCLA, 2011].

Binomial (or binary) logistic regression is a form of regression which is used when the dependent variable is a dichotomy and the independent variables are of any type. The goal is to find the best set of coefficients so that cases that belong to a particular category will, when using the equation have a very high calculated probability that they will be allocated to that category. This enables new cases to be classified with a reasonably high degree of accuracy as well.

The dependent variable and the independent variables used in the logistic regression to model the determinants of the means of transport to health facilities are summarized Table 3 .

Using the variable names above the logistic

regression or prediction equation for this study is specified as:

$$\text{Logit}(p) = \log(p/(1-p)) = \beta_0 + \beta_1 \cdot X_1 + \beta_2 \cdot X_2 + \beta_3 \cdot X_3 + \beta_4 \cdot X_4 + \beta_5 \cdot X_5 + \beta_6 \cdot X_6 + \dots$$

Where p is the probability of success.

X1 = Dailyexp1

X2 = Peraccess1

X3 = Gender1

The remaining variables are continuous

X4 = Income

X5 = Distanctohf

X6 = Transcostohf

X7 = Age - Age of respondent

X8 = Rural/Rural status

Writing the logistic regression equation more formally:

Let y be the binary outcome variable indicating failure/success (in our study it will be walk/don't walk) with 0/1 and p be the probability of y to be 1 and let X1, ..., Xk be a set of predictor variables. The independent variables are either dichotomous or continuous. Where the data are categorical (i.e. options are in categories such as in assessing the perception of access) the options are recoded into a dichotomous variable as "accessible" and "not accessible".

One way to assess our logistic regression model is to examine the number of the cases of the dependent variable it is able to predict correctly. We will then compare the level of accuracy of the prediction of the null model with that of the model containing all our

Table 3. Definition of the data in the regression model

Variable name	Meaning	Options	Reason for choice of variable
	Dependent		
Trek	Means of getting to the health facility	Trek = 1 Vehicle = 0	The variable to be predicted. We want to know why some household trekked to health facilities.
	Independent variables (Categorical)		
Dailyexp	Daily expenditure by respondent	Below \$2 = 1 Above \$2 = 0	To know the effect of daily expenditure on mode of transport to health facilities
Peraccess	Perception of access to health facility	Far = 1 Not far = 0	How perception of access determines whether a household will walk to health facility
Gender	Gender of respondent	Male = 1 Female = 0	How being a male or female affect whether the head of household will walk to health facility or not
Rural_Urb	Rural/urban status of settlement	Rural = 0 Urban = 1	The rural/ urban status of household will determine whether a household will walk to health facility
	Independent variables (Continuous)		
Income	Income of respondent		Income of the household will determine the ability of the household to use a vehicle or not.

Table 4. Classification table for the model

Observed	Predicted			Percent correct
	Use of vehicle 0	Trek 1		
Use of vehicle 0	98	10		90.7
Trek 1	20	22		52.4
Overall percentage				80.0

Computed by author with SPSS 15, 2011

Table 5. Omnibus test: the model chi square

Chi-square	Degrees of freedom	Significance
53.413	8	0.00

Source: Computed by author with SPSS 15, 2011

Table 6. Model summary

Cox & Snell R square	Nagelkerke R square
0.295	0.425

independent variables (full model) as shown in Table 4. The table shows that 98 cases are observed to be 0 and are correctly predicted to be 0. Also cases that are not correctly predicted are shown. Ten cases are observed to be 0 but are predicted to be 1. The overall percent of cases that are correctly predicted by the model is 80%. Table 4 above shows that our model was able to classify 80.0% of the cases correctly.

The model chi square shown in Table 5 is used to test the overall significance of the predictors in the regression model. The analysis here involves comparing the null model (a model containing only the constant) and the full model (a model with all the predictor variables). Table 5 shows a chi-square value of 53.413 and a probability of  $p < 0.00$ . Thus the indication is that the null model has a poor fit and that the predictors have significant effect in predicting the dependent variable.

There is no close analogous statistic in logistic regression to the coefficient of determination ( $R^2$ ). The model summary in Table 6 below provides some approximations to  $R^2$  and they are called pseudo  $R$ -squares. Cox and Snell's  $R$ -square attempts to imitate  $R$  square based on "likelihood", but its maximum can be (and usually is) less than 1.0 making it difficult to interpret. Here it is indicating that 29.5% of the variation in the dependent variable is explained by the logistic model.

The Nagelkerke modification that does range from 0 to 1 is a more reliable measure of the relationship. Nagelkerke's  $R$  will normally be higher than the Cox and Snell measure. In our case it is 0.425 indicating a moderate relationship of 42% between the predictors and the prediction.

The logistic coefficients ( $\beta$ ) are the values for the logistic regression equation for predicting the dependent variable from the independent variable. They are in log-odds units. The coefficients tell you about the relationship between the independent variables and the dependent variable, where the dependent variable is on the logit scale. These estimates tell the amount of increase (or decrease, if the sign of the coefficient is negative) in the predicted log odds of success that

would be predicted by one unit increase (or decrease) in the predictor, holding all other predictors constant. For the independent variables which are not significant, the coefficients are not significantly different from 0, which should be taken into account when interpreting the coefficients. In table 7 the Wald statistic and associated probabilities provide an index of the significance of each predictor/coefficient in the equation. The Wald statistic has a chi-square distribution. The simplest way to assess Wald is to take the significance value and if less than 0.05 reject the null hypothesis as the variable does make a significant contribution.

The logistic regression coefficients are in log-odds units and difficult to interpret, so they are often converted into odds ratios. This has been done in table 7 by exponentiating the coefficient ( $e^B$ ), and the result are displayed at the right-most column of the table and labeled "Exp(B)".

In our study, we note that the independent or predictor variables: Rural\_Urb (Settlement status) ( $p=0.00$ ) and Transcostohf (transport cost to health facility) ( $p=0.00$ ) contributed significantly to the prediction of the regression model. The researcher may want to drop independent variables from the model when their effect is not significant by the Wald statistic. The Exp(B) column in Table 7 presents the extent to which raising the corresponding predictor variable by one unit influences the odds ratio. We can interpret  $\text{Exp}(\beta)$  in terms of the change in odds. If the value exceeds 1 then the odds of an outcome occurring increase; if the figure is less than 1, any increase in the predictor leads to a drop in the odds of the outcome occurring.

Households in urban areas are not likely to walk to health facility compared to households in rural areas. For a household in an urban area (coded as 1), we expect to see 0.119 decrease ( $\beta = -2.125$ ) in the odds of trekking (the chances of trekking decreases) to the health facility often used. Another way to express this in terms of percent change is to say that the odds for a household in an urban area to walk to the health facility often used is (100% - 11.9%) 88.1% lower than the

Table 7. Variables in the equation

Variable	B	S.E	Wald	df	Sig.	Exp(B)
Dailyexpen	0.712	0.560	1.617	1	0.203	1.491
Percaccess	1.458	0.836	3.039	1	0.81	4.297
Gender	-0.435	0.455	0.915	1	0.339	0.647
Income	-0.026	00.014	3.801	1	0.051	0.974
Distancohf	0.103	0.388	0.07	1	0.791	1.108
Transcostohf	0.039	0.09	17.219	1	0.000	1.962
Age	0.008	0.019	0.155	1	0.694	1.008
Rural_Urb	-2.125	0.546	15.144	1	0.00	0.119
Constant	2.339	1.061	4.860	1	0.27	10.371

odds for a household in a rural area.

From Table 7 we can derive that for every one unit increase in transport cost, we expect to see 1.962 increase in the odds of trekking ( $\beta = 0.039$ ) to the health facility frequently used. If the transport cost should increase by one unit, holding all other predictors constant, the odds for a household to trek to the health facility frequently used will increase by 96%.

Apart from settlement status and transport cost that are significant in explaining the outcome of the logit model we also examined the relationship between the amount expended daily by the households and the mode of getting to the health facility (trek/ non trek. The  $\text{Exp}(\beta)$  value associated with Dailyexp (Daily expenditure) is 1.491 ( $\beta = 0.712$ ). Hence when

Dailyexp is raised by one unit (i.e. if the respondent is poor - spends less than \$2 or N320 a day and coded as 1) the odds ratio is 1.491 times as large and therefore households that are poor are 49% times more likely to walk to the health facility they frequently used.

We derive the following predicting logistic regression equation from the parameters in table 7.

$$\text{Log}(p/1-p) = 2.339 + 0.712\text{Dailyexpen} + 1.458\text{Percaccess} - 0.435\text{Gender} - 0.026\text{Income} + 0.103\text{Distancohf} + 0.039\text{Transcostohf} + 0.008\text{Age} - 2.125\text{Rural\_Urb}$$

Where p is the probability of walking to the health centre. The equation above can be used to predict the outcome for a particular case.

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