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Addis Ababa University
College of Natural and Computational Sciences
Department of Statistics

Factors associated with the number of antenatal care visits in four disadvantaged regions of Ethiopia

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A thesis submitted to the Department of statistics in partial fulfillment of the requirements for the degree of Master of Science in Statistics

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This is to certify that the thesis prepared by Tilahun Abate, entitled: Factors associated with the number of antenatal care visits in four disadvantaged regions of Ethiopia and submitted in partial fulfillment of the requirements for the Degree Master of Science in Statistics (Biostatistics) complies with the regulations of the University and meets the accepted standards with respect to originality and quality.

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Factors associated with the number of antenatal care visits in four disadvantaged regions of Ethiopia: Application of Zero Inflated Negative Binomial model

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A Thesis Submitted to Addis Ababa University Department of Statistics in Partial Fulfillment of the Requirement for Degree of Masters of Science in Statistics (Biostatistics)

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Abstract

The coverage and the quality of antenatal care play a significant role in maternal health. Antenatal care is crucial to managing preventive measures and identifying potential complications which will occur at birth. The objective of the study is to examine the determinant factors of the socioeconomic and demographic characteristics on the frequency of antenatal care attendances in Somalia, Afar, Gambela and Benishangul Gumuz regions of Ethiopia. The 2016 Ethiopian Demographic and Health Survey (EDHS), which is obtained from Central Statistical Agency (CSA) were used for this study. A total of 2386 child-bearing mothers were selected. Among the child-bearing mothers who received ANC from skilled providers about 1116 (47.8%) of the child-bearing mothers had not received ANC service during their pregnancy period, while 687 (28.8%) of the child-bearing mothers had received one up to three times ANC service. In total, 583 (24.33%) of childbearing mothers had received 4 and more times ANC service, whereas 1250 (52.2%) had visited at least one time. Four models Poisson, negative binomial, zero inflated Poisson and zero-inflated negative binomial regression models were fitted to the data. Zero Inflated Negative Binomial Regression model was found to be relatively better to fit the data. Through the analysis, Husband's education, literacy, decision maker for healthcare, and media exposure like listening radio, watching television and , Wealth index, and age of women were found to be important factors significantly associated with the use of ANC visits. The average number of ANC visit (2.08) in the four economically disadvantaged regions did not meet the recommended number (at least four) of ANC visits by WHO. From the descriptive statistical result ,it seems that, on average Gambelaregion child-bearing mothers had higher number of ANC visits than Afar, Somali and Benishangul Gumuz regions. There is an indication that there were differences observed among the four regions on the average number of antenatal care visits. Thus we recommend that special needs in promoting ANC across regions should be prioritized and structured for better implementations in all respective regions.

Abbreviations

AIC Akai Information Criterion

ANC Antenatal Care

AOR Adjusted Odds Ratio

BIC Bayesian Information Criterion

CSA Central Statistical Agency Demographic and Health Survey

DF Degrees of freedom

HIV Human Immunodeficiency Virus

IRR Incidence Rate Ratio

MDG Millennium Development Goal

NB Negative Binomial

PMTCT Prevention of Mother to Child Transmission

PNC Postnatal care

PRM Poisson Regression Model

SDG Sustainable Development Goal

UNICEF United Nations International Children's Emergency Fund

WHO World Health Organization

ZIP Zero –Inflated Poisson

ZINB) Zero –Inflated Negative Binomial

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CHAPTER ONE

INTRODUCTION

1.1 Background of the Study

Antenatal care (ANC) is a care provided by skilled health personnel to a pregnant woman throughout her pregnancy (WHO2016). The primary aim of ANC is to promote and protect the health of women and their unborn babies during pregnancy so as to achieve at the end of a pregnancy a healthy mother and a healthy baby. Currently, it is regarded as a basic component of maternal health care on which the life of mothers and babies depends(CSA 2011).

It is a key strategy for reducing maternal morbidity and mortality directly through detection and treatment of pregnancy-related illness or indirectly through detection of women at risk of complications of delivery and ensuring that they deliver in a suitably equipped facility (TrinT.,2007)

Developing countries have low ANC coverage and there is a wide gap of ANC between mothers in developed and developing countries. In developed countries, 97% of women make at least 1 ANC visit. In developing countries, coverage of at least 1 ANC visit is low, that is, 69% in sub-Saharan Africa (WHO,2018). Complications of pregnancy and childbirth are the leading causes of disability and death among mothers in the reproductive age (15-49) groups in developing countries (Making Pregnancy Safer, 2009) .Ethiopia is also one of the countries with the highest maternal mortality ratios in the world (CSA,2011).

In Ethiopia, an estimated 2.9 million women give birth every year. Of these, approximately 25,000 women and girls die each year and more than 500,000 suffer from complications including obstetric fistula (Fekadu M., (2014)).

Health care services during pregnancy and after delivery are important for the survival and

wellbeing of both the mother and the infant. Skilled care during pregnancy, childbirth, and the postpartum period are important interventions in reducing maternal and neonatal morbidity and mortality. As highlighted in the 2015-16 Health Sector Transformation Plan (HSTP), maternal and newborn health are priorities for the Government of Ethiopia (MOH, 2015). The HSTP key components are delivery at a health facility, with skilled medical attention and hygienic conditions; reduction in complications and infections during labour and delivery; timely postnatal care that treats complications from delivery; and education of the mother on care for herself and her infant. The goal of the reproductive health program is to reduce the maternal mortality ratio to 199 maternal deaths per 100,000 live births and the neonatal mortality rate to 10 per 1,000 live births by 2020.).(CSA,2016)

The high number of maternal deaths in some areas of the world reflects inequities in access to health services and highlights the gap between rich and poor. Almost all maternal deaths (99%) occur in developing countries. More than half of these deaths occur in sub-Saharan Africa and almost one-third occur in South Asia. More than half of the maternal deaths occur in fragile and humanitarian settings. The maternal mortality ratio in developing countries in 2015 is 239 per 100, 000 live births versus 12 per 100, 000 live births in developed countries. There are large disparities between countries, but also within countries, and between those women living in rural versus urban areas(WHO,2015).

The quality of ANC is dependent on the qualifications of health providers and the number and frequency of ANC visits. Adverse pregnancy outcomes can be minimized or avoided altogether if ANC is received early in the pregnancy and continued through delivery. The World Health Organization (WHO) and United Nations Children's Fund recommend that a woman without complications should have at least 4 antenatal visits, the first of which should take place during the first trimester. Antenatal care represents a significant opportunity to reach a large number of pregnant women (Mrishoetal., 2009).

Reducing maternal health complication is directly or indirectly assured if and only if the child-bearing mother takes ANC service or visited doctor as well as health-care providers. Furthermore, the prevalence of ANC service utilization of child-bearing women is absolutely low in developing countries. Regular antenatal visits are necessary to establish confidence

between the woman and her health-care provider, to individualize health promotion messages, and to identify and manage any maternal complications or risk factors associated with child-bearing mothers((Abyot N.et.al (2019)).

1.2 Statement of the Problem

Maternal mortality is due to different causes and problems. These problems and causes are not specifically characterized. Thus to reduce the maternal mortality a wide range of interventions are required. The coverage and the quality of antenatal care play a significant role in maternal health. Antenatal care is crucial to managing preventive measures and identifying potential complications which will occur at birth.

There are many studies on the field of maternal care in which researchers often examined the effects of covariates on patients using logistic regression models, but this model cannot account for the number of ANC a mother visited. Hence, such data can be more explored using Poisson, negative binomial, zero inflated, and hurdle negative binomial models. Because of counting, common data analysis procedures can be adopted directly. With recognizing the underlying factors associated with the number of antenatal care visit utilization of child bearing mothers in Four disadvantaged regions (Somalia, Afar, Gambela and BenishangulGumuz) were attempted to study.This study focused on these regions because of the fact that there are competing realities of life in Ethiopia, where a family's circumstances can largely be determined by which region they live in, or whether they are in a rural or urban setting.In Somali , Afar ,Gambella and Benishangul regions, many families depend on the seasonal movement of livestock and are subject to recurrent droughts. Most children live more than an hour away from the nearest school, a trip to the market may take the better part of a day, and access to basic services is limited(Berhe M.et.al(2020).

In turn with the above veracity, the study attempted to turn up with possible clarification and commendation after having clear thought upon the state of affairs by giving own emphasis to answer the following research questions: What is the frequency and distribution of ANC visit

utilization of child-bearing mothers in study areas? What are the determinants significantly influencing ANC visits?

1.3 Significance of the Study

Promoting the antenatal care service definitely declines the number of maternal deaths. The evaluation of the interdependence of socioeconomic and demographic characteristics with utilizing any ANC is also measuring interrelationship among those characteristics and the maternal health. Therefore; making an assessment on the contemporary situations of the utilization of antenatal care and its symbiotic factors is vital and has a reflex action to measure possession of maternal health. This study will therefore, help the stakeholders to reduce maternal and infant mortality and clarifying the main determinant factors that significantly affect the number of ANC visit utilization of child-bearing mothers. In general, the results of this study will give information to concerned bodies in setting policies, strategies, and further investigation for ANC visit utilization

1.4 limitations of the study

The limitations of the current study were cost and time constraints to study on primary data sources. And thus, information on the distance to the nearby health facility that might have impact on the utilization of ANC was not included in the survey.

1.5 Objectives of the Study

1.5.1 General Objective

The general objective of the study is to examine the association of the socioeconomic and demographic characteristics on the frequency of antenatal care attendances in Somalia, Afar, Gambela and BenishangulGumuz of Ethiopia.

1.5.2 Specific Objectives

- To estimate the average coverage of ANC in Somalia, Afar, Gambela and BenishangulGumuz regions of Ethiopia.

CHAPTER TWO

LITRATURE RIVIEW

A cross-sectional study in Nepal conducted by Neupane et al. (2020) on Andersen's model on determining the factors associated with antenatal care services showed a strong association between both contextual and individual determinants and frequency of ANC visits. Contextual factors such as province, household wealth index, ethnicity, and individual level predisposing factors such as age along with enabling factors (partner's education level, exposure to multimedia) contributed to the recommended number of ANC visits. Thus, concluded that ANC attendance is vital for live birth, and intervention target at various levels is recommended. Effective National Safe Motherhood Policy meets SDGs' target by addressing healthy reproductive needs. To make National Safe Motherhood policy effective, the factors associated with 4+ ANC visit needs to be addressed appropriately with developing special health promotion program with a focus on the vulnerable and disadvantaged. Moreover, the flexible use of multimedia should be encouraged to improve maternal health literacy(Neupane et al. (2020)).

A study conducted by Teklesilasie and Deressa (2018) used logistic regression analyses with odds ratio and 95% confidence interval were computed to test homogeneity of the two groups' baseline characteristics and examined the association between husbands' involvement in antenatal care and women's use of skilled attendants during birth. The results of their study revealed that women who reported at least one antenatal care visit in which their husbands accompanied them were 6.27 times (95% Confidence interval: 4.2, 9.3) more likely to use skilled birth attendants compared to women attended antenatal care alone(Teklesilasie W. and Deressa W. (2018))

According to the Ethiopian Demographic and Health Surveys report, the proportion of women age 15-49 who received any ANC from a skilled provider has increased from 27% in 2000, to 28% in 2005, 34% in 2011 and 62% in 2016. The proportion of women who received the recommended four or more ANC visit increased from 10% 2000 in to 32% in 2016 (CSA, 2016).

A study conducted by Tarekegn et al (2014) used logistic regression and the result of the study showed that 34% of women had ANC visit 11.7% used skilled delivery attendance and 9.7% of

women had postnatal health checkup. Education of women, ethnicity, parity, place of residence, women's autonomy and household wealth had significant association with the use of maternal health services. Women who completed higher education were more likely to use ANC (AOR=3.8, 95% CI:1.8,7.8); skilled delivery attendance (AOR=3.4, 95% CI:1.9,6.20) and PNC (AOR=2.3, 95% CI: 2.0,5.2). Women from the urban areas use ANC (AOR = 2.3, 95% CI: 1.9-2.9), skilled delivery attendants (AOR = 4.9, 95% CI:3.8,6.3) and PNC service (AOR = 2.6, 95% CI:2.0,3.4) more than women from rural areas. Women who have had ANC visits during the index pregnancy were more to subsequently use skilled delivery attendants (AOR=1.3, 95% CI=1.1-1.7) and PNC (AOR=3.4, 95% CI:2.8-4.1). Utilization of ANC, delivery and PNC services is more among autonomous women than others. And therefore, maternal health in Ethiopia is very low. Socio-demographic and accessibility related factors are major determinants of service utilization. There is a high inequality in service utilization among women with difference in education, household wealth, autonomy and residence. ANC is an important entry point for subsequent use of delivery and PNC service (Tarekegn, et al (2014)).

A study conducted by Zegeye et al (2018) used multivariable logistic regression model to identify significant factors associated with prevention of mother to child transmission service utilization. Poisson and negative binomial regression models were applied, considering the number of ANC visits as a dependent variable. The explanatory variables were age; educational status; type of occupation; decision making power in the household; living in proximity to educated people; a neighborhood with good welfare services; location (urban high-HIV prevalence and rural low-HIV prevalence); transportation accessibility; walking distance (in minutes); and household income status. Significant results were reported at p-values of <0.05 and <0.001. And the result of the study showed that, long walking distance, low household income and living in a rural setting are the significant factors associated with low ANC service utilization. The primary strategies for a holistic policy to improve ANC/PMTCT utilization should thus include improving the geographical accessibility of ANC/PMTCT services, expanding household welfare and paying more attention to remote rural areas (Zegeye, et al (2018)).

A study conducted by Birhan and Seretew (2020) used multivariate regression reanalysis to identify factors that contribute for the change in antenatal care visit, and found out that, among the reproductive age women the rate of an acceptable antenatal care visits was increased from 16% in 2005 to 35% in 2016 in Ethiopia. In the multivariate analysis, about 29% of the increase

in acceptable antenatal care visit was due to a difference in composition of women (endowments) across the surveys. Residence, religion, husband educational attainment, and wealth status was the main source of compositional change factors for the improvements of an acceptable antenatal care visit. Almost two-thirds of an overall change in acceptable antenatal care visit was due to the difference in coefficients/ change in behavior of the population. Religion, educational attainment (both women and husband), and residence are significantly contributed to the change in full antenatal care visit in Ethiopia over the last decades (Birhan and Seretew (2020)).

A study conducted by Yeneneh et al. (2018) used Bernoulli by applying Kulldorff methods using the SaTScan™ software to analyze the purely spatial clusters of ANC utilization and found out that spatial variation was found in ANC utilization at regional levels. Of a total of 7909 women interviewed, 3361 (42.5%) women utilized ANC by skilled health care providers. The highest utilization of ANC was spatially clustered in Addis Ababa. On the other hand, Somali Region was the lowest in ANC utilization. Of a total of 7304 women interviewed in 2005, 2016 (27.6%) utilized ANC. Of a total of 7967 women interviewed in 2011, 2127 (26.7%) utilized ANC. Out of these, the highest ANC utilization was in Addis Ababa, Dire Dawa, Hareri and Gambela sequentially, while Amhara and Somali Region showed the lowest ANC utilization. And also used multiple logistic regression analysis to identify predictors affecting ANC utilization and found out that, of the total respondents, in 2000, 7061 (88.6%), in 2005, 6673 (91.3%), and in 2011, 6720 (85%) were rural residents. The majority of women ages during birth were between 25 to 29 years old. Women who had 2 to 4 birth order were 3259 (43.8%) in 2000, 3026 (41.4%) in 2005, and 3464 (40.9%) in 2011. In 2000, 6539 (82.0%), in 2005, 5734 (78.5%), and in 2011, 5270 (66.6%) women were uneducated. Of the total women in 2005, 1520 (20.8%) and in 2011 1739 (22.0%) were in the lowest wealth quintile. Despite the low coverage of ANC utilization, all regions registered an increasing trend from 2000 to 2011 (Yeneneh et al. (2018)).

A study in Kenya, conducted by Afulani et al. (2019) assuming the response to the individual items to generate ANC service provision and experience of care scores, used both linear and logistic regression to examine predictors and found out that the average service provision score was 10.9 (SD=2.4) out of a total of 16. Most women received some recommended services once, but not at the frequency recommended by the Kenyan Ministry of Health. About 90% had their blood pressure measured, and 78% had a urine test, but only 58 and 14% reported blood pressure monitoring and urine test, respectively, at every visit. Only 16% received an ultrasound at any

time during ANC. The average experience score is 27.3 (SD=8.2) out of a total score of 42, with key gaps demonstrated in communication. About half of women were not educated on pregnancy complications. Also, about one-third did not often understand the purposes of tests and medicines received and did not feel able to ask questions to the health care provider. In multivariate analysis, women who were literate, employed, and who received all their ANC in a health center had higher experiences scores than women who were illiterate (coefficient=1.52, CI:0.26,2.79), unemployed (coefficient=2.73, CI:1.46,4.00), and received some ANC from a hospital (coefficient=1.99, CI: 0.84, 3.14) respectively. The wealthiest women had two times higher odds of receiving an ultrasound than the poorest women (OR=2.00, CI: 1.20, 3.33) (Afulaniet al.2019).

CHAPTER THREE

DATA AND METHEODOLOGY

3.1 DATA SOURCE

The source of data for this study was the 2016 Ethiopian Demographic and Health Survey (EDHS), which is obtained from Central Statistical Agency (CSA). It is the fourth survey conducted in Ethiopia as part of the worldwide Demographic and Health Surveys project.

The 2016 Ethiopian Demographic and Health Survey, was designed to provide estimates for the health and demographic variables of interest for the following domains. Ethiopia as a whole; urban and rural areas (each as a separate domain); and 11 geographic administrative regions (9 regions and 2 city administrations), namely: Tigray, Affar, Amhara, Oromiya, Somali, Benishangul-Gumuz, Southern Nations, Nationalities and Peoples (SNNP), Gambela and Harari regional states and two city administrations, that is, Addis Ababa and Dire Dawa. The principal objective of the 2016 EDHS was to provide current and reliable data on fertility and family planning behavior, child mortality, adult and maternal mortality, children's nutritional status, use of maternal and child health services, knowledge of HIV/AIDS, and prevalence of HIV/AIDS and anemia. We used IBM SPSS Statistics 24 and Stata14 statistical software packages for data analysis.

3.2 Variables in the study

The response (dependant) and explanatory variables included in the survey are defined as follows:

Response variable

The primary outcome variable of this study is the number of any ANC visits of pregnant women in the five years preceding EDHS 2016.

Explanatory variables

The explanatory variables that would be included are explained as follow. The choice of these variables is guided by different literatures as the determinant factors of antenatal care service. These categories of the independent variables were coded to make it appropriate for further analysis and interpretation using different statistical models.

3.2.1 Coding and Description of Variables

Tab 3.1 Coding and Descriptions of Variables

No	Variable	Description	Type	Coding
1	Number of ANC visit	the number of any ANC visits of a pregnant women in the five years preceding EDHS 2016)	Count	
2	Age in 5 year group of respondent (Age)	This variable refers to age group of mother's at the time of the survey	Categorical	1="15-19",2="20-24",3="25-29", 4="30-34",5="35-39",6="40-44" and 7="45-49"
3	Educational status of respondent	This variable refers to the highest educational level mother's attained	Categorical	0 =" No Education", 1 =" Primary", 2 ="Secondary" & 3="Higher"
4	Educational level of husband or Partner	Similar to educational level of the mothers	Categorical	0 =" No Education", 1 =" Primary", 2 ="Secondary"&3="Higher"
5	Mass Media Exposure	Refers that the frequency of listening to radio, watching TV & reading Newspaper	Categorical	0 ="Not at all",1 =" less than once a week", 2 ="at least once a week" and 3="almost every day"
6	Work status	In the survey, this was defined as if mother's has been working in any field other than household work in the seven days before the survey	Categorical	0 = "no", 1 ="yes"
7	Place of residence	This is a dichotomous variable (urban and rural) according to where the woman was living at the time of the survey	Categorical	1= "Urban" and 2="rural"
8	Religion	This variable refers to Religion of mother's at the time of the survey	Categorical	1="Orthodox",2="Catholic",3 ="Protestant", 4="Muslim",5="traditional", 6="Others"
9	Wealth index	Composite score of several indicators of household possession. This Measured by a	Categorical	1=" Poorest",2="poorer",3 =" Middle" , 4=" Richer "and 5 =" richest"

		was based on the questions about whether the household has items and facilities as piped water, toilet, type of floor used, electricity, radio, television and bicycle. Then according to the answer, each asset was given weight. Each household then was assigned a score according to each asset and the scores were summed for each household		
10	Sex of household	This is classified as male or female. Based on the answer from the usual residents of the households on who the head of the household is	Categorical	0 = "Female" and 1 = "Male"
11	Region	Refers to Region where the woman was living at the time of the survey	Categorical	1 = "Afar", 2 = "Somalia", 3 = "BenishangulG", 4 = "Gambela"
12	Current marital status	Refers to the marital status of women at the time of the survey.		
13	Literacy	Refers to the capability of the respondent to read at the time of the survey.	Categorical	0 = "Cannot read at all", 1 = "able to read only part of sentence", 2 = "able to read the whole sentence", 3 = "no card with required language", 4 = "Blind/visually impaired"
14	Person who decides on respondents health care	Refers to who usually is empowered to decide on respondents health care at the time of the survey	Categorical	1 = "Respondent alone", 2 = "Respondent and husband/partner", 3 = "Respondent and other persons", 4 = "Husband/partner alone", 5 = "Someone else", 6 = "Others"

3.3 METHEODOLOGY OF ANALYSIS

3.3.1 Count regression Models

3.3.1.1 Poisson Regression Model

Poisson regression model is the usual starting point to analyze count data. This model provides a standard framework for the analysis of count data. It can be used to model the number of occurrences of an event of interest.

The Poisson regression assumes that the observed counts are generated from a Poisson distribution which is given by:

$$P(Y = y | \lambda) = \frac{e^{-\lambda} \lambda^y}{y!}; y = 0, 1, 2, \dots \quad (3.1)$$

where, Y is a random variable and the values y's are the possible outcomes or the realizations of the random variable Y. Within a specified time interval, the Poisson response variables defined in $N = 0, 1, 2, \dots$ represents number of events occurred. The Poisson distribution is characterized with the property that the mean and the variance are assumed to be equal to λ . That is

$$E(Y) = V ar (Y) = \lambda \quad (3.2)$$

The possible value of λ is clearly a positive number but not necessarily an integer. This characteristic of the distribution is described as "equi-dispersed".

The Poisson regression model is derived from the Poisson distribution by parameterizing the relation between the mean parameter λ_i and covariates (regressors) x. And thus it follows that,

$$\lambda_i = \exp(X_i \beta) \quad (3.3)$$

$$= \exp(\beta_0 + \beta_1 X_1 + \dots + \beta_p X_p)$$

Due to the canonical log-link functions property described above the Poisson regression models are also called Poisson log-link or simply log-linear models, Poisson distribution is belongs to the exponential classes of probability distribution functions. Thus the generalized linear model approach can be used to relate the Poisson response to the explanatory variables. Given p explanatory variables and taking the natural logarithms, we have the following:

$$\ln\lambda_i = \beta_0 + \beta_1 X_{i1} + \beta_2 X_{i2} \dots + \beta_p X_{ip} \quad (3.4)$$

Equality of the mean and variance is referred to as the equi-dispersion property of the Poisson. In real life situations this property is frequently violated. Thus there is an over dispersion or otherwise under dispersion in the data. Over dispersion is thus due to the variance exceeding the mean of the response variable. Over-dispersion is closely related to the presence of unobserved inter-individual heterogeneity, but it can also arise from occurrence dependence between events. Zero event counts are often dominant, leading to a skewed distribution. Also, there may be a great deal of unobserved heterogeneity in the individual experiences of the event in question. Unobserved heterogeneity leads to over-dispersion; that is, the actual variance of the process exceeds the Poisson variance even after regressors are introduced. In these cases alternative models such as Negative Binomial, Zero-inflated Poisson model and Zero-inflated Negative Binomial models could be used to solve over dispersion. In this study, all these possible models will be considered and the model with the best fit will finally be selected.

3.3.1.2 Negative Binomial Model

Under the PRM the observed heterogeneity among the sample members is specified by the rate λ_i as a function of the observed X_i 's. see Equation 3.3.above. In practice the PRM is inadequate to fit due to over dispersion. The PRM underestimates the amount of dispersions in the observed data. Hence the negative binomial regression model solves this failure by using the unobserved heterogeneity parameter α .

For count response variable the conditional negative binomial distribution is given by the following equation.

$$P(Y/X) = \frac{\Gamma(y+\alpha^{-1})}{y!\Gamma(\alpha^{-1})} \left(\frac{\alpha^{-1}}{\alpha^{-1}+\mu}\right)^{\alpha^{-1}} \left(\frac{\mu}{\alpha^{-1}+\mu}\right)^y \quad (3.5)$$

3.3.1.3 Zero-Inflated Models

The main motivation for zero-inflated count models is that real-life data frequently display over dispersion and excess zeros than expected. Zero-inflated count models provide a way of modeling excess zeros in addition to allowing for over dispersion. For observation i , process 1 is chosen with probability ω and process 2 with probability $1-\omega$. Process 1 generates only zero counts, whereas process 2, $g(y|x)$, generates counts from either a Poisson or an NB. In general:

$$Y = \begin{cases} 0, & \text{with probability } \omega \\ g(y|x), & \text{with probability } 1 - \omega \end{cases} \quad (3.6)$$

The probability distribution of a zero inflated Poisson random variable Y is given by

$$P(Y=y) = \begin{cases} \omega + (1 - \omega)e^{-\mu} & \text{for } y=0 \\ (1 - \omega) \frac{\mu^y e^{-\mu}}{y!} & \text{For } y= 1,2,3,\dots \end{cases} \quad (3.7)$$

3.3.2 Handling Over dispersions and Zero -Inflation

3.3.2.1 Zero Inflated Poisson Regression Model

The zero-inflated Poisson (ZIP) regression model is a modification of the standard Poisson regression model and it allows for an overabundance of zero counts in the data. The assumption of the Poisson distribution which is the equality of the mean and the variance is violated because data are often over dispersed when there are excess zeros than expected by the standard Poisson distribution. Thus the Poisson distribution underestimates the dispersion of the observed counts. The over dispersion occurs when the single parameter λ of Poisson distribution is unable to fully describe event counts. Hence this Poisson regression model in this regard is inadequate due to excess zeroes. Thus, Zero-inflated Poisson regression model can be taken for handling the occurrences of a higher proportion of zeroes than the parent Poisson distribution. Lambert (Lambert, D. (1992)).

The ZIP model assumes that some zeroes occur by a Poisson process and others were not even eligible to have the event occur. Some explanatory variable may have a significant effect to produce zero counts to the response variable. But it is not possible to identify which zero count is due to either of the processes. The essential idea is that the data come from two regimes. In the first process the outcome is always a zero count, while in the second the counts follow a standard Poisson process (David Giles(2010)). Thus, the two processes of the ZIP models for generating its count data can be described by the following probability distribution.

$$P(y_i = r) = \begin{cases} \omega_i + (1 - \omega_i)e^{-\lambda_i}, & \text{where } r = 0 \\ (1 - \omega_i)\frac{e^{-\lambda_i}\lambda_i^r}{r!}, & \text{where } r = 1, 2, 3, \dots \end{cases} \quad (3.8)$$

The mean and the variance of the ZIP distribution are, respectively:

$$E_{ZIP}(y_i) = (1 - \omega_i)\lambda_i$$

and

$$\text{Var}_{ZIP}(y_i) = (1 - \omega_i) + \frac{\omega_i}{1 + \omega_i} [(1 - \omega_i)\lambda_i]^2. \quad (3.9)$$

The variance of the ZIP distribution from the above Equation 3.13 can be equivalently expressed as:

$$\text{Var}_{ZIP}(y_i) = (1 - \omega_i)(\lambda_i + \omega_i\lambda_i^2). \quad (3.10)$$

Hence the variance of the ZIPM is the quadratic function of its mean (Minami et.al (2006)) because of the presence of overdispersion due to the excess zeros in the data yields;

$$\text{Var}_{ZIP}(y_i) = (1 - \omega_i)(\lambda_i + \omega_i\lambda_i^2) > \lambda_i(1 - \omega_i) = E_{ZIP}(y_i). \quad (3.11)$$

ZIP model fits simultaneously two separate regression models. They are the count model and the logistic or probit model. These separate models have their own coefficient estimates using the same or different predictor variables. These coefficients/parameters are β and γ respectively. The parameter vector β is estimated due the measured explanatory variables, and vector γ is parameters for covariates due to the logit or the probit of the probability/proportion of the occurrences of zeroes counts ω_i . The predictor variables have different effects on the two separate processes. The nature of the two parameters β and λ can be explained in the following way. For the covariates B and Z, the Poisson mean $\lambda = (\lambda_1, \lambda_2, \dots, \lambda_k)$,

$$\log(\lambda) = B\beta, \quad (3.12)$$

and for the probability $\omega = (\omega_1, \omega_2, \dots, \omega_k)$

$$\log(\omega_i) = \log(\omega_i (1 - \omega_i)) = Z\gamma; \quad (3.13)$$

where $0 \leq \omega_i \leq 1$. In the Equation 3.12 the $\log \lambda$ is the cononical link for a Poisson regression model, and in the Equation 3.13 $\log(\omega_i)$ is the canonical link for Bernoulli probabilities.

3.3.2.2 Zero-Inflated Negative Binomial Regression Model

For count data that are skewed with highly right tail, the NB distribution is appropriate instead of the Poisson distribution. Furthermore the zero-inflated negative binomial (ZINB) regression appropriate to model over dispersed data with an excess of zeros (Minami et.al (2007)). By this model count outcome is generated by two models. First, a logit model was generated for the “zero,” to assess whether women would be in this group. Then, a NB model was generated for the counts for those women who did not have zero visits. Finally, the two models were combined. probability distribution function of the ZINB regression is given by

$$P(Y=y) = \begin{cases} \omega + (1-\omega) \left(\frac{\alpha^{-1}}{\alpha^{-1} + \mu}\right)^{\alpha^{-1}} & y = 0 \\ (1-\omega) \frac{\Gamma(\alpha^{-1} + y) \mu^{\frac{1}{\alpha}}}{\Gamma(\frac{1}{\alpha})^y} (1 - \mu)^y & y > 0 \end{cases} \quad (3.14)$$

Where α is the overdispersion parameter.

3.4 Model selection

Basically the standard PRM is the starting model to fit for a count response variable. If the count data is characterized by over dispersions, the NB is the alternative model to be fitted. However; the count observations might have excess zeroes than expected in such situations the ZIPRM model could be the appropriate to analyze the dependent variable. If the data have over dispersions and excess zero counts, the ZINB regression model is an alternative for analysis.

3.4.1 Test of over dispersions

An indication of the magnitude of overdispersion for an event in count dataset is simply by comparing the sample mean and variance of the dependent count variable.

If the conditional variance of the response variable exceeds its mean then there is a possibility of overdispersion in the data (A. Colin Cameron and Pravin K. Trivedi (19981)).

The Poisson is the special case of the negative binomial with $\alpha=0$. If α is not equal to zero and actually greater than zero then there is overdispersion. Hence the hypothesis

$$H_0 : \alpha = 0 \text{ vs } H_a : \alpha > 0 \quad (3.15)$$

is a test for overdispersion. The LR test uses -2 times the difference in the fitted log-likelihoods of the two models. Under the null hypothesis (no overdispersion), the PR is preferred to the NB. Alternatively a Wald test can be performed, using the reported t-statistic for the estimated α in the negative binomial model. Then the negative binomial MLE for α is zero, the negative binomial parameter estimates equal the Poisson estimates, and the LR test statistic takes a value of zero (A. Colin Cameron and Pravin K. Trivedi (19981)).

Another useful diagnostic tool detecting overdispersion is the Pearson chi-square statistic. Pearson's chi-square statistic is defined as:

$$\chi^2 = \sum_i \frac{(y_i - \hat{\mu}_i)^2}{\hat{\mu}_i} \quad (3.16)$$

This statistic described in Equation 3.16 above has a limiting chi-square distribution, with degrees of freedom equal to the number of observations minus the number of parameters estimated. If a model is correctly specified and there is no over dispersion, the Pearson chi-square statistic divided by the degrees-of-freedom has an expected value of 1. A ZIP model can account for excess zero. However; if Pearson statistic indicates that there is evidence of model misspecification, with over dispersion being the most likely occurred, then fitting a zero-inflated negative binomial (ZINB) might be a solution.

Akaike Information Criterion (AIC) and Bayesian Information Criterion (BIC) are used as a model fit statistic. And a model with a minimum (AIC) or (BIC) is relatively considered as an appropriate model.

3.5 Model building and variable selection procedures

In order to select the important factors related to the response variable, the backward selection Procedure is used. This strategy is called backward because we are working backward from our

largest starting model that contains all variables to a smaller final model. In this case, the procedure is used to remove covariates with non-significant p-values. This means that variables

that did not contribute to the model based on the highest p-value will be eliminated sequentially and each time a new model, with the remaining covariates will be refitted, until we remained with covariates necessary for answering our research question. that are jointly significant.

CHAPTER FOUR

RESULTS AND DISCUSSION

4.1 Introduction

In this chapter data were analyzed and interpretation of results was given. We start with descriptive statistics to show the distribution of the utilizations of any ANC for different major or fundamental variables.

To investigate the determinants of socioeconomic and demographic features on utilizations of ANC in disadvantaged region of Ethiopia, we compared four count models to choose the appropriate model for our data. That is two basic count models with their respective zero inflated models were compared so that an appropriate model or a better model in fitting the data was selected. Finally we gave interpretations to our results

4.2 Summary of Descriptive Statistics

A total of 2386 child-bearing mothers who have a live birth in the last 5 years preceding the survey were interviewed from four regions (Afar, Somali, BenishangulandGambella), as shown in Table 1. About 1116 (46.8%) of the child-bearing mothers had not received ANC service during their pregnancy period, while 108 (4.5%), 186(7.8%) and 393(24.4%) of the child-bearing mothers had received one, two and three time ANC service respectively. In total, 583 (24.4%) of childbearing mothers had received four and more times ANC service. Table 4.1 reveals that among 2386 women who gave live births in the five years preceding EDHS-2016 in the disadvantaged regions 46.8% do not visited any ANC service. In addition it indicates that the variance of the response variable (27.42) is greater than its mean (2.08) with the measure of the skewness 15.4. Besides, it is shown in figure 4.1 that the percentage of zero counts of the ANC visits is high i.e. almost half. From these we can hypothesize there is overdispersion in the distribution of any ANC visits in the data set.

In table 4.1 we can understand that the pattern of any ANC visits was skewed to the right with higher proportion of zeroes. Furthermore, we can see that the mean value of any ANC visits in the four disadvantageous regions was 2.08 which is less than the WHO recommendations (4+ times).

Tab 4.1 The frequency distribution of number of ANC visit

	Frequency	Percent
No ANC	1116	46.8
1	108	4.5
2	186	7.8
3	393	16.5
4 and above	583	24.4
Total	2386	100
Mean	= 2.08	
Variance	= 27.42	
Skeweness	= 15.4	
Kurtosis	= 281.81	

About two thousand five (83%) respondents were rural women attending ANC of 1.966 on average, whereas 381 (17%) were attending ANC of 3.433 on average. For this, most of the mothers who live in urban were more prevalent of receiving ANC than those who live in rural.

Table 4.2 revealed that as women got higher level of education, their number of ANC visits during pregnancy was increased considerably. About 1,698(71.2%) of women of child bearing age proceeding the survey were not educated and had 1.777 average number of ANC visits where as 128(5.4%) of women with primary education level had 3.813 average number of ANC visits followed by those women of higher education level with 3.063 average number of ANC visits.

The wealth index was also found to be another important indicator for utilizing ANC. Here we found that the change in the amount of income and expenditure of the households were seen with a consistent change in the frequency or the intensity of antenatal care utilization in the four disadvantageous regions during the five years preceding the survey. Women in lower index were loosely linked to ANC as compared to women who belong from the higher wealth indices. Table 4.2 indicate that 57.6 percent with average ANC of 2.298 and 13.9 percent with average ANC

of 3.955 of the poorest and the richest pregnant women respectively were experienced ANC service.

One hundred fourteen (4.8%) child-bearing mothers were below the age of 20 years and ANC was 2.298 averagely. The pregnancy age of the most child-bearing mothers (675 [28.3%]) was between 25 and 29 years, with an average ANC of 2.669.

Table 4.2 indicates that the average ANC visit utilization of childbearing mothers in the 5 years preceding the survey in Afar, Somali, Benishangul Gumuz and Gambela region were 0.560, 0.576, 1.019 and 1.059 respectively. Thus, on the average Gambela region child-bearing mothers had higher number of ANC visits than Afar, Somali and Benishangul Gumuz regions..

Table 4.2 indicates that Catholic (1%), Orthodox (10.5%) and Muslim (68.9%) women who gave live births in the five years preceding the survey had experienced average ANC visits of 3.2, 2.873 and 2.123 respectively.

Women from male headed HH in the study (73%) who gave birth in the five years preceding the survey had experienced 2.262 average number of ANC visits. Whereas those women from female headed house hold (27%) who gave birth in the five years preceding the survey had 2.034 average number of ANC visits.

About 26.8% of women who gave birth in the five years have work, and had average ANC visit of 2.635 and 73.2% of women without work were with the average ANC visit of 2.041.

Table 4.2 indicate that among women who gave birth in the five years preceding the survey and who are autonomous on the decision of health care service (20.7%) had 2.799 ANC visits on the average.

Tab 4.2 The frequency distribution of number of ANC visit

Back ground characteristics	Frequency(percent)	Mean	Std.Dev
Age in 5 years			
15-19	114(4.8%)	2.298	2.194
20-24	562(23.6%)	2.536	4.701
25-29	675(28.3%)	2.669	8.521
30-34	474(19.9%)	2.002	2.061
35-39	414(17.4%)	1.681	5.088
40-44	117(4.9%)	0.795	1.500
45-49	30(1.1%)	0.767	1.223
Education Level			
No education	1,698(71.2%)	1.777	6.031
Primary	504(21.1%)	3.063	4.876
Secondary	128(5.4%)	3.813	1.999
Higher	56(2.3%)	3.571	1.906
Region			
Afar	582(24.4%)	0.560	0.7250
Somali	754(31.6%)	0.576	0.707
Gambela	481(20.2%)	1.019	0.835
Benishangul	569(23.8%)	1.059	0.872

Place of residence			
Rural	2,005(83%)	1.966	6.028
Urban	381(17%)	3.433	2.297
Religion			
Orthodox	251(10.5%)	2.873	2.495
Catholic	25(1%)	3.2	2.141
Protestant	427(17.9%)	2.194	2.284
Muslim	1,645(68.9%)	2.123	6.586
Traditional	25(1%)	0.32	1.108
Others	13(0.7%)	0.846	1.144
Frequency of reading news paper			
Not at all	2,318(97.2%)	2.1665229	5.6995605
Less than once a week	39(1.6%)	3.8461538	1.0396979
At least once a week	29(1.2%)	2.6896552	1.8342661
Frequency of listening to radio			
Not at all	2,034(85.2%)	1.982	5.611
Less than once a week	183(7.7%)	3.683	7.238
At least once a week	169(7.1%)	3.225	2.794
Frequency of watching television			
Not at all	2,069(86.7%)	1.983	5.954
Less than once a week	129(5.4%)	2.519	1.746

At least once a week	188(7.9%)	4.378	2.079
Sex of HH			
Male	1,742(73%)	2.262	5.595
Female	644(27%)	2.034	5.715
Literacy			
Cannot read at all	1,887(79.1%)	1.852	5.770
Read only parts of sentence	199(8.3%)	3.809	2.402
Read whole sentence	186(7.8%)	3.183	2.249
No card with required language	114(4.8%)	3.561	9.174
Wealth index			
Poorest	1,374(57.6%)	1.114	4.019
Poorer	271(11.4%)	2.232	2.059
Middle	233(9.8%)	5.069	12.453
Richer	175((7.3%)	3.52	7.574
Richest	333(13.9%)	3.955	2.029
Work status			
Yes	639(26.8%)	2.635	6.870
No	1,747(73.2%)	2.041	5.090
Husband/Partner education			
No education	1,395(58.5%)	1.409	4.072
Primary	534(22.4%)	3.358	8.601

Secondary	216(9.1%)	3.505	6.775
Higher	241(10%)	3.046	2.015
Person who usually decides on respondent's health			
Respondent alone	493(20.7%)	2.799	9.865
Respondent and husband/Partner	1,314(55.1%)	2.284	4.346
Husband/partner alone	579(24.2%)	1.525	2.031

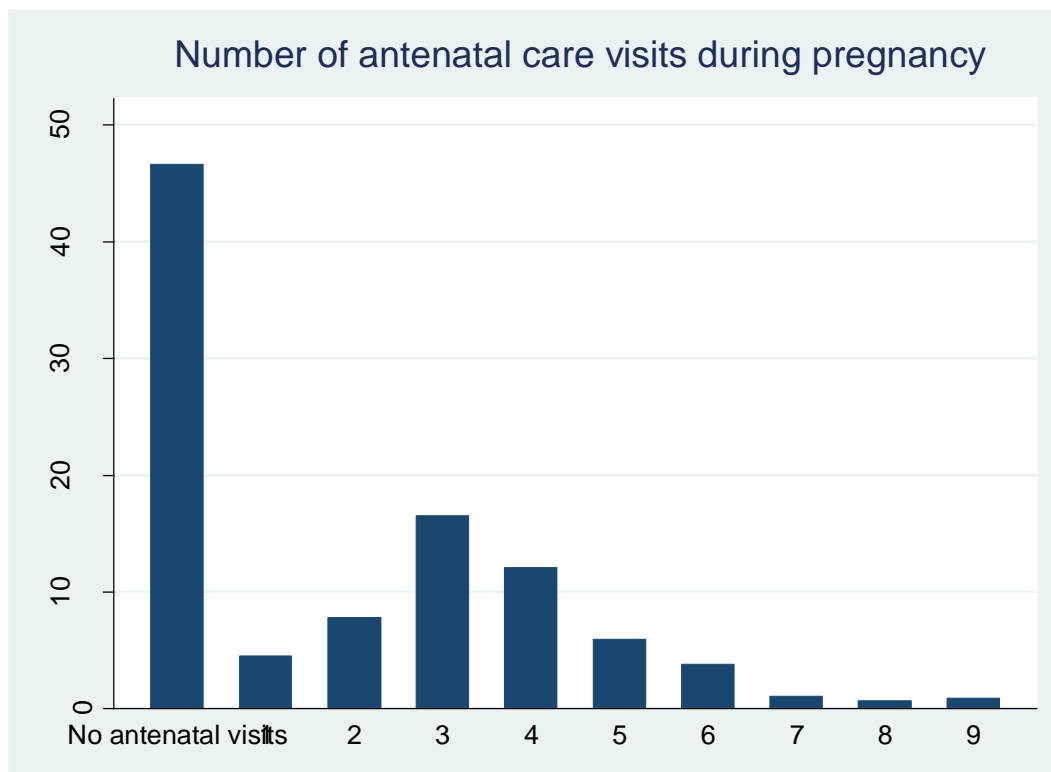


Figure 4.1 Number of antenatal care visits during pregnancy

4.3 Test for over dispersion

Test of overdispersion on standard Poisson regression were made using Deviance and Pearson chi square Goodness of fit test. Thus, the tests confirmed that there is overdispersion in the data.

Model	Deviance goodness of fit (P-value)	df	Deviance Goodness of fit /df	Pearson chi square Goodness of fit	df	Pearson chi square Goodness of fit/df (P-value)
Standard Poisson	7390.92 (<0.001)	2324	3.2	20293	2324	8.73 (<0.001)

Young test of ZIP vs Standard Poisson (Z =33.45 ,P<0.001)

Young test of ZINB vs standard Negative Binomial (Z=66.83,P<0.001)

4.4 Model Comparison

Count data models were compared using Akaike information criterion (AIC) and Bayesian Information criterion (BIC). To incorporate all variables into the model building workflow for count data regression models, their model information would be used to complement standard techniques such as information criteria (AIC) and (BIC). Using a basic model for the number of ANC visit data (Poisson vs. NB models) and/or extra zeros (zero-inflation) are relevant issues in the data at hand. The models upon which the variables are based should use a reasonable first selection of regressor. (a standard specification from the literature or a model involving all potentially relevant variables). In much literature reviewed that the models with smallest AIC and (BIC) values is the better one. The value of young statistics was 66.83 with $P < 0.001$, indicating that over-dispersion due to many zero observations and unobserved heterogeneity. Moreover, both the AIC and BIC values also supported the conclusion that the ZINB model provided the best fit to the frequency of ANC care use data (Table 4.3)

Table: 4.4 AIC and (BIC) values

No	Model	AIC	BIC
1	Poisson	118458.78	12027.88
2	Zero-inflated Poisson(ZIP)	7309.16	7494.04
3	Negative Binomial(NB)	8640.15	8778.81
4	Zero inflated Negative Binomial(ZINB)	5817.76	6008.41

4.5 ZINB regression model analysis

4.5.1 NB part of ZINB Model

Table 4.5 shows the results from the NB part of the ZINB regression model. The model estimation applied Bootstrap to make the parameter estimation more precise and control for minor violations related to the underlying assumptions. The model was run to estimate the

incidence rate ratio (IRR), which measures the rate at which the event (ANC visits) occurs. Table 4.5 illustrates the IRRs estimated across the explanatory variables.

Age at birth was the major variable that had impact on the utilizations ANC in the last five years. Table 4.5 revealed that, the expected number of ANC visits for women of age group, 30-34, 35-39 and 40-44 and 45-49 were respectively equal to 0.754, 0.751, 0.314 and 0.385 times the expected number of antenatal care visits for women of age group 15-19.

Women who were pregnant five years preceding the survey and were able to read whole sentence had 0.791 times the expected number of ANC visits compared to woman who were not able to read at all.

Women's husband/partner education level showed a significant effect on the number of ANC visit during their pregnancy period. As shown in table 4.5 the expected number of antenatal care visits of women whose husband/partner education level were secondary, higher and primary education were 2.018, 1.438 and 1.431 times the expected number of antenatal care visits by women whose husband/partner had no education.

Exposure to different Medias significantly affected the number of ANC visit. Women who listened radio and watched television in less than once a week had 2.23 and 0.601 times the expected number of antenatal care visits compared to those women who did not at all respectively.

Table 4.5 revealed that the expected number of ANC visits of a woman in middle and richest indices were 4.121 and 3.017 times the expected number of antenatal care visits compared to poorest women respectively.

Decision for healthcare being taken by the husbands was significantly associated with ANC visits showing 0.689 times of having the expected number of antenatal care visits compared to decision taken by women alone.

Table 4.5 Parameter estimates of negative binomial part of the ZINB regression model across the explanatory variables

Variables	IRR	Coefficient	Std.error	Z-value	P-value	95% CI
Age in 5 years						
15-19(Reference)						
20-24	0.849	-0.164	0.136	-1.20	0.228	[-0.431 0.103]
25-29	0.871	0.1386	-0.138	-1.01	0.310	[-0.404 0.128]
30-34	0.754	-0.282	-0.142	-1.99	0.047	[-0.561 -0.004]
35-39	0.751	-0.287	0.144	-1.99	0.046	[-0.568 -0.005]
40-44	0.314	-1.159	.202729 4	-5.72	0.000	[-1.557 -0.762]
45-49	0.385	-0.953	0.329	-2.90	0.004	[-1.598 -0.308]
Literacy						
Not at all(Reference)						
Able to read only parts of sentence	0.964	-0.036	0.104	-0.35	0.727	[-0.241 0.168]
Able to read whole sentence	0.791	-0.234	0.112	-2.09	0.037	[-0.453 -0.014]
No card with required language	1.577	0.455	0.134	3.40	0.001	[0.193 0.718]
Frequency of listening to radio						
Not at all(Reference)						
Less than once a week	2.237	0.805	0.103	7.85	0.000	[0.604 1.006]
At least once a week	1.426	0.355	0.106	3.36	0.001	[0.148 0.562]
Frequency of watching television						

Not at all(Reference)						
Less than once a week	0.601	-0.509	0.132	-3.86	0.000	[-0.767-0.251]
At least once a week	1.044	0.043	0.126	0.35	0.728	[-0.202 0.289]
Wealth index						
Poorest(Reference)						
Poorer	1.928	0.656	0.088	7.49	0.000	[0.485 0.828]
Middle	4.121	1.416	0.091	15.53	0.000	[1.237 1.594]
Richer	2.559	0.939	0.107	8.77	0.000	[0.729 1.149]
Richest	3.017	1.104	0.111	9.94	0.000	[0.886 1.322]
Husband/ Partner's Education level						
No education(Ref.)						
Primary	1.431	0.358	0.073	4.93	0.000	[0.216 0.500]
Secondary	2.018	0.702	0.099	7.08	0.000	[0.5080.897]
Higher	1.438	0.363	0.107	3.40	0.001	[0.154 0.573]
Respondent work						
No(Reference)						
Yes	1.235	0.211	0.084	3.092	0.002	[1.080 1.413]
Person who usually decides on respondent's health						
Respondent alone(Reference)						
Respondent and husband/Partner	0.753	-0.284	0.069	-4.08	0.000	[-0.419 -0.147]
Husband/partner alone	0.689	-0.371	0.084	-4.440	0.000	[-0.535 -0.207]
_cons	1.384	0.325	0.139	2.321	0.020	[0.051 0.599]
Alpha	0.078	1.081	0.060			[0.969 1.206]

4.5.1 Zero-inflation part of ZINB Model

Table 4.6 revealed that, the probability of being in the excess Zero group for women of age group, 30-34 and 40-44 were decreased by 19.7% and 29.5% respectively as compared to the reference group (i.e., pregnant women of age group 15-19).

The model also suggested that the probability of being in the excess zero group was 13.2% lower for women residing in Benishangul region compared to those in Afar region.

Women who were pregnant five years preceding the survey and with no card with required language had 21.9% increase in the probability of being in the excess Zero group compared to woman who were not able to read at all.

Women who read news paper at least once a week had 16.8% decrease in the probability of being in the excess zero group compared to those women who did not read at all. Similarly women who listened radio and watched television in less than once a week had an increase in the probability of being in the excess Zero group by 26.1% and a decrease by 33% respectively compared to those women who did not listen and watch at all.

Table 4.6 revealed that the probability of being in the excess zero group of a woman in middle and richer indices was increased by 58.7% and 66.7% respectively when compared to poorest women.

As shown in table 4.6 probability of being in the excess zero group of a woman with primary and secondary level of their husband/partner education were increased by 35.3% and 37.6% compared to husband/partner with no education respectively.

The probability that women were in the zero group decreased by 27.4% and 22.8% respectively when the women had decision on healthcare by the husband and both compared to women who had decision by her own.

Table 4.6 Zero-inflation part of ZINB model parameter estimation for ANC Visit for the Selected Variables

Variables	IRR	Coefficient	Std.error	z	P-value	95% CI
Age in 5 years						
15-19(Reference)						
20-24	0.916	-0.088	0.099	-0.887	0.375	[-0.282 0.106]
25-29	1.117	0.111	0.099	1.117	0.264	[-0.084 0.305]
30-34	0.803	-0.209	0.106	-1.976	0.048	[-0.416 -0.002]
35-39	1.009	0.009	0.111	0.081	0.936	[-0.208 0.226]
40-44	0.705	-0.350	0.170	-2.061	0.003	[-0.683 -0.017]
45-49	0.594	-0.521	0.288	-1.810	0.070	[0.367 0.873]
Region						
Afar(Reference)						
Benishangul	0.868	-0.142	0.069	-2.067	0.039	[-0.276 -0.007]
Literacy						
Not at all(Reference)						
Able to read only parts of sentence	0.940	-0.061	0.064	-0.904	0.366	[-0.195 0.072]
Able to read whole sentence	0.949	-0.052	0.081	-0.611	0.541	[-0.221 0.116]
No card with required language	1.219	0.198	0.090	2.193	0.028	[0.021 0.375]
Frequency of reading news paper						
Not at all(Reference)						
Less than once a week	0.887	-0.120	0.134	-0.890	0.374	[-0.383 0.144]
At least once a week	0.681	-0.384	0.166	-2.312	0.021	[-0.710 -0.058]
Frequency of listening to radio						
Not at all(Reference)						
Less than once a week	1.261	0.232	0.067	3.437	0.001	[0.100 0.364]
At least once a week	1.145	0.135	0.073	1.844	0.065	[-0.009 0.279]

Frequency of watching television						
Not at all(Reference)						
Less than once a week	0.670	-0.401	0.093	-4.316	0.000	[-0.583 -0.219]
At least once a week	1.022	0.022	0.079	0.274	0.784	[-0.134 0.177]
Wealth index						
Poorest(Reference)						
Poorer	1.164	0.152	0.070	2.182	0.029	[0.015 0.288]
Middle	1.587	0.462	0.065	7.148	0.000	[0.335 0.589]
Richer	1.667	0.511	0.078	6.547	0.000	[0.358 0.664]
Richest	1.318	0.276	0.076	3.653	0.000	[0.128 0.424]
Husband/ Partner's Education level						
No education(Reference)						
Primary	1.353	0.303	0.054	5.632	0.000	[0.197 0.408]
Secondary	1.276	0.244	0.069	3.513	0.000	[0.108 0.380]
Higher	1.043	0.042	0.078	0.542	0.588	[-0.111 0.196]
Person who usually decides on respondent's health						
Respondent alone(Reference)						
Respondent and husband/Partner	0.772	-0.259	0.052	-4.980	0.000	[-0.361 -0.157]
Husband/partner alone	0.726	-0.320	0.064	-5.014	0.000	[-0.445-0.1 95]
_cons	3.769	1.327	0.103	12.866	0.000	[1.125 1.529]
Alpha	0.237	0.237	0.015			[0.209 0.268]

4.6. Discussion

In the current study, 2386 women who gave birth at least once within the last 5 years before the survey were interviewed from four regions (Afar, Somali, Benishangul and Gambela). The child-

bearing mothers who attend at least one ANC visits were 52.2% and 24.33% of women of child bearing had at least 4 visits, but in the study done in Kaffa, Sheka, and Bench Maji Zones of Southwestern Ethiopia(Abiyot N.et.al(2019)),out of 1715interviewed women (91.9%) of them had at least one ANC visit during their pregnancy. Majority of women, (66.6%), made at least 4 ANC visits during their pregnancy., which is much more than the proportion for current study.

The current study showed that the age at pregnancy is significantly influencing the utilization of ANC. The study also showed that the attendance is higher for pregnant women whose age is in between 30 and 34 and get lower as age increases. It is supported by a nationally representative sample survey in Ethiopia; receipt of maternity care was found to vary by age, and other sociodemographic factors (CSA, 2011)

Use of public media sources like listening radio and watching television increases the awareness of people on health and other matters. In this study, use of these public media sources significantly affected the use of ANC services. Frequency of listening to radio has significant increasing effect than frequency of watching television. This may be due to the low proportion of households who have access to television and news paper in the regions. A study in Nigeria has shown that community media saturation was found to be a strong predictor of maternal health service utilization (Babalola,(2009)).

Decision making power of women can have a significant effect on the ability of women to seek health services and/or contribute to delays in accessing and receiving medical care even in places where services are readily available(Ahmed et al.,2010)). In this study, women who have been able to decide on health care by themselves were more likely to have the expected number of ANC visits than women whose health care decision was controlled by other people. This may be because if resources are controlled by others women do not have the freedom to use services whenever they need care. This may be because if resources are controlled by others, women do not have the freedom to use services whenever they need care

This study has found that household wealth status is significantly associated with the utilization of ANC services. Women who are from household with middle and a higher wealth quintile are more likely to utilize ANC services than those who are from the poorest wealth households. This result is consistent with other similar studies (Arthur E. (2014)). This is expected since access to health services utilization in Ethiopia mainly depends on out of pocket payment (Ministry of

Health of Ethiopia(2010)). Though the services forANC are exempted, women are expected to pay for medications and additional transportation costs contribute to the high cost of seeking care and maydeter women from utilizing services. This has also been revealed in other similar studies (Ahmed etal.,2010).

This study has found thatofhusband/ partner's education levelhad a significant effect on the utilization of the ANC services.. These results have been consistently supported by many other studies which showed a positive influence of education on maternal health service utilization (Mengesha ZB etal.,2013)]. Educated husbands may have a better communication with their wives and willingness to discuss the use of maternal health services. They may also provide more autonomy to their wives (Ahmed etal.,2010)).

CHAPTER FIVE

CONCLUSIONS AND RECOMMENDATIONS

This study showed that in the four economically disadvantaged regions availing of at least one ANC visit was 53.2% and at least 4 visits was 24.40%, which is very low. Husband's education, literacy, decision maker for healthcare, and media exposure like listening radio, watching television Wealth index,, and age of women were the important factors which were significantly associated with the use of ANC visits. Thus, we conclude that socio economic and demographic factors determine the number of ANC, and intervention at various levels is recommended.

The average number of the frequency (2.08) of visiting for ANC in the four economically disadvantaged regions did not match to the recommended number (at least four) of ANC visit by the WHO..Benishangul region had relatively better proportion of women who had ANC visit(40.73%) followed by Gambella(35%) and Somali region had the least proportion of ANC visit(12.23%) almost similar with Afar region (16.37).Thus we concluded that there were differences observed among regions on the proportion of women who had utilization of antenatal cares and we recommend that special needs in promoting ANC across regions should be prioritized and structured for better implementations in all respective regions.

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June,2021

Addis Ababa, Ethiopia.

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Annex

A1. Output for Poisson regression model

A1.2. Akaike's information criterion and Bayesian information criterion for Poisson regression model

Akaike's information criterion and Bayesian information criterion

Model	Obs	ll(null)	ll(model)	df	AIC	BIC
.	2,386	-7280.076	-5893.39	31	11848.78	12027.88

Note: N=Obs used in calculating BIC; see [\[R\] BIC note](#).

A2. Output for Zero inflated Poisson regression model

```

Zero-inflated Poisson regression      Number of obs   =    2,386
                                      Nonzero obs     =    1,277
                                      Zero obs       =    1,109

Inflation model = logit              LR chi2(29)    =    751.39
Log likelihood = -3622.58             Prob > chi2    =    0.0000
  
```

Number_of_antenatal_care_visit	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
Number_of_antenatal_care_visit						
Age_in_5_year_groups						
20-24	-.1116438	.0710388	-1.57	0.116	-.2508773	.0275897
25-29	.1444443	.0710288	2.03	0.042	.0052304	.2836582
30-34	-.2969999	.0770933	-3.85	0.000	-.4481001	-.1458998
35-39	-.0497284	.0797853	-0.62	0.533	-.2061047	.106648
40-44	-.3974366	.1262846	-3.15	0.002	-.64495	-.1499233
45-49	-.6865756	.2230251	-3.08	0.002	-1.123697	-.2494545
Region						
Somali	.1809214	.0448475	4.03	0.000	.0930219	.2688209
Benishangul	-.2029249	.0498343	-4.07	0.000	-.3005983	-.1052516
Gambela	-.0922599	.0503937	-1.83	0.067	-.1910296	.0065099
Type_of_place_of_residence						
Rural	.0906433	.0454952	1.99	0.046	.0014743	.1798123
Education_level						
Primary	-.1047387	.0402888	-2.60	0.009	-.1837034	-.0257741
Secondary	.00618	.0673321	0.09	0.927	-.1257885	.1381484
Higher	-.3581268	.0896522	-3.99	0.000	-.533842	-.1824117
Sex_of_HH_head						
Female	-.1040097	.0349938	-2.97	0.003	-.1725963	-.035423
Frequency_of_reading_news_paper						
Less than once a week	-.0878405	.0937939	-0.94	0.349	-.2716732	.0959923
At least once a week	-.4512545	.1212409	-3.72	0.000	-.6888822	-.2136267
Frequency_of_watching_television						
Less than once a week	-.3135012	.066047	-4.75	0.000	-.4429509	-.1840515
At least once a week	.0627085	.0515299	1.22	0.224	-.0382882	.1637053
Wealth_index						
Poorer	.1647464	.0510987	3.22	0.001	.0645948	.2648979
Middle	.5253245	.0437733	12.00	0.000	.4395303	.6111186
Richer	.6142715	.0543934	11.29	0.000	.5076625	.7208805
Richest	.2994706	.060277	4.97	0.000	.1813297	.4176114
Husband_partner_education						
1	.3891898	.0376855	10.33	0.000	.3153275	.4630521
2	.2652745	.0490423	5.41	0.000	.1691532	.3613957
3	.1549897	.0571417	2.71	0.007	.0429939	.2669855
Person_who_decide_on_heakth_care						
Respondent and husband/partner	-.3349096	.0355461	-9.42	0.000	-.4045788	-.2652405
Husband/partner alone	-.4247991	.0444831	-9.55	0.000	-.5119844	-.3376138
Someone else	-6.670358	1286769	-0.00	1.000	-2522027	2522013
Other	-7.689499	1301318	-0.00	1.000	-2550544	2550529
_cons	1.362627	.0895722	15.21	0.000	1.187068	1.538185
inflate						
Number_of_antenatal_care_visit	-45.85093	7519.823	-0.01	0.995	-14784.43	14692.73
_cons	23.7523	4462.442	0.01	0.996	-8722.474	8769.979

A2.1. Akaika's information criterion and Bayesian information criterion for zipm model

Akaike's information criterion and Bayesian information criterion

Model	Obs	ll(null)	ll(model)	df	AIC	BIC
.	2,386	-3998.276	-3622.58	32	7309.16	7494.036

Note: N=Obs used in calculating BIC; see [\[R\] BIC note](#).

A3. Output for Negative Binomial regression model

Number_of_antenatal_care_visit	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
Age_in_5_year_groups						
20-24	-.1642513	.1363319	-1.20	0.228	-.431457	.1029543
25-29	-.1377876	.1358409	-1.01	0.310	-.4040309	.1284556
30-34	-.2825103	.1421601	-1.99	0.047	-.561139	-.0038817
35-39	-.2866383	.1437112	-1.99	0.046	-.568307	-.0049696
40-44	-1.159421	.2027294	-5.72	0.000	-1.556763	-.7620783
45-49	-.9533475	.3291063	-2.90	0.004	-1.598384	-.308311
Litracy						
Able to read only parts of sentence	-.0364355	.1043165	-0.35	0.727	-.2408921	.1680211
Able to read whole sentence	-.2339587	.1119772	-2.09	0.037	-.45343	-.0144873
No card with required language	.4554086	.133922	3.40	0.001	.1929263	.717891
Frequency_of_listening_to_radio						
Less than once a week	.8051719	.1025455	7.85	0.000	.6041864	1.006158
At least once a week	.3550131	.105683	3.36	0.001	.1478783	.562148
Frequency_of_watching_television						
Less than once a week	-.5087931	.131683	-3.86	0.000	-.7668869	-.2506992
At least once a week	.0436414	.125669	0.35	0.728	-.2026653	.289948
Wealth_index						
Poorer	.6562659	.0875942	7.49	0.000	.4845844	.8279474
Middle	1.416121	.0911813	15.53	0.000	1.237409	1.594833
Richer	.9394973	.1071475	8.77	0.000	.729492	1.149503
Richest	1.104104	.1110729	9.94	0.000	.8864052	1.321803
Husband_partner_education						
1	.3584193	.0727008	4.93	0.000	.2159284	.5009102
2	.7022999	.0992029	7.08	0.000	.5078657	.8967341
3	.3634732	.1069482	3.40	0.001	.1538586	.5730879
Person_who_decide_on_heakth_care						
Respondent and husband/partner	-.2838024	.0694897	-4.08	0.000	-.4199998	-.147605
Husband/partner alone	-.3715474	.0837256	-4.44	0.000	-.5356466	-.2074481
Someone else	-20.31075	11816.37	-0.00	0.999	-23179.97	23139.35
Other	-19.60845	6316.117	-0.00	0.998	-12398.97	12359.75
_cons	.3250711	.139975	2.32	0.020	.0507251	.599417
/lnalpha	.0782664	.0556996			-.0309029	.1874356
alpha	1.081411	.0602342			.9695697	1.206153

LR test of alpha=0: chibar2(01) = 3099.22

Prob >= chibar2 = 0.000

]

A3.1. Akaike's information criterion and Bayesian information criterion for Negative Binomial regression model

Akaike's information criterion and Bayesian information criterion

Model	Obs	ll(null)	ll(model)	df	AIC	BIC
.	2,386	-4630.414	-4296.074	24	8640.149	8778.806

Note: N=Obs used in calculating BIC; see [\[R\] BIC note](#).

A4.1. Akaike's information criterion and Bayesian information criterion for ZIBN model

Akaike's information criterion and Bayesian information criterion

Model	Obs	ll(null)	ll(model)	df	AIC	BIC
.	2,386	-3015.17	-2875.877	33	5817.753	6008.407

Note: N=Obs used in calculating BIC; see [\[R\] BIC note](#).