Individual and Area Level Factors of Low Birth Weight in Ethiopia: Evidence from Ethiopian Demographic Health Survey 2016: Multilevel Modeling

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Abstract

Background Low birth weight is one of the critical issues in Ethiopia that causes many babies' short-term and long-term health consequences. In Ethiopia, low birth weight is increasing; however, limited evidences of multilevel factors associated with low birth weight in the study setting, Ethiopia. **Objective** The objective of this study was to assess individual and area level factors of Low Birth Weight in Ethiopia: from Ethiopia Demographic and Health Survey 2016. **Methods** The data were extracted from the 2016 Ethiopia Demographic and Health Survey. For analysis a sample of 2110 birth weights within five years preceding the survey were included. The analysis was carried out using STATA software version 14. A two level mixed effects logistic regression analysis was used to estimate both the fixed effects of the individual and contextual factors and the random effects of the between cluster difference. Adjusted Odds Ratio with 95% Confidence Interval to express measures of association and Intra Class Correlation to express measures of variation were used. **Results** A total of 2110 children nested within 445 clusters were included in the analysis. Among them, 13% were with low birth weight. The ICC implied 11.7% of the variance in low birth weight was attributable to Area level unobservable factors. At individual level; multiple birth (AOR=2.74; 95%CI: 1.450-5.184), preterm birth (AOR=4.83; 95%CI: 2.644-8.830), anemic mothers (AOR=1.49; 95% CI: 1.069-2.092), six and above birth order (AOR=0.42; 95%CI: 0.242-0.752), mothers with primary educational level (AOR=0.61; 95%CI: 0.418-0.896) and secondary/higher educational level (AOR=0.39; 95%CI: 0.252-0.612) as well as region from Area level were significantly associated with low birth weight. **Conclusion** The results of this study showed that multiple births, anemic mothers, birth order, not-educated mothers and preterm gestational age at birth were significant factors of low birth weight. Hence, switching off/on the significant factors accordingly could reduce the risk of having low birth weight child.

Background

Low birth weight (LBW), which is defined by the world health organization (WHO) as birth weight less than 2500 grams, is associated with a higher risk of neonatal and infant mortality and morbidity, and a greater risk for adverse health outcomes, cognitive development and school performance problems than those born with normal weight. The subcategories of low birth weight are: very low birth weight (VLBW) which is <1500gram, extremely low birth weight (ELBW) which is <1000gram. [1-3].

Low birth weight caused due to either preterm (before 37 weeks of gestational age) or intra-uterine growth retardation (IUGR) and poor health care during pregnancy. The IUGR is often as a result of maternal nutrition either before conception or during pregnancy as well as fetal problem [4].

Low birth weight has public health importance including increased neonatal mortality and morbidity, physical and psychomotor development delay. Besides infant LBW are more likely to develop significant disabilities and there are long term health implication of future chronic disease [4]. LBW due to restricted fetal growth affect the person throughout life and is associated with poor growth in childhood and a

higher incidence of adult disease, such as type 2 diabetes, hypertension and cardiovascular disease. An additional risk for girls is having smaller babies when they become mothers [5].

Based on Literatures, the factors associated with premature delivery and low birth weight include: socio economic status, maternal education and occupational, residence, maternal anthropometric status such as: maternal stature, Mid Upper Arm Circumference (MUAC) and multiple pregnancies [6, 7].

Triggering LBW, previous history of preterm/LBW/IUGR, maternal age, birth interval, inadequate weight gain in pregnancy, infectious and improper nutrition are other predictors of LBW. From the maternal factors, stress, smoking, and use of alcohol, pollution, violence and genetic factor are some of the determinant factors of LBW [8].

In 2012, the world health assembly resolution endorsed a comprehensive implementation plan on maternal, infant and young child nutrition, which specified six global nutrition targets for 2025. The third target of this policy briefly covers 30 percent reduction of LBW. The goal is to achieve a 30% reduction of the number of infant born with a weight lower than 2500 gram by the year 2025. This would translate in to 3.9 percent relative reduction per year between 2012 and 2025 [9].

Methods

Low birth weight (LBW), which is defined by the world health organization (WHO) as birth weight less than 2500 grams, is associated with a higher risk of neonatal and infant mortality and morbidity, and a greater risk for adverse health outcomes, cognitive development and school performance problems than those born with normal weight. The subcategories of low birth weight are: very low birth weight (VLBW) which is <1500gram, extremely low birth weight (ELBW) which is <1000gram. [1-3].

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1.1 Statement of the Problem

About 20 million infants world-wide accounting for 15.5 percent of all births are born LBW, 95.6 percent of them in developing countries. In developing countries 16.5 percent of infants are born LBW, 13 percent in sub Saharan Africa. LBW is a major public health problem in under-resourced settings [10]. There are 15 million preterm birth (PTB) annually, and more than 1 million of infants born preterm die due to early complications. The rate of PTB is increasing, particularly in sub-Saharan Africa and South Asia where over 60 percent of global PTBs occur [11].

Babies born low birth weight are 37 percent more likely to die during infancy compared to those of normal weight if other factors are held constant. Therefore LBW is strongly negatively associated with infant survival [12]. In a study conducted in East Africa, preterm babies and babies with LBW were found to account for 52 percent of newborn deaths in East Africa [13].

Being born with LBW is generally recognized as a disadvantage for the infant, family and country. Preterm birth is a direct cause of 28 percent of the 4 million neonatal deaths that occur globally every year. Direct or indirect, LBW may contribute to 60 percent to 80 percent of all neonatal death. LBW infant are at higher risk of early growth retardation, infection, developmental delay and death during infancy and childhood [14]. Studies have shown that infants weighing less than 2500 gram were approximately 20 times more likely to die than heavier babies [5].

Based on the united nation report in Ethiopia the prevalence pattern of LBW in 2000 GC was 15 percent but in 2005 GC report was 20.5 percent [15]. The pattern of LBW in Ethiopia was increased from 11 percent to 13 percent in 2011 and 2016 respectively. Based on a different researches done in Ethiopia was reported the prevalence of LBW in Addis Ababa 11 percent, and in Jima 22.5 percent and Gondar 17.4 percent and the prevalence of LBW in Tigray was 9.9 percent [3, 15, 16].

In Ethiopia, using a single level analysis, there are a lot of studies done on LBW, however; the multilevel factors associated with LBW, are not well addressed. For data with a hierarchical nature, using single level logistic regression can bias the parameter estimation. Multilevel models allow one to account for the clustering of subjects within clusters of higher-level units when estimating the effect of subject and cluster characteristics on subject outcomes and gives us appropriate parameter estimation for nested data.

Limited evidences are available regarding predictors of LBW in Ethiopia at nationally representative sample.

So this study has taken a step further from the routine EDHS report by further analysis of DHS data using advanced analysis model to assess multilevel factors of LBW using a multilevel logistic regression model and provides context specific information to program planners and policy makers.

2. Objectives

2.1 General Objective

To assess the individual and Area level factors of low birth weight in Ethiopia evidence from EDHS 2016.

2.2 Specific Objectives

To identify the individual level factors associated with low birth weight in Ethiopia EDHS 2016.

To identify the Area level factors associated with low birth weight in Ethiopia EDHS 2016.

Results

1.1 Study area and period

Ethiopia is situated in the Horn of Africa with 9 Regional States and two city administrations. The capital city of Ethiopia is Addis Ababa. Study was conducted in all nine geographical regions and two administrative cities of Ethiopia which are included by EDHS 2016 [3]. Study period of this study is from May to December of 2018.

The current health policy of Ethiopia gives much more emphasis on prevention and the health promotion components of health care that should be able to resolve most of the health problems of the population [17].

1.2 Data source

The child dataset used for this analysis was the 2016 EDHS. It is the latest and the nationally large scale dataset of demographic and health survey that was conducted by the Central Statistical Agency (CSA) from January 18, 2016 to June 27, 2016 with nationally representative sample from 9 regions and two administrative cities. The total study participant to this study was 2110 infants whose weights were recalled. Details of sampling design and selection of sample are available in the Ethiopia Demographic and Health Survey 2016 EDHS reports [3].

1.3 Study design

A cross sectional study design was used to identify multilevel factors associated with LBW from the 2016 EDHS data collected by the CSA.

1.4 Population

1.4.1 Source population

All live births in the five years born to women's of reproductive age of 15-49 years who were residents of the nine regions and two administrative cities of Ethiopia during the survey.

1.4.2 Study population

All live births in the five years born to women's of reproductive age of 15-49 years, who were residents of the selected households in the selected enumeration areas during the survey.

1.4.3 Sample population

All live births in the five years born to women's of reproductive age of 15-49 years, who were residents of the selected HHs in the selected enumeration areas during the survey, Study population who fulfilled the inclusion criteria.

1.5 Eligibility criteria

1.5.1 Inclusion and exclusion criteria

All Birth Weighted infants in EDHS 2016 sampled areas were included in the study.

All mothers who had not weighted their children and do not know their child's weight were excluded.

1.5.2 Operational and standard definition

Normal birth weight- is a weight class greater than or equal to 2500 gram.

Low Birth Weight- is a weight at birth less than 2500 gram.

Individual level factors: A variable operating at the lowest level or individual level which included children's, parents and household characteristics.

Area level factors: The term Area refers to clustering of individuals within same geographic environment.

1.6 Sample size and Sampling procedures

Each region was stratified into urban and rural areas yielding 21 sampling strata. Samples of enumeration areas were selected independently in each stratum in two stages. In the first stage an enumeration areas 645 (202 urban areas and 443 rural areas) were selected. In the second stage of selection a 28 households per cluster were selected with an equal probability systematic selection from

the newly created household listing. Overall, 18,008 households were selected of which 17,067 were occupied. Selected households were visited and interviewed. All women aged 15-49 years were eligible to be interviewed. Of all the child dataset related to birth weight, 2110 infants are eligible to this study.

Figure 2: Graphical representation of sampling procedure for low birth weight in Ethiopia from EDHS 2016.

1.7 Data quality control

Data related to the outcome variable of low birth weight was selected and extracted from the child dataset of EDHS 2016. Further data cleaning, labeling, coding and recoding were done for all selected variables. Categorization was done for continuous and categorical variables using information from different literatures accordingly.

1.8 Study variables

1.8.1 Dependent variable

The outcome variable is low birth weight.

The dependent variable for the i^{th} birth weight was represented by a random variable with two possible values coded 1 and 0. So, the response variable of the i^{th} birth weight was measured as a dichotomous variable.

($Y_i=1$ if low birth weight was occurred, otherwise $Y_i=0$)

1.8.2 Independent variables

The explanatory variables were considered at two levels individual and Area level factors.

Individual level variable

Individual level factors considered in this study includes: Age of mother, maternal age at first birth, gestational age at birth, birth order, type of birth, sex of the child, sex of household head, marital status, mother's educational level, husband/partner of mothers' educational level, maternal occupation, husband/partner of mothers' occupation, cigarette smoking, household wealth index, religion, household size, media exposure, body mass index.

Area level variables includes: In addition to region, place of residence, Area media exposure, Area educational status, Area poverty status. The aggregated Area level predictor variables were constructed by aggregating individual level values at cluster level and binary categorization of the aggregated variables were done based on the distribution of the proportion values calculated for each cluster(Area).

1.9 Statistical methods of Analysis

Multilevel Modeling: Nature of nested data makes the uses of traditional regression methods inappropriate because of the assumption of independence among individual within the some group, assumption of equal variance across groups which an inherent in traditional regression methods are violated. Therefore, multilevel model is a type of regression analysis for multilevel data where the dependent variable is more appropriate for hierarchically structured data, such as the DHS to estimate the robust standard error. So in this study multilevel binary logistic regression analysis was employed in order to account for the hierarchical nature of the DHS data and the binary response of the outcome variable.

1.9.1 Data analysis

1.9.2 Descriptive analysis

Frequency and percentage were reported for categorical variables and continuous explanatory variables. In addition cross tabulation was showed the proportion of different categories of each characteristic with respect LBW.

1.9.3 Multilevel Analysis

Bivariate Multilevel Logistic Regression Analysis

Bivariate MLRA was employed to explore association between dependent variable and a wide range of independent variables. Variables with p-value ≤ 0.25 entered multivariable logistic regression which controls the undesirable effects of confounding variables [18].

Multivariable Multilevel Logistic Regression Analysis

Multilevel Logistic Regression Model was fitted to examine the individual and Area level factors that are associated with low birth weight at p-value of ≤ 0.25 during the Bivariate Multilevel Logistic Regression Analysis. Variables with p-value of less than 0.05 were considered as significant predictors. The result was presented with odds ratio (AOR) and 95% confidence interval (CI).

Model specification

In this multilevel analysis it has set up of two level models. The level one individual variables and second level is the Area level. The analytical strategy in the case of multilevel analysis consists of **four models**.

The first model which is usually called the “empty” or “null model” is fitted without explanatory variables. In other words, it contained no covariates, but decomposes the total variance in to individual and Area components. The empty model is used to determine whether the overall difference between communities and individual on LBW were significant.

$$Y = \ln [P_{ij}/1-P_{ij}] = b_{0j} + u_{0j} \dots\dots\dots (1)$$

In the above equation, P_{ij} is probability of LBW, b_{0j} is the overall regression intercept when all predictors were adjusted to zero and u_{0j} is the residuals at the Area level.

The second model referred to as the “individual model” included individual-level characteristics. This is to allow the assessment of the association between the outcome variable and individual level characteristics. The model containing the individual level variables is used to determine whether the variation across communities could be explained by the characteristics of the individual residing within that Area or not.

$$Y_{ij} = \beta_{0j} + \beta_1 X_{1ij} + \dots + \beta_n X_{nij} + u_{0j} + e_{ij} \dots \dots \dots 2$$

In the above model: β_{0j} is the intercept, β_1 is the regression coefficient (regression slope) for the explanatory variable, X_{1ij} is number of individual’s level factors and e_{ij} is the usual (random error term).

A third model contains only the Area level characteristics to allow the assessment of the impact of the Area level variables on the outcome variables.

$$Y_{ij} = \beta_{0j} + \beta_1 Z_1 + \dots + \beta_{nj} Z_{nj} + e_{ij} + u_{nj} \dots \dots \dots 3$$

Each cluster has different intercept β_{0j} , slope coefficients β_1 , Z_{nj} is number of Area level factors, and u_{nj} random residual error terms at the cluster level.

Lastly a fourth model generated which is called “final model”. This includes explanatory variables at both the individual and Area level simultaneously. The final model is used to test for the independent effect of Area contextual variables above and over the individual variables. The simultaneous inclusion of both individual and Area level predictors in the multilevel logistic regression model permits: (1) the examination of Area effects after individual level confounders have been controlled for: (2) the examination of individual level characteristics as modifiers of the Area effect (and vice versa); and (3) the simultaneous examination of within and between Area variability in outcomes, and of the extent to which between Area variation is explained by individual and Area level characteristics.

The formula for the final models expressed as:

$$\text{Log} [P_{ij} / (1-P_{ij})] = b_{0j} + b_1 X_{1ij} + b_1 Z_{1j} + \dots + u_{0j} + e_{ij}$$

- P_{ij} is the probability of LBW i^{th} birth weight in the j^{th} Area
- b_{0j} is the log odd of the intercept
- $b_1, b_{\dots}, b_{nj}$ are the regression coefficient estimate the data
- $X_{1ij}, \dots, X_{nij}$ are the covariates (independent variables) which may be defined at the individual level
- $Z_{1j}, \dots, Z_{nj}$ are the covariates (Area variables) which may be defined at the Area level
- u_{0j} are random error at the Area level

- e_{ij} are random error at the individual level

Parameter Estimation Method

The parameters that have to be estimated are the fixed coefficients b_0 , b_1 , etc and the random parameters $s^2_{u_0}$.

In the multilevel model, fixed effects (measure of association) refer to the individual and Area covariates and expressed as Adjusted Odds Ratio (AOR) and 95% confidence interval. The random effects are the measure of variation with LBW across communities. The ratio of the variance at the Area level to the total variance is referred to as the intra-class correlation coefficient (ICC). The precision is measured by the standard error (SE) of the independent variables [19]. The result of random effects (which are the measure of variation) are expressed as Variance Partition Coefficient (VPC) (which in this study is equal to ICC), and proportional change in variance (PCV). As a result of the dichotomous nature of the outcome variable in the study, the VPC calculated based on the linear threshold model method which converts the individual level variance from the probability scale to the logistic scale, on which the Area level variance is expressed [20]. In the other words, by using the linear threshold model, the unobserved individual outcome variable follows a logistic distribution with individual level variance s^2_e equal to $p^2/3$ ($=3.29$). In this case, the VPC corresponds to the ICC, which is a measure of general clustering of individual outcome of interest in the communities.

The ICC is calculated as: $ICC = (s^2_u / (s^2_u + p^2/3))$

ICC is the proportion of Area variance out of the total variance (Area plus individual variance) s^2_u is the variance of the Area level, $p^2/3 = 3.29$ and represent the fixed individual variance.

Area differences with LBW may be attributable to contextual influences or differences in individual composition of communities (including unobserved individual characteristics) [47]. In view of this, while adjusting for the individual characteristics in the multilevel model, same part of the compositional differences were taken in to consideration to explain some of the Area difference observed in the empty model. Thus the equation for the proportional change in Area variance is:

$$PCV1 = (VN1 - VN2) / VN1$$

$VN1$ – is the Area variance in the empty model and $VN2$ is the Area variance in the model including either individual level characteristics or Area level characteristics or both individual and Area level characteristics [20].

The Wald test was used to test the null hypothesis that a parameter value is zero or that a group of parameters are jointly zero. The latter case applies when testing the significance of categorical variable. Linear functions of parameters can also be tested. If the null hypothesis is true, the test statistic is distributed as approximately χ^2 with r degrees of freedom, where r is the number of functions that are

being tested [21]. Hence, the significance of freedom variation at each level will be tested with the Wald test, and p values <0.05 were considered to be significant to reject the null hypothesis.

Model diagnostics

Multi-collinearity diagnostic evaluation was done using variance inflation factor (VIF), and thus value of VIF greater than 10, gives evidence of multi-collinearity. Interaction effects were assessed between individual and Area level explanatory variables [22].

Model fit statistic

Receiver operating characteristics (ROC) curve was used to assess general accuracy of the model to the data set using the area under receiver operating characteristics. ROC curve is a commonly used measure for summarizing the discriminatory ability of a binary prediction model.

Relative goodness-of-fit tests were conducted using both Akaike's information criteria (AIC) for each of the models and compared. In general, it might be best to use AIC together in model selection. This was done because AIC was the appropriate selection in multilevel analysis than other methods. Since multilevel data have a different sample size at different levels and compared to AIC [23].

Discussion

This study used multilevel logistic regression analysis to address these two level factors. Accordingly the individual and Area level factors such as the type of birth, gestational age at birth, anemia status, birth order, mother's educational level and region have a significant association with low birth weight.

Child whose mother's with multiple births had high odds of LBW compared to those with single birth. Study done in Nigeria also shows multiple birth children were found to be significantly positively associated with LBW [24]. Similarly, a multilevel analysis conducted in Ethiopia also showed that multiple-birth children are 1.964 times more likely to have a LBW than single births [25]. Twin gestations are commonly associated with delivery of LBW infants because sometimes leading to complications like anemia, high blood pressure and early labor [26].

Gestational age at birth was also found to be an important factor in determining low birth weight in this study. Preterm gestational ages at birth had higher odds of low birth weight compared with full term birth. Studies conducted in Tanzania, Ghana and Uganda also shows gestational age less than 37 weeks were found to be significantly positively associated with LBW [27-29]. This is also consistent with study conducted in Gondar, Ethiopia [30].

This study found that anemic mothers were positively associated with low birth weight. Similar studies conducted in northern Tanzania and Ethiopia also shows mothers who are anemic are significantly vulnerable to small birth size than those who had not anemic mothers[25, 31]. Some studies have

demonstrated a strong association between low hemoglobin before delivery and LBW babies. Maternal anemia is commonly considered a risk factor for LBW babies [32].

Even though having a birth order of 6 and above showed odds of low birth weight compared to lower birth order (one up to three), other study conducted in Malawi revealed that birth order of 2–3 and 4–5 are associated with lower risk of being small at birth compared to those with first birth order [33]. But another studies conducted in 2014 in Ethiopia birth order of the child appeared to have no significant effect on determining the baby's size at birth [25]. During pregnancy, structural changes take place in the uterine spiral arteries, increasing blood flow with beneficial effects for fetal growth [34].

Maternal educational level appeared to be a very important determinant of the LBW in this study. The risk of LBW is significantly higher for children whose mothers have no education than children whose mothers have primary or secondary/higher level of education. Other studies done in Northern Tanzania, Nigeria, Kenya and Malawi implied that mothers' education has an association with delivery of LBW. Children of higher education mothers have reduced chance of being small at birth than children of mothers with no education [24, 27, 33, 35]. Previous single level studies done in Ethiopia showed a consistent result with this study [25, 36]. Low educated mothers are more frequently malnourished, have unhealthy habits, chronic diseases and inadequate prenatal care and this in turn related with mothers delivering small birth size infants [25].

Concerning the regional disparity children born in Afar, Amhara, Oromiya and Southern nations, nationalities and people were at a higher odds of LBW than children who born in Tigray. Similarly, previous study conducted In Ethiopia showed that children whose mothers reside in Afar, Amhara and Addis Ababa were more likely to be LBW as compared to those from the reference category (Dire Dawa) [25]. The observed high odds of LBW in these regions may be attributed to differences in nutrition, socio-economic status, health care services, and other cultural and life style differences among these regions.

Limitation: There were no cluster level weights for conducting weighted analysis and some variables may not be collected at the date of the event.

Conclusions

In this study the overall percentage of low birth weight in Ethiopia remained higher. The results of this study showed that multiple births, anemic mothers, birth order, not-educated mothers and preterm gestational age at birth were significant factors of low birth weight. Also, there exist considerable differences in baby's size at birth among regions. Hence, switching off/on the significant factors accordingly could reduce the risk of having low birth weight child.

Declarations

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1.2. Funding

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1.3. Ethics approval and consent to participate

Ethical approval was obtained from the research ethics review committee of Mekelle University, College of Health Sciences and the consent to participate was guaranteed by the ethics review committee and DHS program since an already existing data were used. However, confidentiality, and privacy of the data were assured.

1.4. Availability of data and materials

The dataset used in this study had not been released to public. According to relevant regulations, the data cannot be shared publicly. However, it can be obtained from the corresponding author up on reasonable request.

1.1 Authors' Contributions

Kibrom Taame Weldemariam, the principal investigator, has participated in designing and planning of the research; and in extraction and analysis of the data as well as in preparing the manuscript. Dr. Haftom Temesgen Abebe, Kebede Embaye Gezae and Tsegay Tekulu Tsegay have participated in planning, designing and analyzing the study. All authors read and approved the final manuscript.

1.5. Consent for publication

Not applicable

1.6. Competing interests

We all authors declared that we have no competing interests.

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Tables

Table 1a: Low birth weight distribution by individual level characteristics, EDHS 2016, Ethiopia, 2018

Characteristics	Frequency weighted	%	Normal BW	LBW	LBW %
Religion					
Orthodox	1006	47.7	871	135	13
Protestant	469	22.2	423	45	10
Muslim	620	29.4	522	98	16
Others	15	0.7	15	0	0
Age of mother (in year)					
15-24years	536	25.4	470	66	12
25-34 years	1208	57.2	1037	171	14
35 and above years	366	17.4	325	41	11.2
Member of at household					
1-5	1299	61.6	1119	180	14
6 and above	811	38.4	712	99	12
Sex of child					
Male	1077	51.1	960	117	11
Female	1033	48.9	871	161	16
Sex of HH head					
Male	1681	79.7	1476	205	12
Female	429	20.3	356	73	17
Maternal age at 1st birth					
Less than 18 years	557	26.4	479	79	14
18 and above years	1553	73.6	1353	200	13
Continued-					
Gestational age at birth					
Term	2047	97	1797	250	12
Preterm	63	3	34	29	46
Child is single or twin					
Single birth	2045	96.9	1792	252	12
Multiple birth	65	3.1	39	26	40
Marital status					
Not married	20	1	17	2	15
Married	2090	99	1814	276	13
Birth order					
1-3	1535	72.7	1342	193	13
4-5	318	15.1	252	66	21
6 and above	257	12.2	237	20	8
Smokes cigarettes					
No	2102	99.6	1824	277	13
Yes	8	0.4	7	1	13
Anemia level					
Anemic	399	19.9	333	66	17
Not anemic	1603	80.1	1397	207	13
Mothers educational level					
No education	609	28.9	498	111	18
Primary	803	38	714	88	11
Secondary/higher	698	33.1	619	79	11.3
Maternal occupation					
Not Employed	1253	59.4	1073	179	14
Employed	857	40.6	758	99	11

Husband education level					
No education	414	20.9	339	75	18
Primary	690	34.8	585	107	15
Secondary/higher	878	44.3	791	89	10
Husband occupation					
Not employed	904	45.6	747	157	17
Employed	1078	54.4	986	110	10
Nutritional status					
Underweight	289	14.2	233	56	19
Normal	1361	66.7	1176	185	14
Overweight/obese	390	19.1	355	35	9
wealth index					
Poor	363	17.2	306	57	16
Middle	294	13.9	243	51	17
Rich	1453	68.9	1283	170	12
Media exposure					
Not exposed	713	33.8	611	102	14
Exposed	1397	66.2	1220	177	13

Table 1b: Low birth weight distribution by Area level characteristics, EDHS 2016, Ethiopia, 2018

Characteristics	Frequency	%	Normal BW	LBW	%
Region					
Tigray	294	13.96	272	22	7
Afar	9	0.44	7	2	22
Amhara	288	13.64	224	64	22
Oromia	602	28.53	523	79	13
Somali	73	3.44	65	8	11
Benishangul	36	1.71	33	4	11
SNNPR	448	21.21	389	59	13
Gambela	12	0.57	11	1	8
Harari	13	0.63	12	1	8
Addis Ababa	306	14.49	271	35	11
Dire Dawa	29	1.36	26	3	10
Residence					
Urban	1026	48.63	914	112	11
Rural	1084	51.37	917	167	15
Area poverty					
Low	1424	67.47	1244	180	13
High	666	32.37	588	97	14
Area education					
Low	2001	94.83	1738	263	13
High	109	5.17	93	16	15
Area media exposure					
Low	1052	49.88	895	157	15
High	1058	50.12	937	121	11

Table 2: multivariable multilevel logistic regression analysis of individual and Area level factors associated of low birth weight 2016

Characteristics	Empty	Individual	Area	(Individual & Area)
	Model 1	Model 2 AOR (95% CI)	Model 3 AOR (95% CI)	Model 4 AOR (95% CI)
Child is single or twin				
Single birth		Ref		Ref
Multiple birth		3.25(1.670-6.355)		2.74*(1.450-5.184)
Birth order				
1-3		Ref		Ref
4-5		0.80(0.501-1.296)		0.87(0.570-1.326)
6 and above		0.40(0.205-0.801)		0.42*(0.242-0.752)
Gestational age				
Term		Ref		Ref
Preterm		6.69(3.361-13.35)		4.83*(2.644-8.830)
Anemia level				
Not anemic		Ref		Ref
Anemic		1.37(0.974-1.954)		1.49*(1.069-2.092)
Mothers educational level				
No education		Ref		Ref
Primary		0.61(0.396-0.941)		0.61*(0.418-0.896)
Secondary/higher		0.43(0.252-0.752)		0.39*(0.252-0.612)
Region				
Tigray			Ref	Ref
Afar			3.98*(1.509-10.499)	3.83*(1.463-10.044)
Amhara			3.74*(1.762-7.950)	3.50*(1.649-7.428)
Oromiya			2.23*(1.113-4.483)	2.19*(1.085-4.411)
Somali			1.93(0.957-3.891)	1.32(0.625-2.818)
Benishangul			1.13(0.54-6-2.371)	1.35(0.630-2.906)
SNNPR			1.92(0.999-3.707)	2.39*(1.236-4.644)
Gambela			1.79(0.890-3.602)	1.71(0.846-3.471)
Harari			0.74(0.340-1.629)	0.73(0.330-1.623)
Addis Ababa			1.96*(1.030-3.760)	1.73(0.935-3.203)
Dire Dawa			1.32(0.667-2.637)	1.23(0.627-2.437)
Random effect parameters				
Area level variance and (SE)	Empty 0.434 (0.1869)	Individual 0.300 (0.1801)	Area 0.247 (0.1572)	Individual and Area 0.155 (0.1576)
ICC (%)	11.7	8.4	7	4.5
PCV (%)	Ref	33.6	44	64.3
Model fit statistics AIC	1469.67	1340.07	1463.08	1333.33

Abbreviations

AIC Akaike's Information criterion

APH Ante-partum hemorrhage

CSA	Central Statistical Agency
DHS	Demographic health survey
EAs	Enumeration Areas
EDHS	Ethiopian demographic health survey
ICC	Intra class Correlation
IUGR	Intra-Uterine growth retardation
LBW	Low Birth Weight
MUAC	Mid-Upper Arm Circumference
NBW	Normal Birth Weight
NMR	Neonatal Mortality Rate
AOR	Adjusted Odds Ratio
SNNP	Southern Nation, Nationalities and People
VPC	Variance partition coefficient
WHO	World health organization

Figures

Figure 1

Figure 2: Graphical representation of sampling procedure for low birth weight in Ethiopia from EDHS 2016.

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