

MACROECONOMIC DETERMINANTS OF CHILD MORTALITY: AN INVESTIGATION USING A CROSS-SECTIONAL ANALYSIS OF 102 COUNTRIES

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KEYWORDS

Child mortality, GDP, well-being, Pérez-Moreno, Blanco-Arana, Bárcena-Martín.

ABSTRACT

Child mortality remains a critical global challenge, particularly in developing regions such as Africa and South Asia. Despite significant global progress, with child mortality rates dropping from 61% to 37% since 1990, approximately 5 million children under the age of five died in 2020 alone. This research investigates the key determinants of under-five child mortality using cross-sectional data from 102 countries across all income levels for the year 2019. Drawing on existing literature, the study examines the impact of health expenditure, food inflation, gender parity in literacy, aggregate income, and carbon emissions on child mortality. A multiple linear regression model was employed, and assumptions of the classical linear regression were tested and satisfied. The results indicate that increased female literacy and public health expenditure are significantly associated with reductions in child mortality, while food inflation is positively associated with higher mortality rates. In contrast, GDP per capita and CO₂ emissions were found to be statistically insignificant in the global context. These findings suggest that governments should prioritize female education and invest more in public health to reduce under-five mortality. Future studies are encouraged to incorporate more variables and broader datasets to further explore regional differences and causal mechanisms.

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Introduction

Child mortality is one of the most controversial topics in development economics. The severity of the issue can be seen in developing countries, especially African and South Asian

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countries. Although there has been a significant reduction (from 61% to 37%) in child mortality since 1990, the issue is still one of the urgent concerns of the globe. In 2020, the number of children who are under 5 years of age was 5 million. In other words, 13800 children died per day in 2020 (UNICEF, 2021). According to the World Health Organization (2018), improvement in the well-being of children is one of the purposes of the Sustainable Development Goals (SDGs). The SDG Goal 3.2.1 is to eliminate under-5 child mortality by 2030. In this regard, this research is conducted to find out the main determinants of child mortality around the globe. Thus, the aim of the research is to recommend policies to reduce child mortality.

Literature review

In the literature, some of the existing knowledge, which is about the impact of health expenditure, inflation, and literacy on child mortality, is mentioned.

Having used a dynamic panel data approach with world data, Ray and Linden (2019) concluded that public expenditure on health is a significant factor for child mortality and investigated that 10% increase in public expenditure on health will lead to the reduction of infant mortality by 2.31%, on average. Furthermore, it is revealed that there is a significant negative relationship between health expenditure and child mortality rates in low-, lower-middle- and upper-middle-income countries. However, the relation is insignificant in high-income countries (Dhrifi, 2019).

Volatility in food prices might cause to lessen the purchasing power parity of poor families and can lead to undernourishment and high child mortality. In the study of Lee et al. (2013), it is summarized that high and volatile food prices have a significant and negative effect on child mortality, especially in the least developed countries. Similarly, in the research of Lee et al. (2015), it is concluded that rising food prices have a significant effect on infant and child mortality in the least developing countries. The results of the study showed that 10% increase in food price inflation results 2.3% increase in infant mortality rate, on average.

The study of Pradhan et al. (2017) revealed that female literacy is more significant than male literacy in achieving a reduction in child mortality. Similarly, Latif, Ashraf, and Zeeshan's (2019) study in Pakistan presented that male literacy is an insignificant factor for child mortality, while female education is considered a significant factor.

After analyzing the crises of 1982-85 and 1995-96 in Mexico, Cutler et al. (2002) found that economic crises led to an increase in the mortality rate of children and elderly up to 9%. Having used the cross-sectional data of the least developed countries, GDP per capita is revealed to be a significant determinant for child mortality during downturns, while it is not significant during expansions (Pérez-Moreno, Blanco-Arana, and Bárcena-Martín, 2016).

In the research of Zakaria, Tariq, and Husnain (2019), carbon emission is found to be a significant factor in determining child death in South Asia and it is stated that 1% increase in carbon emission will increase child mortality by 1.254%. The study about energy use in Africa revealed that 1% increase in carbon emission will lead to 4% increase in child mortality, on average (Shobande, 2019).

To summarize, investigations of previous studies revealed that health expenditure, inflation, and carbon emission are significant factors for child mortality. However, it is ambiguous whether mother education is a more significant factor in the reduction in child mortality than father education. Finally, the effect of the aggregate income is subtle according to the business cycle.

Materials and Methods

Data overview

According to the empirical literature review, public health expenditure, food inflation, female and male literacy, aggregate income and carbon emission are found to be the essential determinants of child mortality and, thus, variables are chosen accordingly. All figures, which are used in this report, are obtained from the database of World Bank. Furthermore, cross-sectional data of 108 countries from all income divisions have been analyzed to determine the factors of child mortality in 2019². The detailed information about the variables is given in the table below. In addition, the raw data of all variables is available in Appendix 1.

Summary of variables

Variables (Abbreviations)	Variable definitions	Measurement units
Under-five mortality rate (U5MR)	Probability (per 1000) of dying a child before the age of 5	Percentage
Domestic general government health expenditure per capita, PPP (HE)	Public expenditure on health per capita at purchasing power parity	USD
Inflation (Consumer Price Index) (INF)	Annual percentage change in the cost of acquiring a basket of goods	Percentage
Gender parity index in literacy rate (GPILR)	The ratio of literate female to literate male	Percentage
GDP per capita (GDPpercapita)	GDP divided by mid-year population.	USD
CO2 emissions (CO2)	Carbon dioxide emission	Metric tons per capita

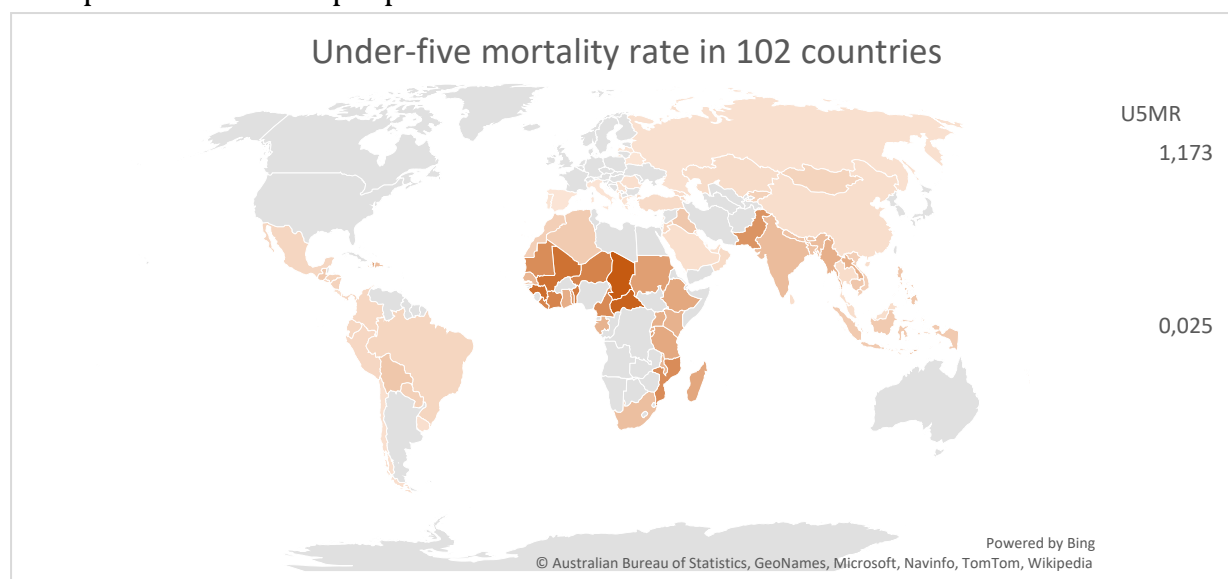
Table 2. Descriptive statistics of the variables

Variable	Mean	Std. Dev.	Min	Max
U5MR	30.21373	27.19534	2.5	117.3
INF	4.431791	7.130132	-2.814698	63.29251

² Exemption: The values of the gender parity index in literacy rate (GPILR) are taken as a 5-year average due to the limitation in the data availability.

GPILR	96.53506	9.713281	52.5805	104.718
lnHE	5.328456	1.611706	1.908986	7.953927
CO2	3.664968	5.247188	.0527946	32.41564
lnGDPpercapita	11.81064	2.556589	6.232973	18.22755

As shown in Table 1, the under-five mortality rate (U5MR) has the highest standard deviation. Furthermore, the average U5MR is 30.21 per 1000 children and the maximum and minimum probability is 117.3 (Chad) and 2.5 (Singapore), respectively (see the map below). The same interpretation can be proposed for all other variables.



As HE and GDP per capita have high values in USD, it might cause heteroscedasticity and non-normal distribution of error term. Therefore, it is decided to use ln of both variables.

Estimation

The model of the analysis is as follows: $U5MR = F(HE, INF, GPILR, GDPpercapita, CO2)$

Thus, the population regression function is generalized:

$$U5MR_i = \beta_0 + \beta_1 \cdot INF_i + \beta_2 \cdot GPILR_i + \beta_3 \cdot lnHE_i + \beta_4 \cdot CO2_i + \beta_5 \cdot lnGDPpercapita_i + \varepsilon_i$$

In this equation, β_0 (constant) is an average child mortality excluding all the variables in the model; ε_i is an error term; $\beta_1, \beta_2, \beta_3, \beta_4, \beta_5$ are partial regression coefficients.

Table 1. Correlation matrix

	U5MR	INF	GPILR	lnHE	CO2	lnGDPpercapita
U5MR	1.0000					
INF	0.1779	1.0000				
GPILR	-0.8199	-0.0180	1.0000			
lnHE	-0.8511	-0.1962	0.5622	1.0000		
CO2	-0.4410	-0.1385	0.2568	0.5915	1.0000	
lnGDPpercapita	0.1030	-0.1914	-0.0722	-0.1470	-0.0616	1.0000

The table above shows correlations among variables. It should be mentioned that there is no

perfect colinear relation between regressors. The highest one is 59.15%, which is between CO2 emissions and Health Expenditure. According to correlation, it is estimated that an increase in CO2 emissions, health expenditure, and female literacy can lead to a decrease in child mortality. On the other hand, an increase in inflation might lead to an increase in child mortality. Lastly, the effect of aggregate income is subtle. All these results are theoretically consistent with the findings in the literature review.

Estimated results³

Starting with the overall significance of the model, the following hypothesis is set:

$$H_0: \beta_1 = \beta_2 = \beta_3 = \beta_4 = \beta_5 = 0$$

H_a : Not all slope coefficients are equal to zero

Decision making: Reject H_0 , if $F > F_\alpha$ or $\text{Prob}(F) < \alpha$

As $\text{Prob}(F)$ is equal to 0.0000, it can be concluded that at least one of the predictors is significant in the model.

Table 2. Overall significance of the model

SOURCE	SS	DF	MS
MODEL	67172.9635	5	13434.5927
RESIDUAL	7525.29731	96	78.3885136
TOTAL	74698.2608	101	739.58674

Number of obs	102
F(5, 96)	171.38
Prob > F	0.0000
R-squared	0.8993
Adj R-squared	0.8940
Root MSE	8.8537

The R-squared is 89.98%, which represents that 89.98% of the variation in the child mortality can be expressed with the regressors: domestic general government health expenditure, inflation, literacy rate (GPI), aggregate income and CO2 emissions. Furthermore, the difference between Adjusted R-squared and R-squared is low (0.53%), which can represent that the predictors are selected correctly. To sum up, it can be concluded that the chosen model fits the sample data.

Table 3. Overall significance of the model

SOURCE	SS	DF	MS
MODEL	67172.9635	5	13434.5927
RESIDUAL	7525.29731	96	78.3885136
TOTAL	74698.2608	101	739.58674

³ Ordinary least squares (OLS) regression is used to find out the main factors of child mortality.

Number of obs	102
F(5, 96)	171.38
Prob > F	0.0000
R-squared	0.8993
Adj R-squared	0.8940
Root MSE	8.8537

Regarding the significance of the individual regression coefficients, the following hypothesis is set:

$$H_0: \beta_n = 0$$

$$H_a: \beta_n \neq 0$$

In the equation above, n identifies the regression coefficients. $n \in [1; 5]$

Decision making: Reject H_0 , if $|t| > t_{\frac{\alpha}{2}}$ or $\text{Prob}(F) < \alpha$

Table 4. Estimated results

U5MR	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
INF*	0.226514	0.130332	1.74	0.085	-0.03219	0.48522
GPILR***	-1.40778	0.111181	-12.66	0.000	-1.62847	-1.18709
lnHE***	-9.80232	0.822335	-11.92	0.000	-11.4347	-8.17
CO2	0.205641	0.209717	0.98	0.329	-0.21064	0.621926
lnGDPpercapita	-0.05141	0.358175	-0.14	0.886	-0.76238	0.659565
_cons	217.1948	10.23685	21.22	0.000	196.8748	237.5148

***(*) show that value is statistically significant at 1% (10%) level

From the Table, it is revealed that inflation in a basket of foods is significant in 10% significance level, which is consistent with the research of Lee et al. (2013). The estimated result implies that 1% point increase in inflation, ceteris paribus, tends to increase the under-five mortality rate by 0.22% point, on average. In fact, the estimated result of Lee et al. (2015) is approximately the same as the result above. In other words, an increase in the prices of commodities drops purchasing power of people and some people cannot afford to buy necessities. Consequently, undernourishment and child mortality will rise.

The gender parity index of literacy rate is found to be significant at 1% significance level (see Table 4. The interpretation of the result is that 1% point increase in the ratio of literate female to literate male, holding all other variables constant, tends to decrease child mortality by 1.4% points. This result represents that the increase in female literacy has a more significant effect on child mortality than male literacy, which is in line with the conclusions of Pradhan et al. (2017) and Latif, Ashraf, and Zeeshan(2019). Generally, women are engaged to take care of a child in most countries. Literate mothers are efficacious to control the nutrition and health of a child.

Similarly, domestic general government health expenditure per capita (HE) is significant at 1% significance level (see Table 4). The estimated result implies that 1% increase in HE, *ceteris paribus*, will lead to a decrease in the under-five mortality rate by 0.098% points. The finding is consistent with the studies of Ray and Linden (2019) and Dhrifi (2019). An increase in health expenditure improves health atmosphere in a country and reduces child mortality. Though the signs of coefficients are logically correct and corroborate the researches of Cutler et al. (2002) and Shobande (2019), the results showed that CO₂ emissions and aggregate income are insignificant factors for child mortality. According to Pérez-Moreno, Blanco-Arana and Bárcena-Martín (2016), aggregate income becomes insignificant factor for child mortality during expansions. In cross-sectional data, it is ambiguous to determine the stage of the business cycle of a country. Thus, it might be the reason why GDP per capita becomes insignificant in the estimated results. Regarding CO₂ emissions, Zakaria, Tariq and Husnain (2019) and Shobande, (2019) analyzed only the least developed countries. However, the sample data of this research consists of countries from all income divisions. Thus, the carbon dioxide emissions might be insignificant due to generalized data.

Assumption testing

The assumptions of classical linear regression model are examined to identify whether the regression output is valid or not.

1. Zero covariance between the error and regressors

From the covariance matrix, it can be concluded that there is roughly zero correlation between the error and regressors. This means that the regressors are independent from the error term.

2. Zero mean value of disturbance

The mean value of the disturbance term is approximately equal to zero ($\varepsilon_i = -1.61 \cdot 10^{-9}$). This means that there is no specification bias in the model or the model is correctly specified.

3. Homoscedasticity

The Breusch-Pagan test and Cameron & Trivedi's decomposition of IM-test is used to test homoscedasticity (see Table 6). The following hypothesis is set:

H_0 : Constant variance

H_a : Non – constant variance

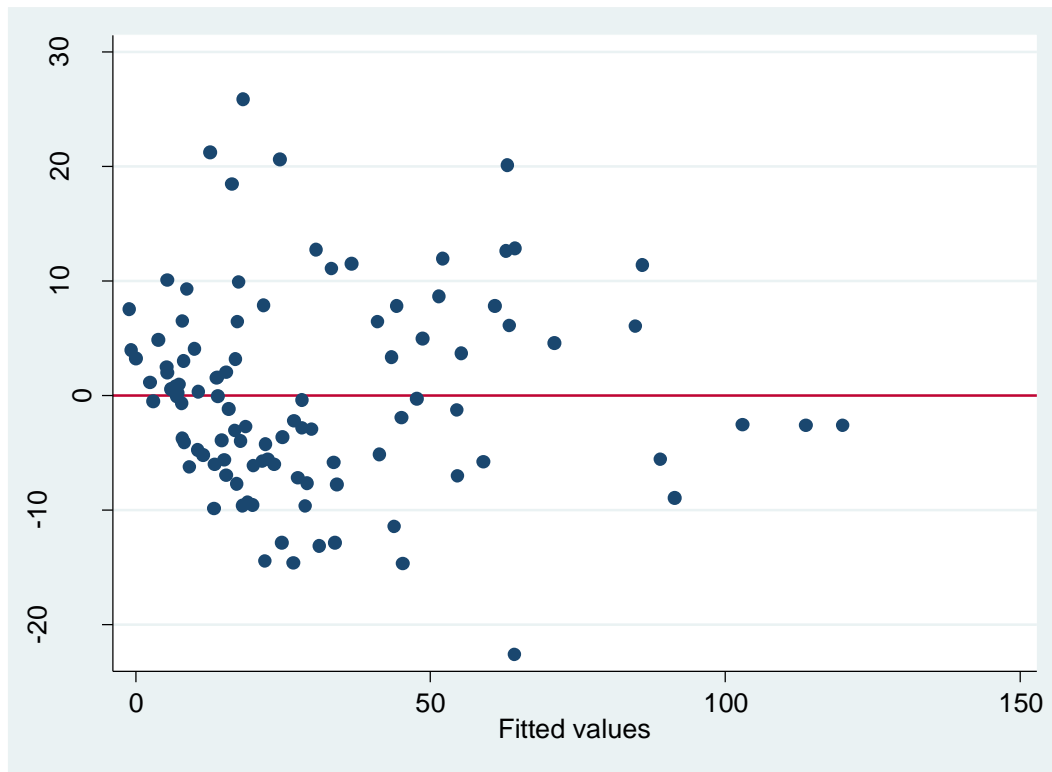
Table 5. Cameron & Trivedi's decomposition of IM-test

Source	Chi2	df	P
Heteroscedasticity	16.32	20	0.6966
Skewness	4.42	5	0.4911
Kurtosis	1.06	1	0.3030
Total	21.80	26	0.6997

Decision making: Reject H_0 if Prob < α .

Prob(chi2) = 0.2715 (Breusch-Pagan) and $P = 0.696$, so we fail to reject H_0 . Thus, the error term has a constant variance. Furthermore, it is visible from the Figure 1 (Appendix 4) that squared residuals do not follow any residual.

Figure 1. Estimated squared residuals



4. No autocorrelation

As the sample data is cross-sectional, we cannot check for serial autocorrelation.

5. No perfect multicollinearity

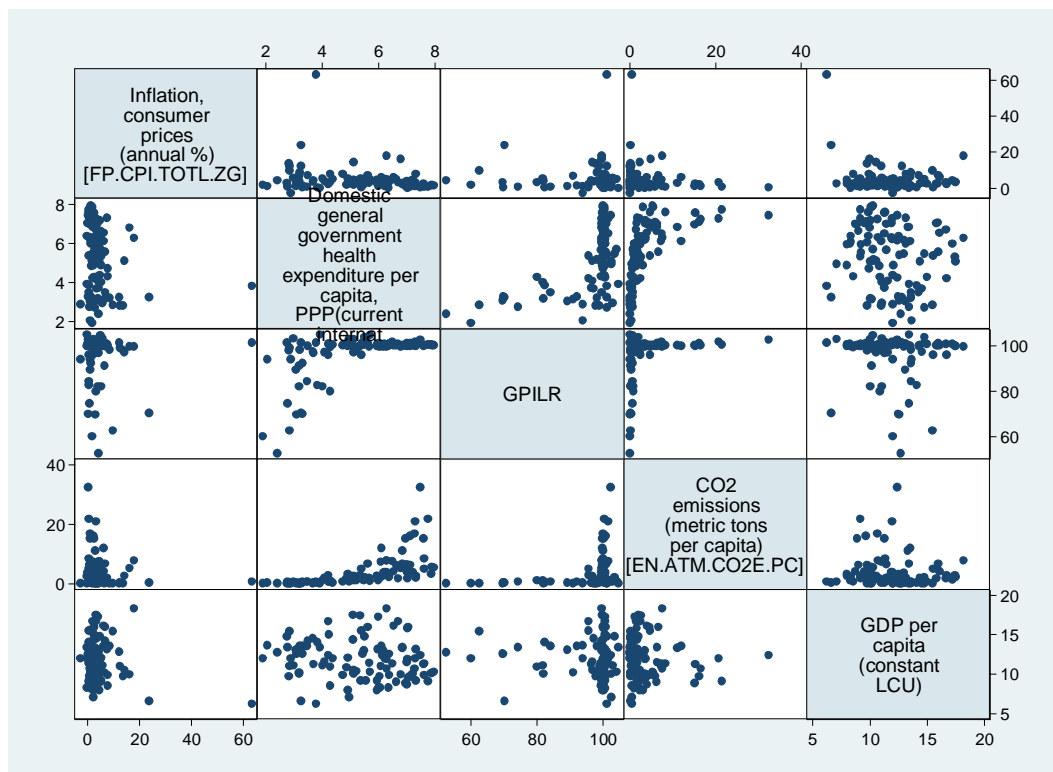
Correlation matrix showed that the highest correlation between variables are 0.5915, which is not perfect collinear (See Table 2, App. 2).

Table 6. Variance-inflating factor and Tolerance

Variable	VIF	1/VIF
lnHE	2.26	0.441837
CO2	1.56	0.640930
GPILR	1.50	0.665488
INF	1.11	0.898745
lnGDPpercapita	1.08	0.925589
Mean VIF	1.50	

From Table 7, it is known that variance-inflation factor is less than 5 for all regressors, which indicates no perfect multicollinearity. Finally, it is obvious from the scatterplot (Figure 2, App. 3) that there is no perfect multicollinearity between regressors.

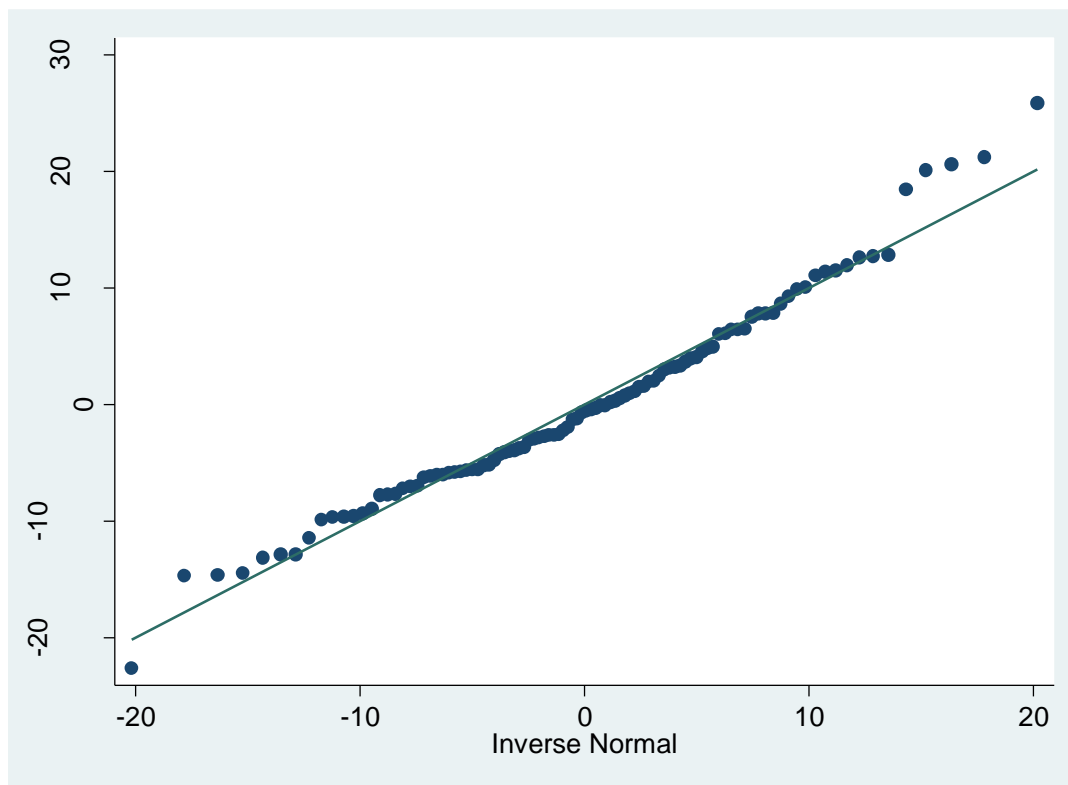
Figure 2. Scatterplot of independent variables



6. Normality

Figure 3 (App.3) shows that the residuals are normally distributed. Moreover, the result of D'Agostino test showed that p-value (0.0765) > 0.05 , which means we fail to reject null hypothesis of normality of residuals.

Figure 3. Q-Q probability plot



Conclusion and Limitations

This study examined child mortality in 102 countries from all income divisions. After checking for OLS assumptions, it can be concluded that the OLS estimators as the best linear unbiased estimators (BLUE), and the model fits the sample data. To sum up, female literacy and domestic health expenditure are found to be fundamental determinants of under-five mortality in any country. In other words, literate mothers and high health expenditures will increase the well-being of children. Furthermore, the main limitations of the research are time constraints and unavailability of data for literacy rate.

To decrease under-five mortality, governments should:

- Concentrate more on female literacy
- Rise government health expenditure

Future researches should consider more countries and add more control variables to increase the precision of the research and fill the research gap in the study of child mortality.

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Appendices

Appendix 1: Raw data

CountryName	U5MR (%)	INF(%)	CO2(metric per tons)	lnGDPpercapita	lnHE	GPILR (%)
Senegal	41.7	0.461	0.622	13.587	3.479	84.025
Nepal	30.6	4.061	0.428	11.265	3.698	96.649
Armenia	12.1	2.520	1.880	14.408	5.080	100.053
Sri Lanka	7.5	2.135	0.998	13.008	5.477	100.546
Sao Tome and Principe	18.1	7.857	0.663	9.899	4.732	100.145
Egypt, Arab Rep.	20.9	14.401	2.502	10.522	5.109	97.013
Tonga	11.9	5.032	1.841	9.199	5.349	100.192
Bangladesh	32.4	5.544	0.513	11.057	2.964	103.062
Belarus	3.4	4.872	6.284	9.044	6.629	100.053
Kyrgyz Republic	19.2	1.543	1.740	10.823	4.851	100.120
China	8.5	2.075	7.352	11.000	6.092	100.000
Kazakhstan	10.3	6.019	12.062	13.559	6.104	99.911
Georgia	9.7	2.615	2.539	9.240	5.944	99.842
Liberia	82.5	23.564	0.274	6.594	3.253	70.260
Vanuatu	26.3	2.331	0.615	12.361	4.228	100.642
Albania	9.5	2.028	1.940	13.175	5.973	100.582
Azerbaijan	21.5	2.269	3.221	8.001	4.906	99.952
Morocco	20.3	1.804	1.851	10.197	5.112	99.347
Congo, Rep.	47.6	1.153	0.614	13.617	3.301	92.267
Malaysia	8.4	0.885	7.600	10.675	6.297	100.381
Montenegro	2.9	2.611	4.050	8.736	6.998	99.738
Iran, Islamic Rep.	13.9	18.014	7.693	18.228	6.285	99.651
Honduras	17.4	4.347	1.019	10.018	5.103	102.576
Lebanon	7.4	6.077	4.040	16.057	6.540	100.177
Cambodia	27.7	2.459	0.687	14.995	4.254	100.766
Gambia, The	53.1	6.521	0.250	10.147	3.165	91.099
Cabo Verde	15.8	1.256	1.140	12.587	5.419	101.160
Thailand	9.4	1.064	3.714	11.945	6.212	100.605
Mongolia	16.8	6.812	6.725	15.937	5.598	100.898
Niger	83.5	2.968	0.102	12.624	3.076	69.630
Russian Federation	6.3	2.878	11.127	13.322	6.839	100.099

India	36.3	3.945	1.800	11.548	4.075	96.964
Serbia	5.8	1.960	6.522	13.432	6.829	100.005
Samoa	17.8	4.197	1.632	9.227	5.546	100.598
Latvia	4.2	2.534	3.959	9.548	7.034	100.080
Peru	13.8	1.509	1.697	9.724	6.051	99.648
Turkey	10.7	16.332	5.015	9.968	6.793	99.608
Greece	4.2	0.626	6.083	9.730	7.106	99.809
Vietnam	21.3	3.540	2.699	17.415	5.326	99.827
El Salvador	13.8	1.090	1.061	8.261	5.991	100.624
Iraq	27	0.367	4.895	15.516	5.374	95.795
Guatemala	25.4	3.752	1.114	10.318	5.198	98.413
Jordan	16	4.462	2.479	7.992	5.957	100.299
Chad	117.3	4.275	0.069	12.732	2.403	52.580
Central African Republic	111.1	1.612	0.071	11.975	1.910	60.050
Guinea	100.4	9.826	0.251	15.453	2.843	62.420
Indonesia	24.6	3.198	2.178	17.478	5.091	99.961
Malawi	43.1	12.420	0.087	12.837	3.244	101.218
Ethiopia	53.2	13.831	0.149	9.729	2.810	97.894
Mexico	14.6	4.899	3.741	11.897	6.294	100.139
Chile	7.2	2.435	4.625	15.917	7.049	100.090
Singapore	2.5	0.439	8.399	11.338	7.610	100.041
Philippines	27.9	5.212	1.334	12.051	4.898	100.686
Uganda	47.4	2.624	0.143	14.809	2.738	101.249
Ecuador	13.9	-0.224	2.314	8.344	6.370	100.008
United Arab Emirates	7	3.069	20.797	11.930	7.291	101.612
Romania	7.4	4.625	3.845	10.038	7.195	100.020
Oman	11.1	0.881	15.192	8.894	6.960	100.228
Qatar	6.6	0.256	32.416	12.388	7.468	102.327
Maldives	7.6	-0.133	3.704	11.848	7.015	100.704
Costa Rica	8.4	2.221	1.652	15.768	7.010	100.191
Portugal	3.6	0.994	4.841	9.851	7.606	100.075
Brazil	15.2	3.665	2.042	9.882	6.382	100.633
Mauritius	15.4	3.216	3.264	12.554	6.373	100.619
Uruguay	7.3	7.607	1.890	13.144	7.308	100.617
Nicaragua	17.3	4.947	0.806	10.234	5.704	104.230
Saudi Arabia	7.7	2.458	15.269	11.265	7.586	99.959
Brunei	11.2	1.025	16.645	10.666	7.258	100.155

Darussalam						
Paraguay	20.2	3.976	1.210	17.219	5.985	100.732
Spain	3.3	1.675	5.520	10.126	7.887	100.033
Myanmar	46.8	6.872	0.605	14.257	3.496	99.733
Burundi	59	-2.815	0.053	11.960	2.882	94.051
Italy	3.2	1.137	5.376	10.257	7.954	100.030
Colombia	14.1	3.240	1.601	16.660	6.699	100.648
Mozambique	75.6	3.911	0.225	10.016	3.175	82.011
Kuwait	8.7	0.543	21.623	9.158	7.752	100.476
Madagascar	53.7	8.594	0.128	13.590	3.206	98.266
Benin	90.9	0.645	0.689	13.434	2.766	74.459
Pakistan	69.5	5.078	0.982	11.033	4.017	81.841
Lao PDR	47.5	2.040	2.661	16.679	4.214	95.865
Algeria	23.8	4.270	3.592	13.015	6.182	99.650
Seychelles	14.5	3.703	6.407	11.440	6.990	101.044
Malta	6.5	1.158	3.198	10.103	7.869	100.506
Tanzania	52.1	3.494	0.206	14.546	3.678	97.274
Togo	68.8	0.928	0.286	13.115	3.068	89.313
Bhutan	29.6	2.723	1.829	11.410	5.680	99.594
Sudan	60.1	63.293	0.483	6.233	3.796	101.334
Palau	18	2.110	16.191	9.655	7.112	100.933
Bolivia	27.4	2.272	2.000	8.353	6.061	100.010
Panama	15.4	0.762	2.428	9.211	7.306	99.757
Kenya	44.3	4.690	0.358	11.996	4.381	100.556
Mali	97.4	0.300	0.295	12.511	3.293	69.910
Ghana	48.1	7.809	0.541	8.559	4.294	99.404
Haiti	64.1	12.481	0.299	10.988	2.839	99.138
Mauritania	75.5	3.049	0.908	10.892	4.284	80.016
Rwanda	43.4	-0.311	0.088	13.429	3.915	104.718
Cameroon	77.3	1.074	0.342	13.659	2.050	93.947
Dominican Republic	34.8	3.564	2.364	12.369	6.133	100.438
Cote d'Ivoire	83.2	0.359	0.395	14.100	3.845	82.324
Timor-Leste	45.1	2.294	0.505	7.132	4.952	102.590
South Africa	33.9	4.505	7.497	11.278	6.487	101.541
Gabon	44.1	4.749	2.175	14.775	5.493	103.617

Table 7. Estimated results

U5MR	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
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INF*	0.226514	0.130332	1.74	0.085	-0.03219	0.48522
GPILR***	-1.40778	0.111181	-12.66	0.000	-1.62847	-1.18709
lnHE***	-9.80232	0.822335	-11.92	0.000	-11.4347	-8.17
CO2	0.205641	0.209717	0.98	0.329	-0.21064	0.621926
lnGDPpercapita	-0.05141	0.358175	-0.14	0.886	-0.76238	0.659565
_cons	217.1948	10.23685	21.22	0.000	196.8748	237.5148

***(*) show that value is statistically significant at 1% (10%) level

Table 8. Covariance matrix

	INF	lnHE	GPILR	CO2	lnGDPpercapita	uhat
INF	50.8388					
lnHE	- 2.25502	2.5976				
GPILR	- 1.24832	8.80113	94.3478			
CO2	- 5.18046	5.00254	13.0863	27.533		
lnGDPpercapita	- 3.48949	- .605547	- 1.79182	-.826098	6.53615	
uhat	2.4e-08	4.0e-08	-1.3e-08	1.7e-07	2.9e-08	74.5079