

# Epidemiologic Modeling of COVID-19 in Amhara Region, Ethiopia: Using The Best and Worst Scenarios Experienced in The World

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## Research Article

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# Abstract

**Background:** The ongoing pandemic of the novel coronavirus (SARS-CoV-2) is currently the first public health agenda. Citizens of resource-poor countries like Ethiopia could be most affected by the pandemic unless aggressive plans are implemented. Despite the prevention and control efforts being made at the national and regional levels, there is no evidence about the worst and best scenarios in the number of COVID-19 cases expected in the Amhara Regional. Besides, there were no efforts to develop predicting models of future values in COVID-19 anticipated cases expected based on the current global trend.

**Objective:** Forecast the latest epidemic situations by estimating and predicting cases and deaths of COVID-19 in the Amhara region.

**Methods:** This study employed models by using the expected number of COVID-19 cases and deaths in Amhara National Regional State using the worst scenarios faced in the world. Data were extracted using a checklist. Existing data on morbidity and mortality by age and sex were collected through a literature search strategy including PubMed/Medline and the Google scholar database. Besides, the Worldometer of the novel COVID-19 was used as a source of data. This study used MS-Excel spreadsheet program for data extraction. COVID-19 modeling was made using Susceptible- Infected-Recovered (SIR), Susceptible-Exposed Infected-Recovered (SEIR), and WHO DisMod II software.

**Result:** Without any pharmacological intervention, based on the SEIR model, the expected number of peak COVID-19 cases could reach about 3 million cases occurring on July 29<sup>th</sup>, 2020. However, with an intervention that could reduce the transmission by 30%, would push the peak period to 22 days (20<sup>th</sup> August). Using the Susceptible–Infected–Recovered model without non-pharmacologic interventions, the expected number of peak COVID-19 cases is 5,947,685, occurring on July 2, 2020. The peak number of Hospitalization is 820, 781, in need of ICU 279,541 and deaths 53,529. With the DisModII modeling, outputs showed that males are more affected than females. Accordingly, the total incident cases in the region were estimated to be 837,348 (474, 809 male and 362,540 female). The expected number of total deaths was also estimated at 44,247 (30, 176 male and 14, 071 female). Also, the middle age group is more infected but less at risk of death. The recovery rate decreases drastically as age increase.

**Conclusion:** The discrepancies in findings from the three models used in this study could be driven mainly by data limitation in the Ethiopia context. However, our results from the different models highlight the need for more intervention on non-pharmacological interventions. Besides, highly influential parameters urge further study for better effective evidence-based decision making.

## Background

Coronavirus is one of the large families of RNA viruses. It causes illness in animals and humans. Originated in China, Wuhan in December 2019 rapidly spread to the globe. Subsequently, the World Health Organization (WHO) declared the outbreak as a public health emergency of

international concern on January 30, 2020, and on March 11, 2020, as a global pandemic(1). Compared with SARS and MERS, COVID-19 has spread more rapidly due increased globalization and the epidemic's focus(2).

Based on the initial report on the first 41 cases of the COVID-19 outbreak, most (73%) of patients infected with COVID-19 were males, with less than half (32%) possessing underlying comorbidities of diabetes, hypertension, and cardiovascular disease. It also reported that the median age of cases was 49 years. Of the initial 41 patients infected, 27 (66%) had been directly exposed to the Huanan seafood market, and the Case Fatality Rate was nearly 2%(3).

A study on the COVID-19 outbreak on the Diamond Princess Cruise ship using the SEIR model with an assumption of relatively homogenous mixing of all people onboard, they estimated the basic reproduction rate ( $R_0$ ) to 14.8 based on calibrating the predicted cumulative number of infections to the observed cumulative number of infections among all people onboard. Accordingly, the daily reproduction rate ( $\beta$ ) was estimated to be 1.48. For the model, considering the infectious period ( $i$ ) of COVID-19 to be ten days, in the situation of no removal, the incubation period or the latent period ( $L$ ), was estimated to be approximately five days (ranging from 2 to 14 days) (4).

A study on the real-time forecasts of the COVID-19 epidemic in China from February 5<sup>th</sup> to February 24<sup>th</sup>, 2020, the latest 5-day forecasts, generated on February 9, estimate an average of 34,509 to 34,596 total cumulative cases in Hubei by February 14. Based on cumulative reported cases as of February 9<sup>th</sup> these estimates correspond with an average of 7409 to 7496 additional cases in Hubei within the next 5 days. Ten days forecasts generated on February 9, estimated 36,854 to 37,230 cumulative cases, on average, in Hubei by February 19. These estimates correspond with an additional 9754 to 10,130 cases in Hubei on average in the next 10 days. Finally, the 15-day forecasts predict a cumulative reported case count between 37,415 and 38,028 cases, on average, in Hubei by February 24(5).

The very well-known model for infectious diseases is SEIR, which was used by the study on COVID-19 outbreak on the Diamond Princess Cruise ship(4, 6).

As it was justified in the study of modeling the epidemic dynamics in China, since COVID-19 outbreak and control characteristics are distinct from existing infectious diseases, the existing epidemic models such as SIR and SEIR cannot be applied to describe the observed data directly. A model called Susceptible, Un-quarantined infected, quarantined infected, confirmed infected (SUQC) model to characterize the dynamics of COVID-19 and explicitly parameterize the intervention effects of control measures, which is more suitable for analysis than other existing epidemic models(7).

However, to use the SUQC model for the Amhara region's case, there is a need to know at least the initial values for the four variables of the model. Among these variables, the number of cumulative confirmed infections,  $C(t)$ , is usually the only variable with daily observed data for model fitting and parameter inference. Besides, the initial value of susceptible individuals  $S(t)$  is approximately equal to the population size. The initially confirmed infections  $C(t)$  is the number of infected obtained from the

region's official report. Other parameters and initial values, including the quarantine rate ( $\gamma_1$ ), the confirmation rate ( $\beta$ ), the initial number of un-quarantine infected cases ( $U_0$ ), and the initial number of quarantine infected cases ( $Q_0$ ) to be estimated by fitting the daily time series of confirmed infections to the model defined as  $\hat{C}=f(\gamma_1, \beta, U_0, Q_0)$ . However, it is too early to use this model since the data for the daily confirmed infections of the region are inadequate, being only six reported at the time of the study conduct.

Furthermore, none of the models had incorporated age and sex as factors for both to acquire and die of COVID-19. Moreover, while DisMod II was developed from the perspective of Global Burden of Disease studies, it may be of use for epidemiologists and public health researchers or anyone else trying to estimate COVID-19 epidemiology. Therefore, before prediction, estimation of some variables and parameters are essential through an epidemiological disease modeling approach using WHO DisMod-II software considering one extreme scenario faced in the world as worst. However, this modeling study tried to use different modeling approaches despite their limitations, including SIR and SEIR. This study aims to forecast the epidemic situation of COVID-19 and estimate and predict cases and deaths of the Amhara region using the world's best and worst scenarios using SEIR, SIR, and DisMod-II.

## Methods

### Study area and period

This study was conducted to model the expected number of COVID-19 cases in the Amhara region using the worst scenarios faced in the world. By the year 2020, the population is approximately 22,189,999 (11,068,826 males and 11,121,173 females), of which 80.58% (40.73% males and 39.85% females) live in rural areas. Amhara region is one of the nine National Regional states in Ethiopia. The region is further divided into 11 administrative Zones (North Gondar, South Gondar, North Wollo, South Wollo, North Shewa, East Gojjam, West Gojjam, Awi, Wag Himra, Oromiya and Bahir Dar special Zone) and 164 administrative Woredas. This study was conducted from April 2020 to May 2020.

### Study design and Levels of Analysis

This study followed a time-series type of study design based on the review of the global and the United States reports, Epidemiological studies, the Worldometer website mortality and morbidity reports of COVID-19. Amhara region's population size and mortality pattern by age and sex structure also employed to estimate and predict the expected number of COVID-19 cases and deaths. The analysis was done at three stages: first at total regional population, second at regional population by sex separately (male and female), and third at regional population by five years age group category.

## Model Overview

# SIR Model

The SIR model is the flows of people between three states: susceptible (S), infected (I), and resistant or recovery (R). In Susceptible-infection-recovery (SIR) Model, we shall by  $S(t)$ ,  $I(t)$ , and  $R(t)$  denote the number of susceptible, infectious, and recovered, respectively, in the population at time  $t$ . Furthermore, it is assumed that the population consists of a constant number of  $N$  individuals and at all times  $S(t)+I(t)+R(t)=N$ . Dr. Alex Hoyt develops the model in an excel template.

## SEIR Model

The susceptible-exposed-infected recovered (SEIR) models the flows of people between four states: susceptible (S), exposed (E), infected (I), and resistant (R). The SEIR model extends the SIR model to include an exposed but non-infectious class. Dr. Alison Hill develops the model. This Shiny app uses an epidemiological model based on the classic SEIR model to describe the spread and clinical progression of COVID-19.

## The DisMod II disease model

It is a straightforward model with just one prevalent state. There are two causes of death: disease and 'all other' causes, which are assumed to be independent. There are 4 transition hazards: incidence, remission, case fatality, and the 'all other mortality' hazard. These transition hazards are age-specific, but DisMod assumes them to be constant within a one-year age interval. A basic assumption of the model is that the 'all other' causes mortality for healthy and diseased people. Hence, the disease model is completely determined by the three transition hazards, such as incidence, remission, and case fatality. Under this assumption, the following three differential equations describe the model:

$$\frac{dS_a}{da} = iS_a + rC_a, \quad \frac{dC_a}{da} = -(f + r)C_a + iS_a, \quad \frac{dD_a}{da} = fC_a$$

Where the three model parameters, representing the transition hazards, are:  $i$ : incidence,  $r$ : remission,  $f$ : case fatality, and the three states are:  $S_a$ : Number of healthy people at age  $a$ ,  $C_a$ : Number of diseased people at age  $a$ ,  $D_a$ : Number of dead people at age  $a$ .

## Data source and Data analysis

This study used several data sources. Extracted data including the clinical and transmission parameters requires for the first Susceptible-Exposed-infected-Recovered (SEIR) were from the code for creating the R shiny application website <https://alhill.shinyapps.io/COVID19seir/>. This code is a SEIR model for COVID-19 infection, including different clinical trajectories of infection, interventions to reduce transmission, and

comparisons to healthcare capacity [https://github.com/alsnhll/SEIR\\_COVID19](https://github.com/alsnhll/SEIR_COVID19). For the Susceptible-Infected-Recovered (SIR) model, took the clinical and transmission parameters data from the COVID-19 SIR modeling excel template done by Dr. Alex Hoyt accessed from <https://1drv.ms/x/s!Aihonr5JPgsehKYeJ...> website on May 14, 2020. For the third DisMod II model, extracted data from an existing data source of mainly the Worldometer website designed for COVID-19 case monitoring <https://www.worldometers.info/coronavirus/>. Besides, information on variables such as the number of populations and mortality by age and sex of the Amhara region obtained from the Regional Health Bureau. Data extracted from the Worldometer website contains information on i: incidence (number of new COVID-19 cases), r: remission (number of recovered from COVID-19), f: case fatality (number of individuals who died of COVID-19) by age and sex <https://www.worldometers.info/coronavirus/coronavirus-age-sex-demographics/>. The data was extracted through the data extraction checklist. The checklist was designed in an MS-excel worksheet. The extracted data entered DisMod-II. Once all available data on COVID-19 have been collated, first, whether the observations are all internally consistent was assessed. Analyzing whether the separate sources of information on remission, incidence, and mortality are consistent was checked using a computer. DisMod-II is a computer software program developed for the GBD that allows the user to check if a set of assumptions on incidence, prevalence, remission, case-fatality rates and observed mortality numbers are consistent with one another. DISMOD-II allowed us to change the age number and size for input and output variables as we require. It was advantageous when the epidemiological parameters or study indicators obtained or extracted from different sources in age groups differ from the age groupings in which we want to present the results for the Amhara region. The incidence rate, case fatality rate, and remission rate of COVID-19 for the Amhara region were estimated as an output using the region's population size by age and sex structure through the DISMOD software using an indirect standardization method. Subsequently, the expected cumulative incident cases and deaths of the region is calculated by multiplying the population size by the estimated incidence rate and case fatality rate as an output for the observed worst (highest) scenario. Two additional models, such as SEIR and SIR, were done for comparison purposes of their output.

## Variables and parameters

For the DisMod-II modeling, the variables used as input were the world's highest incidence rate, case fatality rate, and remission rate by age and sex reported daily since it started in the USA for the last three months (from February 15 to May 13,2020). In addition, the Amhara region's population size by age and sex structure and mortality pattern were the variables used in this model. Moreover, clinical and transmission parameters for the SEIR (Table 1below) and SIR models were used (Table 2below).

**Table 1** Parameters included in SEIR model, Amhara Region, Ethiopia, 2020

Clinical parameters and transmission rate parameters			
Model parameters	Value	Model parameters	Value
Duration incubation period	5 days	Case fatality rate (auto computed)	2.50%
Duration of mild infection	6 days	Population size	22,189,999
% infections that are severe	15%	Initial number infected	1
Duration of severe infection (hospital stay)	6 days	Transmission rate - mild infections	per day
% of infections that are critical	5%	Mild infection ( $R_0 \times 0.15$ )	0.45
Duration critical infection (ICU stay)	8 days	Sever infection ( $R_0 \times 0.05$ )	0.15
Death rate for critical infections	50%	Critical infection ( $R_0 \times 0.03$ )	0.09

**Table 2** Parameters included in SIR model, Amhara Region, Ethiopia, 2020

Model parameters	Value	Definitions
Contact Rate	variable	The number of susceptible peoples an infectious person contact.
Transmissibility	2.0%	The probability that a contact between a susceptible person and an infected person results in infection.
Duration of Infectiousness	10	The period of time (in days) an infected person can pass the infection
% Needing Hospitalization	13.80%	The proportion of infected people who require hospital care.
% Needing ICU Care	4.70%	The proportion of infected people who require ICU care.
Mortality rate	0.90%	The proportion of infected people who die.

## Adjusting the input data

The extracted and entered data in DisMod-II were smoothed using two techniques. The incidence rate and case fatality rate were smoothed or adjusted with a cubic spline interpolation technique since it was observed best fit than other methods. In contrast, the remission rate smoothed with two periods moving average. Finally, the adjusted input data were used to estimate the incidence rate, case



fatality rate, remission rate by age, and sex. The prevalence and mortality rate from all other causes by age and sex were also estimated.

## **Describing the age-sex pattern and temporal trend of COVID-19**

The next step in the analysis was to look into trends by an age-sex pattern of the expected number of new cases and deaths of COVID-19 in the Region. The incidence rate patterns, case fatality rate, and remission rate were described numerically and graphically by age and sex. Furthermore, the temporal trend of COVID-19 is described using SEIR and SIR models.

### **Results And Discussion**

Three models were used to estimate deaths and cases in the Amhara region. Each of them has their strength and limitations. The result obtained from each modeling technique was presented and discussed as follows:

### **Susceptible-Exposed-Infected-Recovered (SEIR) Model**

For this model, parameters mentioned above: clinical parameters, the transmission rate parameters, and simulation values by considering the context of the Amhara region.

Our projection shows that with no intervention, the peak day will be at the beginning of August (150<sup>th</sup> day after a case with coronavirus is confirmed); the region will experience about 4,222,172 cases. On the 100<sup>th</sup> day, the day will be 21<sup>st</sup> June; the cases will reach 21, 917,130. In terms of the population exposed, the peak day will be on July 29<sup>th</sup> with 3,013,605 cases. After putting the different parameters in the model, findings indicate mild infection will be high in the region at 140 days, 31<sup>st</sup> July with 3, 167,679 cases. It is estimated that people to be infected severely and critically in July will be the highest in experiencing 3,310,069 cases and 142,548 cases, respectively. After 125 days, the region will start having people recovered from the pandemic, on that day (16<sup>th</sup> July) with 1,694,316. The worst side of this pandemic is its fatality rate; the number of deaths that will be experienced high in the region estimated will be on the 160<sup>th</sup> day on August 20<sup>th</sup> with 273,458 deaths (Figure 1).

Even though the model predicts the peak number of severe, critical, and recovered cases could be occurred in July, what observed in the region in July was among 33, 166 individuals tested for Covid-19, about 642 and 478 were found test positive, and recovered from Covid-19 respectively. In addition, contrary to the predicted high number of deaths in August, the observed total death in August was 25 in the region. These contradicting findings might be due to mainly two possible reasons: first, the testing capacity of the region was so limited, particularly during the start of the pandemic, this, in turn, might miss infections and deaths from Covid-19; second, took the clinical and transmission parameters used in

the model from other countries study estimates, which might not represent the developing countries' including Ethiopia's context.

To help the evaluation, we generated different scenarios. We tested the potential effects of intervention strategies to control or reduce social contacts. As literature shows, some public health interventions could curb the transmission of the pandemic in different degrees (8). For this study, we used an assumption of the region's implementing of intervention strategies, if able to reduce 30% of the virus's transmission; we estimated the following situations that the region might experience.

The results of the scenario with a 30% reduction in transmission describe the peak day will be at 222 days, on 21st October with 2,407,042 cases (Figure 2). A drastic decrease in the number of cases compared to no intervention model if we increase the days of peak case burden by more than two months. A reduction in the reproduction rate will help in slowing the community transmission.

The result of the scenario with a 40% reduction in transmission was a decrease of about 2 million cases and able to increase the actual daily protection rate around 12 times. The peak day (December 15<sup>th</sup>) will be pushed by beyond four months, compared to the model with no intervention. It can be achieved by reducing both infection and death if strict public health intervention measures are implemented in the region (Figure 3).

### **Susceptible-Infected-Recovered (SIR) Model**

#### **Number of Susceptible, Infected and Recovered Cases for COVID-19 if no Non-Pharmacologic intervention is implemented**

With and without non-pharmacologic intervention (NPI) the susceptible-Infected-Recovered (SIR) model was the most commonly used modeling technique for infectious diseases. Based on the SIR model the expected number of cases by date is shown in Figure 4 below. Without NPI that was designed in Excel as SIR modeling template and taking the parameters shown in Table 2 above as well as the total population of the Amhara region (22, 189, 999) to be susceptible the predicted number of peak COVID-19 cases in the region was estimated to be 5, 947, 685.

This finding emphasizes the peak date of occurrences of cumulative cases, need for hospitalization, and ICU care and death; accordingly, the above peak figures all were expected to be occurring on the same date on July 2, 2020. This prediction is without any non-pharmacologic interventions targeting in minimizing the daily contact rate such as school closure, physical distancing, self-isolation, quarantine, and lockdown.

#### **Number of cases in need of Hospitalization and ICU care and Death from COVID-19 if no Non-Pharmacologic Intervention is implemented**

Assuming that there are no interventions made in the region, the peak number of cases in need of hospitalization was expected to be 820,781 occurring on the date July 2, 2020. Regarding the maximum

number of cases in need of ICU care, it was expected to be 279,541 the same date of peak hospitalization, July 2, 2020, and on the same date, the maximum number of death (53,529) is expected (Figure 5).

### **Number of Susceptible, Infected and Recovered Cases for COVID-19 when a Non-Pharmacologic intervention (NPI) is implemented**

If community interventions mainly focusing on the reduction of contact rate are applied, such as school closure and stay at home lockdown with an assumption of 80% reduction in contact rate (3.35) for at least six weeks from 4/15 to 5/30, the peak expected number of infected cases would be shifted from July 2, 2020, to August 27, 2020 (Figure 6).

### **Number of cases in need of Hospitalization and ICU care and Death from COVID-19 with a Non-Pharmacologic Intervention (NIP)**

When a community intervention of school closure and lockdown are applied with an assumption of an 80% reduction in contact rate (3.35) from 4/15 to 5/30; the peak expected number of cases needing hospitalization and ICU care, and the peak number of deaths would be shifted from July 2, 2020, to August 27, 2020 (Figure 7).

### **Number of Susceptible, Infected and Recovered Cases for COVID-19 with a Non-Pharmacologic intervention (NPI and reactive approach)**

If a non-pharmacologic intervention with a reactive approach is applied, targeting an 80% reduction of the contact rate during the first ten weeks (from April 15/2020 to June 30/2020) followed by a 20% reduction of the contact rate during the next 8 months (from July 1/2020 to March 12/2021) one can expect the following results. The peak number of infected cases would be 4,051,875 on November 10/2020 (Figure 8).

With the same consideration of the NPI mentioned above, the peak number of cases needing hospitalization would be 559,159, while those needing ICU care are 190,438. The estimated death is 36,467 on November 10, 2020 (Figure 9).

If the NPIs are implemented successfully one can see how the date for the peak number of cases, needing hospitalization and ICU care as well as, death could be shifted from being early to late occurrences. This mechanism of delaying the outbreak in turn assists to develop the health institutions' capacity and to have preparedness for emergency response. However, such kinds of NPI couldn't be implemented sustainably due to their socio-economic impact hence releasing the restrictions and reopening schools is mandatory. The questions such as when and how the restrictions to be released depend on the levels of healthcare capacity and the preparedness for emergency response. Therefore, another alternative approach such as NPI of some duration taking the capacity developing time in to consideration with a reactive reduction is required.

If NPIs and a reactive reduction in the contact rate is implemented, keeping all other parameters constant and assuming an 80% reduction in Contact Rate (3.35) for ten weeks (from April 15 to June 30, 2020) and then a reactive reduction of 20% (10.72) for the rest of six months, two important results were observed. The first was the epidemic curve could be flattened as a result the peak number of infected cases reduced from 5, 947, 685 cases to 4,051,875, those in need of Hospitalization decreased from 820,781 to 559,159, needing ICU care from 279,541 to 190,438 and death from 53,529 to 36, 467. The second observation was a pronounced dalliance in the epidemic from July 2 to November 10, 2020. These observations are essential findings that may inform us if this approach is used. It is possible to minimize the risk of acquiring the diseases and dying from it in one hand and get enough time to build capacity and prepare for the emergency response. Despite the important role of NPI in flattening the epidemic curve, low testing coverage and detection rate in the region could result in the hidden epidemic because onward transmission from asymptomatic and undetected infections accounts for 65% infectiousness (9). The peak day determines the duration of the epidemic, which is also dependent on our testing capacity. Testing campaigns can reduce the infection peak because the diagnosed population enters quarantine or treatment center and is less likely to affect the susceptible people(10). Moreover, on the month November, the observed data showed that among 165, 498 individuals who tested for Covid-19, about 6,125 were found test positive, 3,226 recovered, and 89 died.

#### DisMod epidemiological diseases model

We used this model by considering the worst-case scenario to estimate the expected number of COVID-19 cases and deaths. Using the USA COVID-19 infected cases and deaths data by age and sex, it was tried to estimate the regional incidence, case fatality, and recovery rate in DisMod-II software with an indirect standardization method to the region's population size by age and sex. Accordingly, the expected number of infected cases and deaths were calculated by age and sex and presented as follows. Below is the male population graph on COVID-19 cumulative incidence rate, case fatality rate, and remission rate. In the graph, one can observe how the case fatality rate increases as age increases and the remission rate decreases as age increases, whereas the incidence rate remains more or less constant, except it is very low at an early age and low at an older age. Moreover, during an epidemic for an emerging disease not occurred previously, can observe that the cumulative incidence rate is more or less equal to the prevalence rate. This pattern is the same for both males and females though females have lower incidence and case-fatality rates compared to males at similar age while females have a higher remission rate compared to males (Figure 10 and Figure 11).

One can also estimate the expected number of COVID-19 cases, deaths, and recoveries overall or a specific age group by sex in the region using Table 3 and Table 4. The total annual cumulative incident cases in the region were estimated to be 837,348(474, 809 male and 362, 540 female cases). The expected number of total deaths in the region were also estimated at 44,247 (30, 176 male and 14, 071 female deaths).

The Covid-19 data observed in the region so far showed that among 243, 638 total tested peoples, about 6760 were positive for Covid-19 and 3, 552 recovered while 125 deaths were registered. This observation is completely contradicting what was forecasted using the DisMod-II. It might be due to both the region's capacity for testing and the unrepresentative initial rates of the variables taken and used for the model.

The annual cumulative incidence rate and case fatality rate in Table below one can estimate the expected number of COVID-19 cases among the Amhara region's male population, and it is 4.2896% of 11,068,826 male population that is estimated to be 474, 809. In addition, the expected number of COVID-19 deaths among the male population of the Amhara region is 6.3553% of the expected male COVID-19 cases (474, 809) is estimated to be 30, 176 cases.

**Table 3** COVID-19 expected rates of new cases, deaths and remission among males by 5 years age group in Amhara Region, Ethiopia, 2020.

DisMod II input and output, database COVID-19

Males, Disease: COVID-19U (Rates \* 100), sex: Males

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Age	Incidence (rates * 100)	Prevalence (rates * 100)	Remission (rates * 100)	Case fatality (rates * 100)	Mortality (rates * 100)
0-4	0.0044	0.0041	99.34	0.0303	0
5-9	0	0.0003	99.34	0.0389	0
10-15	0.8692	0.5259	99.34	0	0
15-19	3.9207	3.2261	99.34	0.0086	0.0003
20-24	6.7598	6.2527	99.34	0.3143	0.0198
25-29	7.8523	7.7006	99.34	1.2365	0.0953
30-34	7.7227	7.6525	99.3365	3.0425	0.2326
35-39	7.5197	7.2371	99.2863	5.8019	0.4196
40-44	7.9559	7.3171	99.0899	8.9852	0.6582
45-49	9.1516	8.1105	98.8643	11.8817	0.9663
50-54	10.3517	9.1862	98.2221	13.7892	1.2689
55-59	10.0143	9.2429	97.3381	14.1489	1.3056
60-64	8.1336	7.9319	94.7431	13.5075	1.0671
65-69	6.6564	6.5326	91.7705	13.8079	0.8999
70-74	6.4137	6.3709	83.9362	17.1405	1.0926
75-79	6.528	6.4765	78.8656	23.1684	1.5008
80+	6.498	6.7235	65.67	26.9479	1.8199
All	4.2896	3.9548	97.5812	6.3553	0.2514

Focusing on the annual cumulative incidence rate and case fatality rate, it is estimated that the expected number of COVID-19 cases among the Amhara region's female population is 3.2599% of 11,121,173, which is 362, 540 (Table 4). Also, the expected number of COVID-19 deaths among the region's female population is 3.881% of expected female COVID-19 cases (362,540), which are 14,071.

**Table 4** COVID-19 expected rates of new cases, deaths and remission among Females by 5 years age group in Amhara Region, Ethiopia,2020.

DisMod II input and output, database COVID-19					
Females, Disease: COVID-19U (Rates * 100), sex: Females					
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Age	Incidence (rates * 100)	Prevalence (rates * 100)	Remission (rates * 100)	Case fatality (rates * 100)	Mortality (rates * 100)
0-4	0.003	0.0028	99.76	0.0188	0
5-9	0	0.0002	99.76	0.0241	0
10-14	0.6309	0.3806	99.76	0	0
15-19	2.8667	2.348	99.76	0.0054	0.0001
20-24	4.9826	4.5889	99.76	0.1945	0.009
25-29	5.8098	5.691	99.7551	0.7647	0.0436
30-34	5.7111	5.688	99.7225	1.8813	0.1069
35-39	5.552	5.4168	99.6905	3.5875	0.1942
40-44	5.8725	5.5221	99.6056	5.5553	0.3072
45-49	6.7667	6.1735	99.533	7.3453	0.4547
50-54	7.6743	7.0244	99.3477	8.5236	0.5997
55-59	7.4254	7.0686	98.8287	8.7458	0.6172
60-64	6.0132	6.0072	97.1273	8.3496	0.4995
65-69	4.9049	4.8886	95.3137	8.5339	0.4161
70-74	4.7284	4.6734	90.7169	10.591	0.4951
75-79	4.8154	4.7272	88.1846	14.3183	0.6769
80+	4.8035	4.7768	82.56	16.6571	0.797
All ages	3.2599	3.0378	98.9243	3.881	0.1179

With the application of DisMod-II considering the United States of American's (USA) data by age and sex, the annual cumulative incidence rate of COVID-19 was estimated at 837,348. This result is based on the age-sex specific rates observed in the USA after a community transmission started on May 13, 2020Worldometer's report. As clearly described it in the result part, one can look at the age-sex pattern of COVID-19 cumulative incidence, Case fatality and recovery rates.



Unlike the above two models (SIR and SEIR), this model result emphasizes indicating the most affected group of peoples than the peak of several expected cases by date. Moreover, it could not detect the impact of NPI on the magnitude and in delaying peak date of occurrences of COVID-19 with this model. However, contrary to the other models, it indicates the high-risk groups (at least by age and sex) with their number expected to be infected, recovered, and dead; this, in turn, aid deciding about preventing and controlling the pandemic in the region. Accordingly, it showed that cumulative incidence is very low at lower age groups (less than 15 years old) and low at higher age group (greater than 65 years old); however, it is highest among the working-age groups (from 15 to 65 years old). It might be due to their contact rate per day. Since they are mostly unemployed young and old dependents having limited movements, peoples at the lowest and highest age groups have lower average contact rates per day compared to the middle working age groups. It might pronounce if measures like school closure were made. Even though the acquiring probability is higher in the middle age group, the fatality rate is very high among those sixty and above age group people with the lowest recovery rates. This idea is supported by the CDC report that most USA deaths are above 75s (11). Similarly, the WHO's 11<sup>th</sup>-14<sup>th</sup> weekly reports indicated that 60 and above age group peoples are at high risk of death from COVID-19 than the younger age groups(12). In this modeling, it was tried to explore the age-adjusted gender variation through an indirect method of adjustment. As a result, males' cumulative incidence rate is 1.32 times higher than females at 4,290 per 100, 000 and 3,260 per 100, 000 respectively. Similarly, the overall male death rate is 1.6 times higher than females at 6,355 per 100, 000 and 3,881 per 100, 000 respectively. This gender variation in the death rate from COVID-19 is in line with the CDC's report of COVID-19 much more fatal for men, especially taking age into account, it indicated that in New York, the overall male death rate is 1.7 times higher than the female death rate at 228 per 100, 000 and 134 per 100, 000, respectively.

## Conclusion And Recommendation

The SEIR and SIR models inform us that if non-pharmacologic interventions are applied, such as physical distancing, school closure, and others aggressively, we can delay the epidemic peak day and flatten the epidemic curve, better preparedness for the emergency response and developing health care capacity. The DisMod-II modeling informed us about the maximum annual cumulative number of COVID-19 cases that can be occurred by age and sex. Moreover, it clearly showed us the most affected groups of people; thus, though the middle age groups were more at risk for acquiring the infection, they are less at risk of death than the older ones. Similarly, males were more at risk of death than females. This information could help officials for decision making.

The study can guide decision-makers in deciding on the optimal scheduling of intervention possible to delay the peak of infection. The delay in the time of peak infection gives room for better preparedness. The best combination of interventions could less burden the health facilities with the implementation of NPIs. Adequate implementation of NPIs helps in reducing death and flattening the curve. The model predictions could help in applying effective mitigation and suppression strategies. More effective use of

strategies based on the caseload and other factors besides, the risk of specific contexts fastens in combating the pandemic is crucial in controlling the pandemic.

## Declarations

### Ethics approval and consent to participate

Ethical clearance was obtained from ethical review committee of Bahir Dar University, College of Medicine and Health Sciences, School of Public Health and Department of Biostatistics and Epidemiology. Consent to participate is not applicable since this study did not take data from individual person.

### Consent for publication

Consent for publication is not applicable. This study did not take individual person's detail such as name, images, or videos.

### Availability of data and material

The datasets generated and/or analyzed during the current study are available in the [ACCESSIBLE WEB LINKS TO DATASETS]

<https://alhill.shinyapps.io/COVID19seir/> for SEIR model

[https://github.com/alsnhll/SEIR\\_COVID19](https://github.com/alsnhll/SEIR_COVID19) for SEIR model

<https://onedrive.live.com/view.aspx?resid=1E0B3E49BE9E6828!70430&ithint=file%2cxlsx&authkey=!ACcYayKvUyplrds> for SIR model

<https://www.worldometers.info/coronavirus/> for DisMod-II model

<https://www.worldometers.info/coronavirus/coronavirus-age-sex-demographics/> for DisMod-II

### Competing interests

No, I declare that the authors have no competing interests as defined by BMC, or other interests that might be perceived to influence the results and/or discussion reported in this paper.

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No funding was received

### Authors' contributions

G.H, T.T, G.T, D.B and C.M wrote the proposal, extract data, wrote the result and the manuscript. *K.A, A.A, F.A, Y.G, G.D and E.K* reviewed the proposal and the result. All authors reviewed the manuscript.

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## Figures

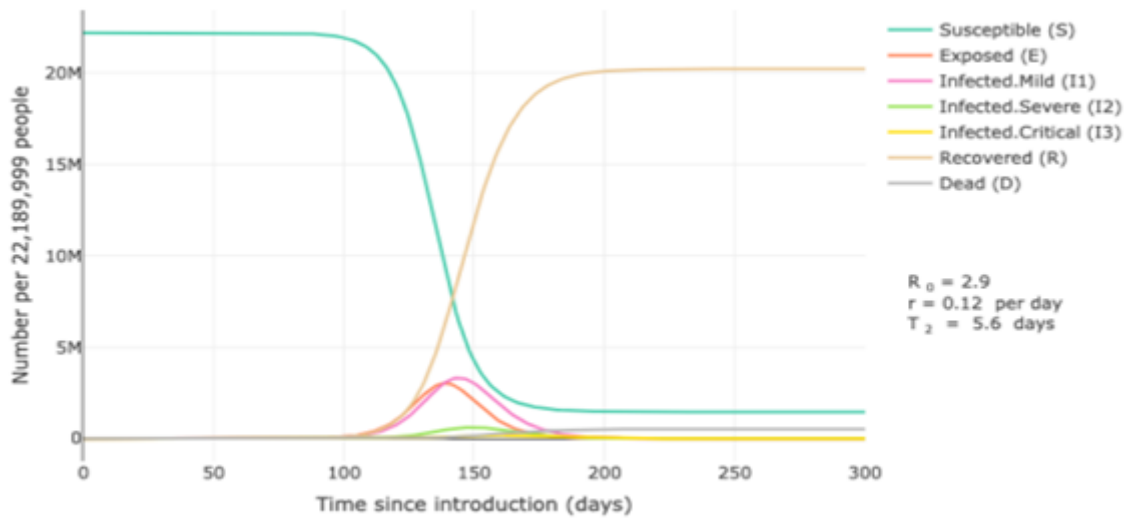


Figure 1

The SEIR model for Covid-19 with no intervention, Amhara Region, Ethiopia, 2020.

### Reduction in predicted COVID-19 after intervention

Simulate the change in the time course of COVID-10 cases after applying an intervention

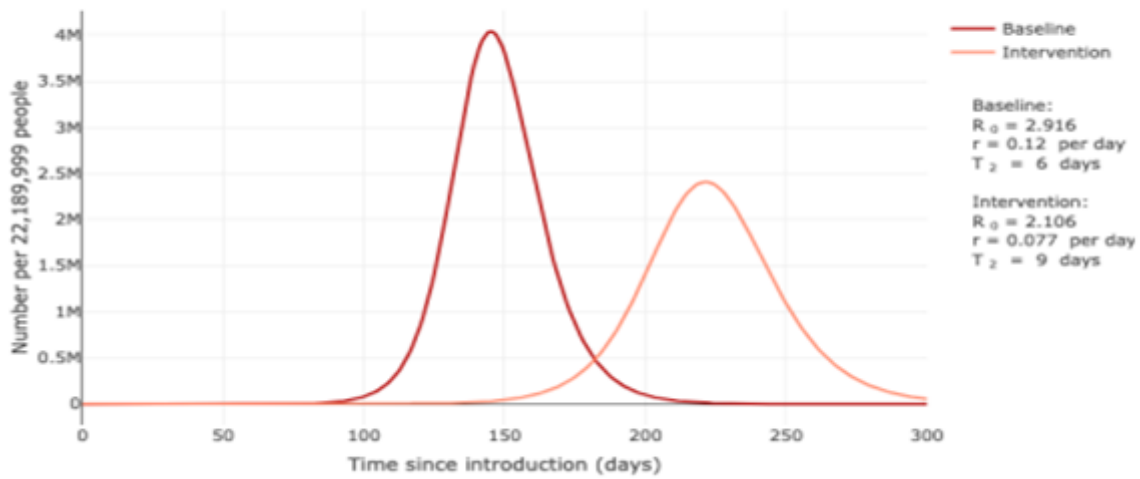
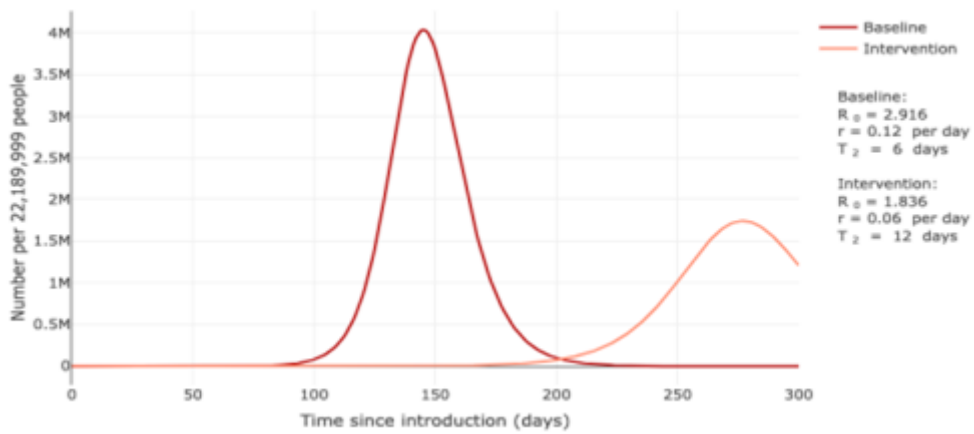


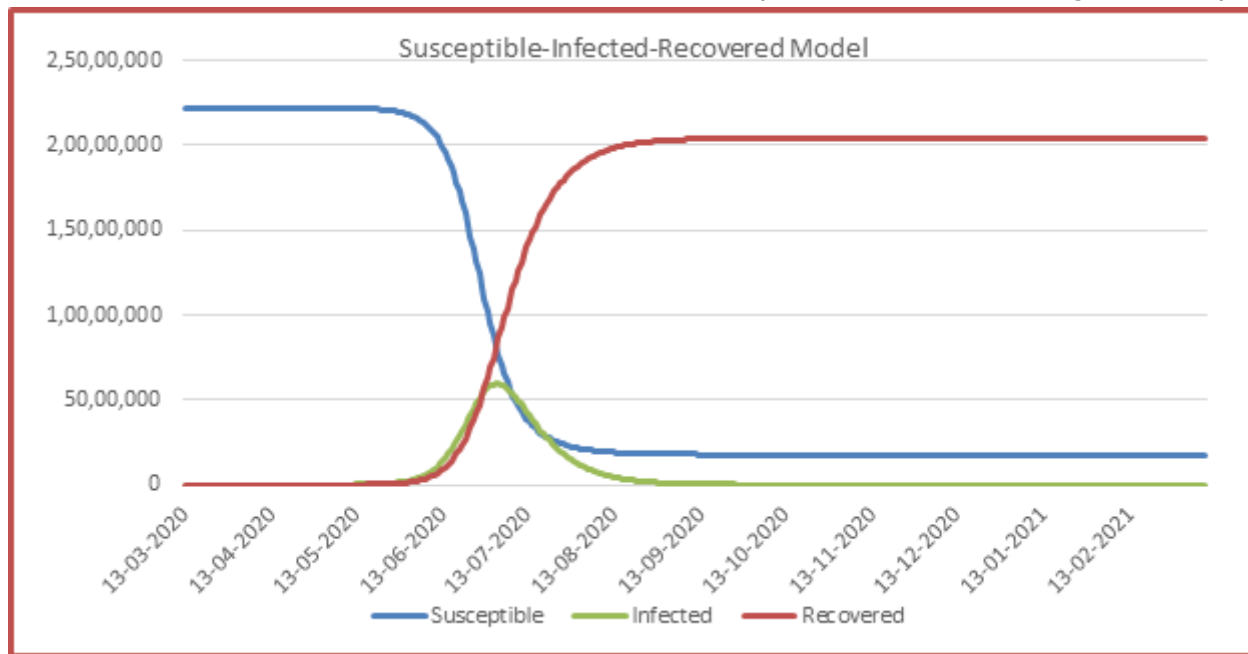
Figure 2

The model with 30% reduction in transmission of the pandemic, Amhara Region, Ethiopia, 2020.



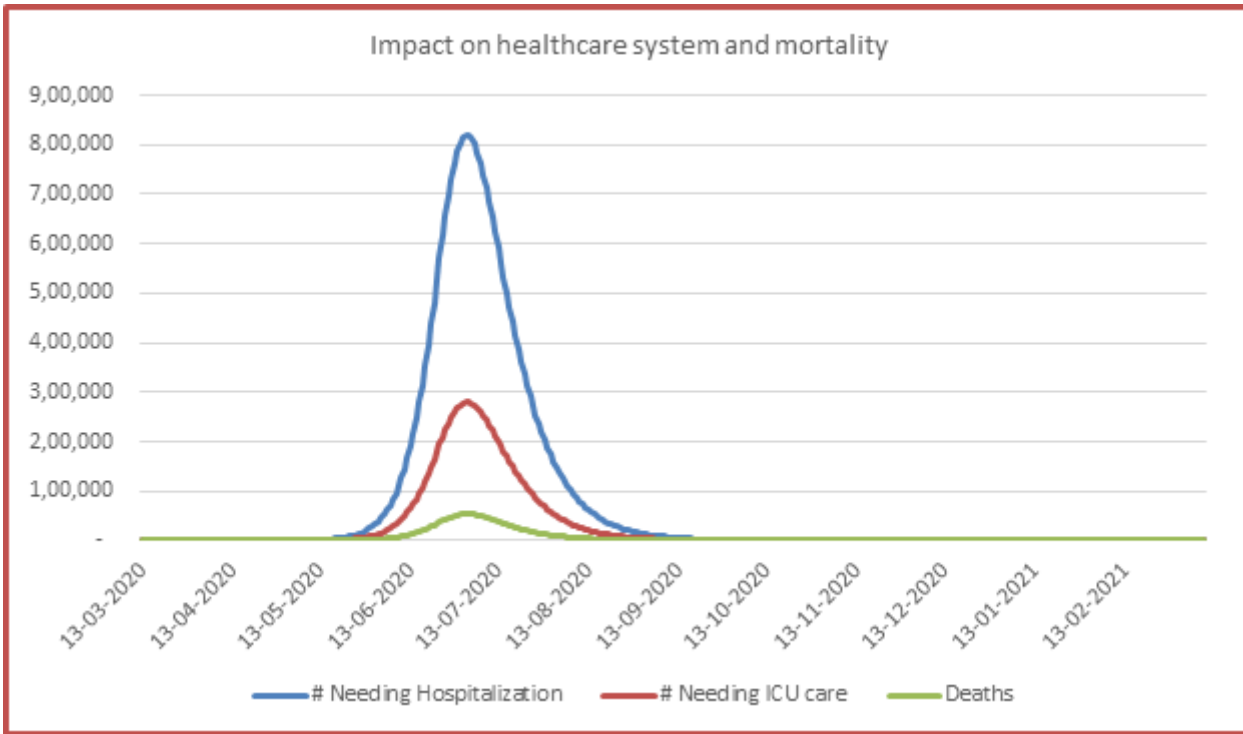
**Figure 3**

The model with 40% reduction in transmission of the pandemic, Amhara Region, Ethiopia, 2020.



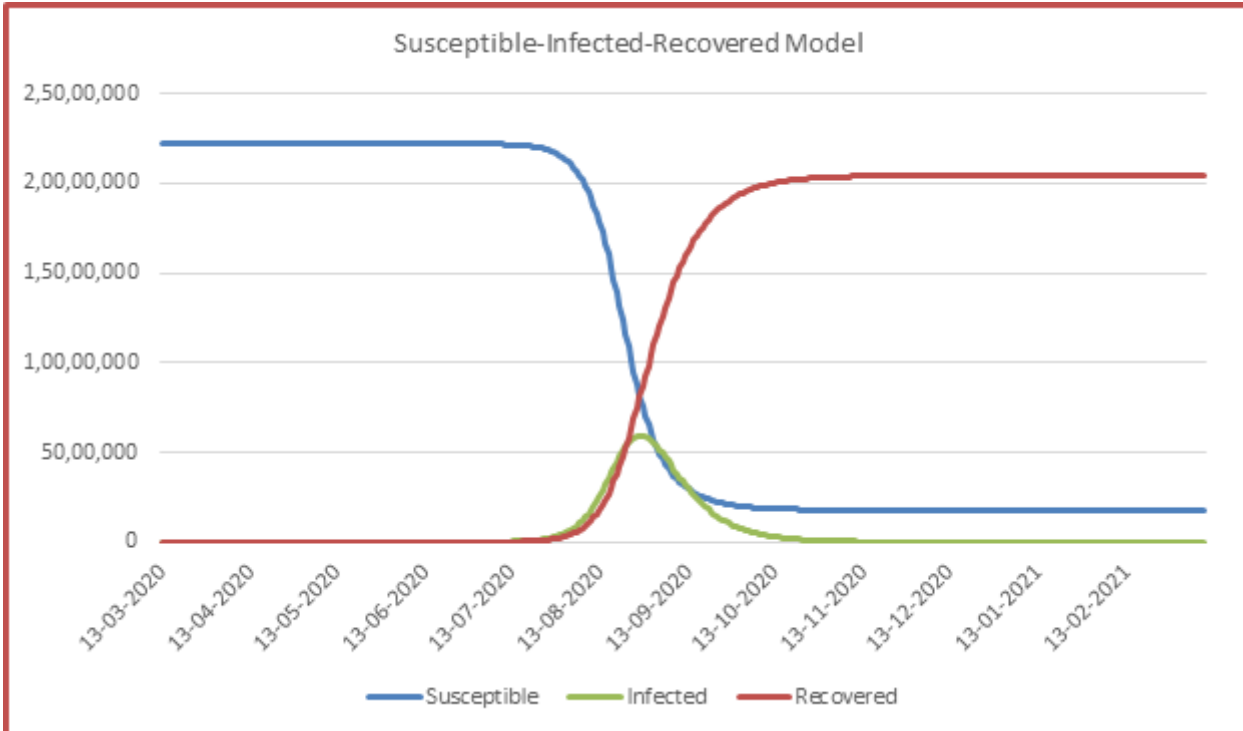
**Figure 4**

Expected Number of Susceptible-Infected-Recovered (SIR) cases for COVID-19 by date if no non-pharmacologic intervention model in Amhara Region, Ethiopia, 2020.



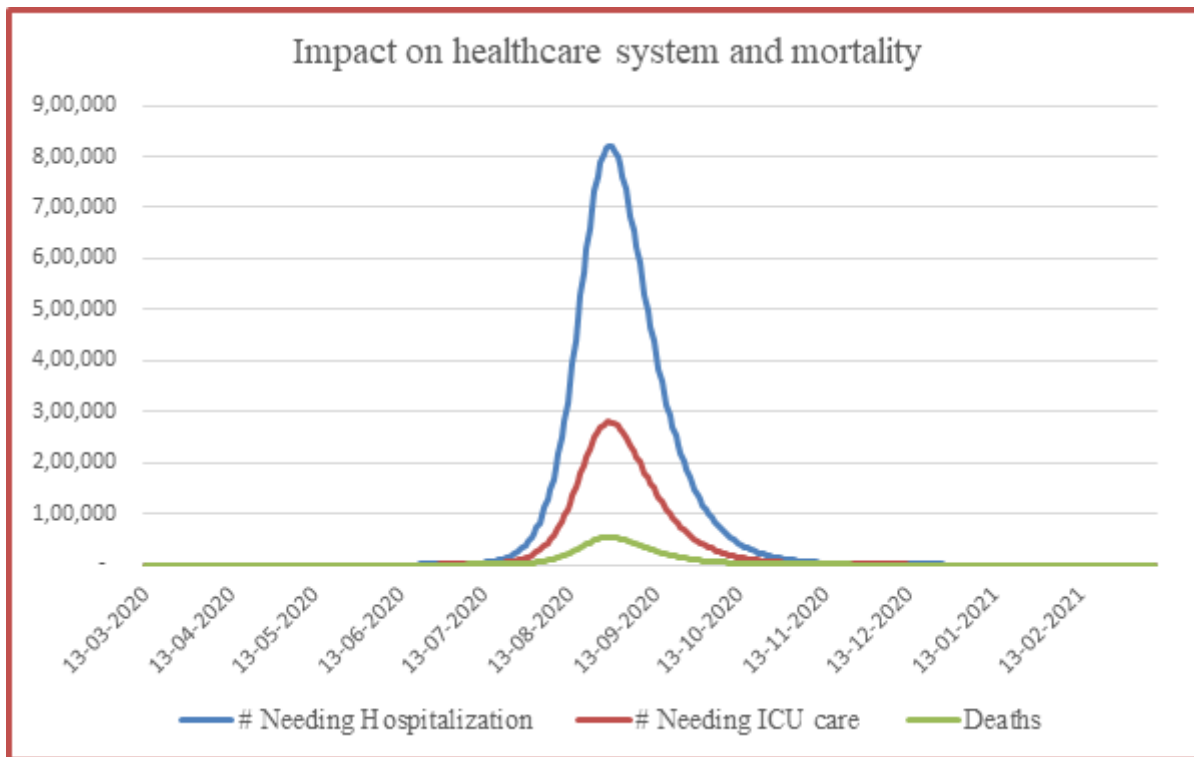
**Figure 5**

Expected Number of cases needing hospitalization and ICU care and death from COVID-19 by date in Amhara Region, Ethiopia, 2020.



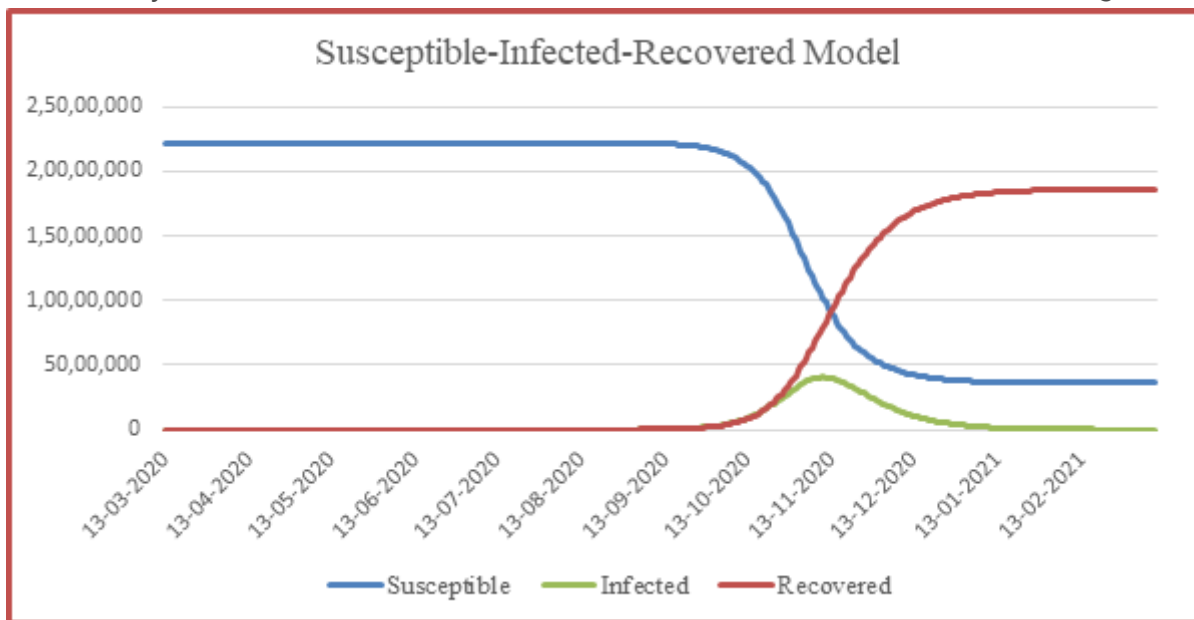
**Figure 6**

Expected Number of Susceptible-Infected-Recovered (SIR) cases for COVID-19 by date with a community intervention aimed at 80% reduction of Contact rate in Amhara Region, Ethiopia, 2020.



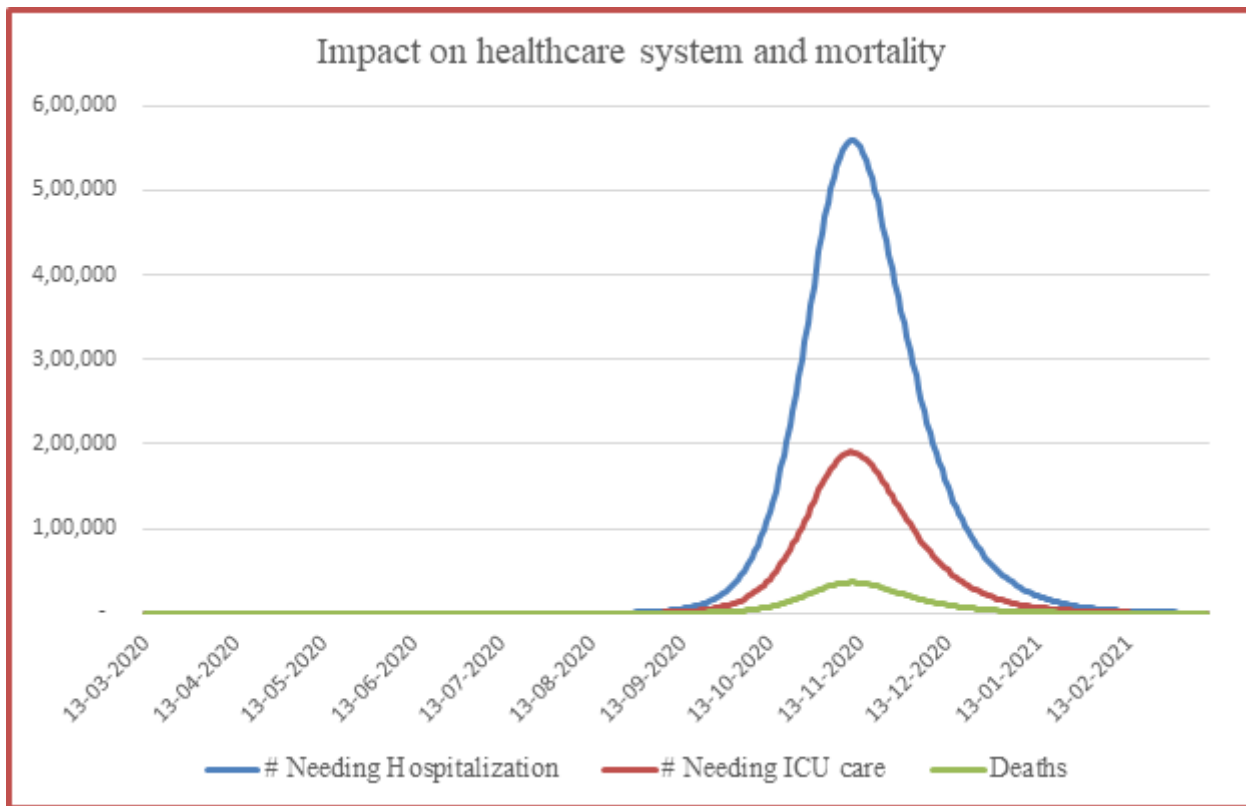
**Figure 7**

Expected Number of cases needing hospitalization and ICU care and death from COVID-19 by date with a community intervention aimed at 80% reduction of Contact rate in Amhara Region, Ethiopia, 2020.



**Figure 8**

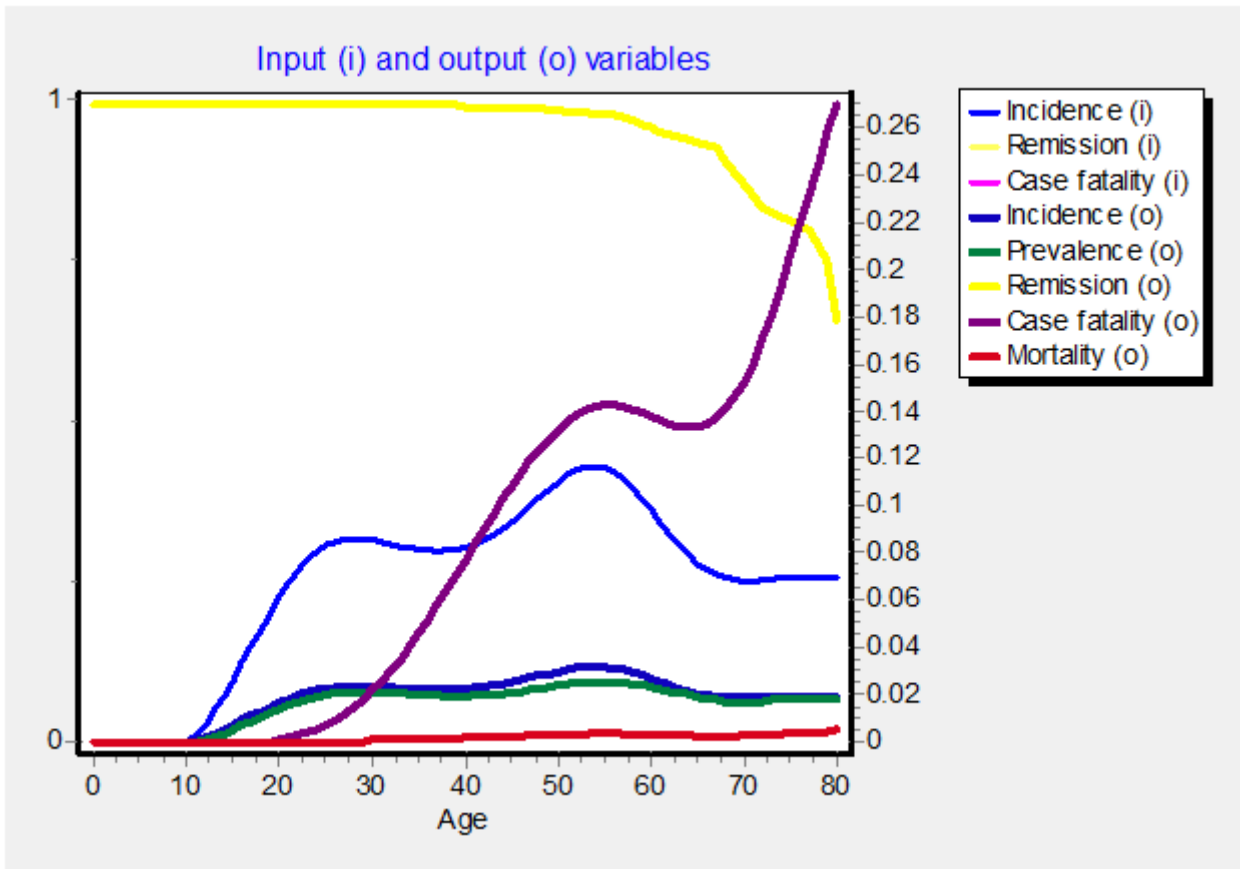
Expected Number of Susceptible-Infected-Recovered (SIR) cases for COVID-19 by date with an intervention aimed at 80% reduction of Contact rate in Amhara Region, Ethiopia, 2020.



**Figure 9**

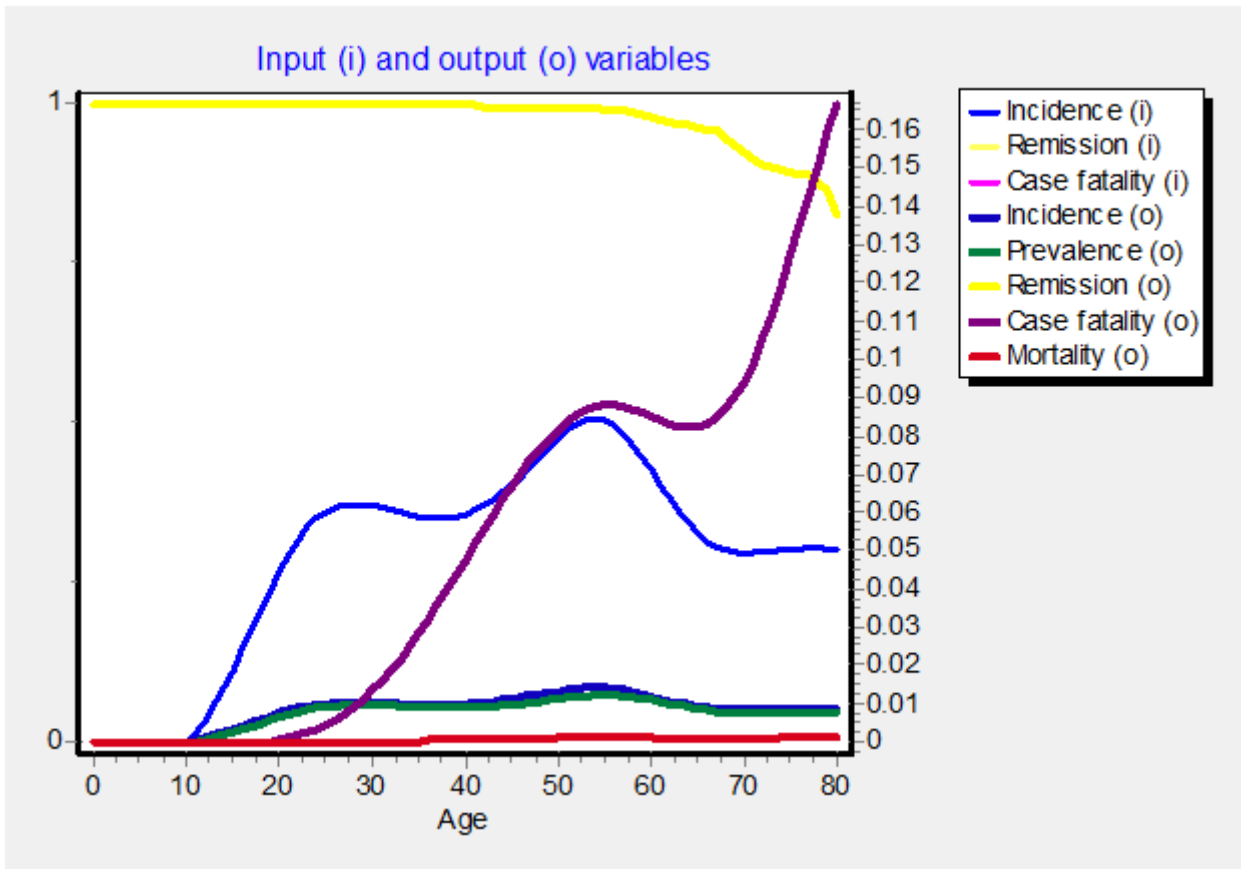
Expected Number of cases needing hospitalization and ICU care and death from COVID-19 by date with a community intervention aimed at 80% followed by 20% reduction of Contact rate in Amhara Region, Ethiopia, 2020.





**Figure 10**

Expected incidence, recovery and case fatality rates from COVID-19 of males by Age in Amhara Region, Ethiopia, 2020.



**Figure 11**

Expected incidence, recovery and case fatality rates from COVID-19 of Females by Age in Amhara Region, Ethiopia, 2020.