

CoVid-19 propagation mathematical modeling: the case of Senegal

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Abstract

The outburst of the CoVid-19 pandemic has raised several questions leading to a complex system in terms of modeling. Indeed, the modeling of the epidemic, at the level of a country, needs considering each of the different sources of contamination as well as the public health authorities strategy, in a specific way. With this in mind, in the present paper, we develop a mathematical model of the CoVid-19 epidemic in Senegal. In the model, the population is subdivided into five compartments: susceptible, infected but asymptomatic, symptomatic, quarantined, and recovered immune people. In addition, due to its important impact in the propagation of the disease, we add one more variable: the number of infected objects. Therefore, the model corresponds to a system of six non-linear ordinary differential equations we submit to an analytical study to prove the relevancy of the model, simulate the evolution of the epidemic, and retrieve epidemiological parameters, namely the infection rate and the basic reproduction number, \mathcal{R}_0 . Based on the senegalese territory CoVid-19 data, we simulate various scenarios as for the evolution of the epidemic in the country, in order to predict the peak and its magnitude with regard to the application of barrier measures. We also explore the option of collective immunity with special protection for vulnerable people. In doing so, non available parameters are identified using some mathematical identification technics.

Keywords: CoVid-19, Mathematical model, *Fitting* strategy, Reproduction number.

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1. Introduction

Since December 2019, the World has been facing a very rapidly spreading pandemic, the CoVid-19 due to a virus named SARS-CoV-2. Starting in Wuhan, China, the disease has almost reached the whole globe in less than 5 months. Despite the lack of full understanding of the pandemic, due to its newness, a number of articles have been released, in mathematical modeling (see [5], [3], [6], [7], [8], [12], [13]) as well as in various factors of propagation of the virus (see [15], [2]). In the early days of the pandemic Liu et al in [5] have developed a CoVid-19 propagation model taking into account, in addition to susceptibles and asymptomatics compartments, a splitting of the symptomatics' into two compartments: the identified and the non identified. They have also addressed other aspects of the pandemic such as the latency phenomenon related to CoVid-19 in [7]. As a crucial factor, age structure is addressed by Griette et al in [3], in the context of Japan. And, on another note, Ogbunugafor et al in [12] came up with a CoVid-19 model which involves the contamination by infected objects.

Moreover, numerous works have been done in the short period of time, which contribute to a better understanding of the CoVid-19 propagation processes. For instance, the work of Doremalen et al in [2] gives a broad overview of the life expectancy of SARS-CoV-2 on different kinds of objects, and in the air. Their paper states that the virus' life expectancy cannot exceed: 72 hours on plastic, 48 hours on stainless steel, 24 hours on cardboard, 4 hours on copper. Such information is of capital importance for modeling the effects of contaminated objects in the propagation of the virus.

Two cornerstones in the contamination process by infected objects are also given by practical and natural human habits. They are *Human Objects Touching* (HOT) and *Human Face Touching* (HFT). These phenomena are studied in a book from Zuccotti, (see [17]) for HOT, and in a paper of Kwok et al in [10] for HFT. As for the influence of climate conditions like temperature and humidity, a study is already done by Wang et al in [15].

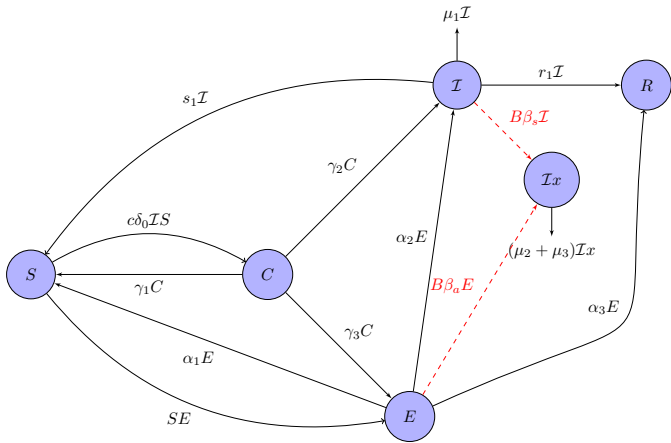
Relying on the relatively abundant literature, in view of the newness of the CoVid-19 pandemic and its virus, the present work presents a model which fits the epidemic management in Senegal. Indeed, the model complies with the different strategies adopted by the public health decision-makers of Senegal. In that respect, we have considered a quarantined compartment and used the related daily data released by the authorities. The other compartments of the present model are quite classical apart from one. Indeed, in terms of human compartments, in addition to the quarantined people compartment, we consider the susceptibles, the asymptomatics, the symptomatics and the removed. And, the novelty in the present paper is the infected objects compartment we have added, and the way we consider newly contaminated objects by properly defining the coefficients. This is critical in avoiding redundancy while counting the contaminated objects.

As for the organization of the paper, after the present introduction, we present the model in detail in Section 2. Then, we do some qualitative analysis in Section 3 to verify the boundedness of the infected objects variable, calculate \mathcal{R}_0 and analyse its sensitivity. The fourth section is devoted to exploiting the data to retrieve the epidemic parameters such as the infectivity rate, and to simulating some scenarios with regard to compliance or not with barrier measures.

2. Derivation of the model

The mathematical model discussed in this paper is a system of 6 non linear differential equations. The variables are defined in the following table .

Variables	Descriptions
S	Susceptible population
E	Infected asymptomatic population
C	Quarantined population
\mathcal{I}	Infected symptomatic population
$\mathcal{I}x$	Infected objects
R	Immune recovered population



$$\gamma_1 = (1 - \delta^{ci}/c)/14\,C$$
$$\gamma_2 = (1 - ra)\delta^{ci}/(3c)C$$
$$\gamma_3 = ra\delta^{ci}/(3c)C$$

$$SE = [(1 - \delta_0)\delta^{ci}\mathcal{I} + \delta^x\mathcal{I}x + \delta^e E]$$
$$B = B\left(1 - \frac{\mathcal{I}x}{B_s\mathcal{I} + B_a E}\right)$$

Parameters	Descriptions	Daily values
δ_0	Identification rate of infected people	0.7
ra	Estimated rate of asymptomatics	0.9
δ^{ci}	Infection rate of susceptibles by symptomatics	Fitting
δ^e	Infection rate of susceptibles by asymptomatics	$0.25 * \delta^{ci}$
δ^x	Infection rate of susceptibles by infected objects	$0.01 * \delta^{ci}$
α_1	The rate asymptomatics step back to susceptibility	$1/1e4*ra/7$
α_2	The rate asymptomatics become symptomatic	$(1-ra)/7$
α_3	The rate asymptomatics become immune	$9999/1e4*ra/7$
r_1	The rate symptomatics become immune	$1/7$
s_1	The rate symptomatics step back to susceptibility	$1/1e4*r_1$
μ_1	The death rate of symptomatics	0.0112
β_a	Number of objects infected by asymptomatics	20
β_s	Number of objects infected by symptomatics	5
B_a	Maximum number of objects infected by asymptomatics per day	400
B_s	Maximum number of objects infected by symptomatics per day	100
μ_2	The average death rate of the SARS-CoV-2 on objects	$1/1$
μ_3	The decontamination rate of infected objects	2
c	The contact rate of humans	Fitting

- The infection rates δ^{ci} , δ^e and δ^x , as well as the daily human number of contacts, c , are fitted to comply with the epidemiological data such as quarantined and under treatment subjects. Note that, due to the results of a recent survey initiated by the Center for Global Development (CGD, <https://www.cgdev.org>) and the *Centre de recherche pour le développement économique et social* (CRDES, <http://crdes.sn>), and leaded by A. Nestour, S. Mbaye and L. Moscoviz (see [11]), after the fitting, the mean value of c is supposed to be 10. In fact, the survey says that the average number of human contacts per capita per day is 10.
- As for the infected objects parameters, we assume that positive asymptomatics and symptomatics infect in average 20 and 5 objects per day respectively; and that the maximum number of objects per day they can infect is 400 and 100 respectively. Some references about objects touching by humans can be found in the book of P. Zuccotti [17].

For the sake of clarity, each of the terms of the model is numbered and is explained below.

$$\begin{aligned}
 \frac{dS}{dt} &= -c\delta_0^{(1)}\mathcal{I}S - [(1-\delta_0)\delta^{ci(2)}\mathcal{I} + \delta^x\mathcal{I}x + \delta^eE]S \\
 &\quad + (1-\delta^{ci(3)}/c)/14C + s_1^{(6)}\mathcal{I} + \alpha_1^{(7)}E \\
 \frac{dE}{dt} &= [(1-\delta_0)\delta^{ci(2)}\mathcal{I} + \delta^x\mathcal{I}x + \delta^eE]S + ra\delta^{ci(5)}/(3c)C - (\alpha_1 + \alpha_2 + \alpha_3)^{(7,8,9)}E \\
 \frac{dC}{dt} &= c\delta_0^{(1)}\mathcal{I}S - [(1-\delta^{ci(3,4,5)}/c)/14 + \delta^{ci}/(3c)]C \\
 \frac{d\mathcal{I}}{dt} &= \alpha_2^{(8)}E + (1-ra)\delta^{ci(4)}/(3c)C - r_1^{(10)}\mathcal{I} - s_1^{(6)}\mathcal{I} - \mu_1^{(11)}\mathcal{I} \\
 \frac{dR}{dt} &= r_1^{(10)}\mathcal{I} + \alpha_3^{(9)}E
 \end{aligned} \tag{1}$$

- (1) stands for people who are identified to have been in contact with infected people, and quarantined.
- (2) stands for not identified susceptible people who become infected.
- (3) stands for people who have been quarantined and eventually proved not to be infected after 14 days.
- (4) stands for people who have been quarantined and eventually proved to be infected.
- (5) stands for people who have been quarantined and eventually proved to be infected but asymptomatic.
- (6) stands for people who have been symptomatic and who have eventually recovered but not immune.
- (7) stands for asymptomatic infected people who have recovered but are susceptible again.
- (8) stands for asymptomatic infected people who become symptomatic.
- (9) stands for asymptomatic infected people who become immune.
- (10) stands for symptomatic people who have eventually recovered and become immune.
- (11) stands for symptomatic people who have died from the disease.

In addition, we consider a compartment of infected objects $\mathcal{I}x$ in the model. That variable fed by the infected individuals is governed by the following ordinary differential equation:

$$\frac{d\mathcal{I}x}{dt} = B\left(1 - \frac{\mathcal{I}x}{B_s\mathcal{I} + B_aE}\right)[\beta_s^{(12)}\mathcal{I} + \beta_a^{(13)}E] - (\mu_2 + \mu_3)^{(14)}\mathcal{I}x, \tag{2}$$

where B is an increasing continuous real function such that $B(0) = 0$.

- (12) stands for the release of the viruses from symptomatic people.
- (13) stands for the release of the viruses from asymptomatic people.

- (14) stands for the death rate of the released viruses. This can be augmented, through μ_3 , by using biocides.

Parameters B_s and B_a are the maximum number of objects symptomatics and positive asymptomatics can infect respectively.

Therefore, the term $B\left(1 - \frac{\mathcal{I}x}{B_s\mathcal{I} + B_aE}\right)$ is intended to avoid counting recontamination of objects while they are infected. Notice that when $\mathcal{I}x$ increases up to $B_s\mathcal{I} + B_aE$, then it starts decreasing due to μ_2 and μ_3 , the mortality and the decontamination rates of the SARS-CoV-2, respectively. The use of function B has the merit of addressing the modeling of environmental contamination in a relatively more explicit way, compared to the models we have met (see [12] , [1]).

It's also worth mentioning that the present model differs significantly from those (see [3], [5], [6], [7], [8], [16]) found in the literature and involving a quarantine compartment. Indeed, in our model, the quarantined people are not a priori qualified, that is, one does not know, contrary to the models we have met, whether the subjects in the quarantine compartment are infected or not. People in the present compartment C may be uninfected, infected but asymptomatic or symptomatic. Such a definition of the quarantine compartment is in complete compliance with senegalese health authorities strategy.

3. Mathematical analysis of the model

3.1. Boundedness of the number of objects variable

Proposition 1. Assume that for some time $t_0 > 0$, the infected objects and humans variables satisfy the following inequality.

$$\mathcal{I}x(t_0) < B_s\mathcal{I}(t_0) + B_aE(t_0).$$

If hypothesis

$$\mu_2 + \mu_3 \geq 1, \quad \alpha_1 + \alpha_2 + \alpha_3 \leq 1, \quad \text{and} \quad r_1 + s_1 + \mu_1 \leq 1, \quad (3)$$

hold, then

$$\mathcal{I}x(t) \leq B_s\mathcal{I}(t) + B_aE(t), \quad \forall t \geq t_0. \quad (4)$$

Proof. Suppose that there exists $t > t_0$ such that the following equality holds:

$$B_s\mathcal{I}(t) + B_aE(t) = \mathcal{I}x(t). \quad (5)$$

Let us now calculate the derivative of $B_s\mathcal{I}(t) + B_aE(t) - \mathcal{I}x(t)$ at time t :

$$\begin{aligned} B_s \frac{d\mathcal{I}}{dt} + B_a \frac{dE}{dt} - \frac{d\mathcal{I}x}{dt} &= B_s \alpha_2 E + B_s (1 - ra)(\delta^{ci}/c)C - B_s (r_1 + s_1 + \mu_1)\mathcal{I} \\ &\quad + B_a ((1 - \delta_0)\delta_1^{ci}\mathcal{I} + \delta^x\mathcal{I}x + \delta^e E)S + B_a ra\delta^{ci}/(3c)C \\ &\quad - B_a (\alpha_1 + \alpha_2 + \alpha_3)E + (\mu_2 + \mu_3)\mathcal{I}x. \end{aligned} \quad (6)$$

Therefore, from hypothesis (3) and identities (5)-(6), we obtain

$$\begin{aligned}
B_s \frac{d\mathcal{I}}{dt} + B_a \frac{dE}{dt} - \frac{d\mathcal{I}x}{dt} &\geq B_s \alpha_2 E + B_s (1 - ra) (\delta^{ci}/c) C \\
&\quad + B_a ((1 - \delta_0) \delta_1^{ci} \mathcal{I} + \delta^x \mathcal{I}x + \delta^e E) S + B_a ra \delta^{ci} / (3c) C \\
&\quad + (\mu_2 + \mu_3 - 1) \mathcal{I}x \geq 0.
\end{aligned}$$

We have proved that at any time $t > t_0$ where $B_s \mathcal{I} + B_a E - \mathcal{I}x = 0$, its derivative is positive. Consequently, inequality (4) is satisfied. \square

3.2. The basic reproduction number

The basic reproduction ratio \mathcal{R}_0 , gives the average number of secondary cases of infection resulting from a single primary infection in a population where everyone is susceptible. In epidemic models, it is worthy of consideration as it is the indicator of persistence or eradication of diseases, by giving a threshold depending on epidemiological parameters. Indeed, an epidemic is said under control (cannot persist in the population) when $\mathcal{R}_0 < 1$.

Using the next generation matrix technique, the basic reproduction number for the CoVid-19 model (2)-(2) is:

$$\mathcal{R}_0 = S_0 \left[\frac{\alpha_2 \beta_s \delta^x - \alpha_2 (\delta_0 - 1) \delta^{ci} (\mu_2 + \mu_3) + (\mu_1 + r_1 + s_1) (\beta_a \delta^x + \delta^e (\mu_2 + \mu_3))}{(\mu_2 + \mu_3) (\alpha_1 + \alpha_2 + \alpha_3) (\mu_1 + r_1 + s_1)} \right]$$

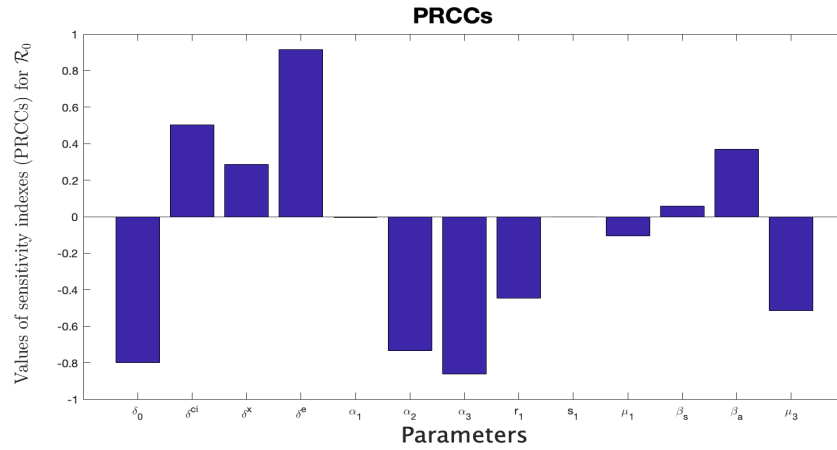
That is:

$$\begin{aligned}
\mathcal{R}_0 = S_0 &\left[\frac{\alpha_2}{\alpha_1 + \alpha_2 + \alpha_3} \frac{(1 - \delta_0) \delta^{ci}}{r_1 + s_1 + \mu_1} + \frac{\delta^e}{\alpha_1 + \alpha_2 + \alpha_3} \right] \\
&+ S_0 \left[\frac{\alpha_2}{\alpha_1 + \alpha_2 + \alpha_3} \frac{\beta_s \delta^x}{(\mu_2 + \mu_3) (r_1 + s_1 + \mu_1)} + \frac{\beta_a \delta^x}{(\mu_2 + \mu_3) (r_1 + s_1 + \mu_1)} \right]. \quad (7)
\end{aligned}$$

In the expression of \mathcal{R}_0 , the first two terms stand for the infections generated by asymptomatic and symptomatic humans, while the last two terms represent contamination caused by infected objects. With this expression, we can see that there are several ways to reduce \mathcal{R}_0 . For example, this can be done by increasing μ_3 , the disinfection rate of infected objects, or by increasing the identification rate, δ_0 .

When a pandemic breaks out, an effective strategy should be developed so that the reproduction number reduces to less than 1 as soon as possible. So, sensitivity analysis is performed to investigate the parameters that exert the greatest influence on this ratio. In this study, we perform the partial rank correlation coefficient (PRCC), which is a global sensitivity analysis technique proven to be one of the most reliable. The sign of the PRCC suggests qualitative relationship between the parameters and \mathcal{R}_0 . A positive value of the PRCC implies correlation between \mathcal{R}_0 and a parameter while negative value proves anti-correlation.

The following figure displays the correlation magnitude between \mathcal{R}_0 and the model parameters such as the transition rates α_2 , α_2 , α_3 , the transmission rates δ^{ci} , δ^e , δ^x and the average numbers of objects β_s and β_a that symptomatics and asymptomatics can infect per day respectively. The figure shows, in particular, that the magnitude of the correlation between the infection rate of objects β_a and \mathcal{R}_0 is important enough to be taken into account in the expansion of CoVid-19.



4. Numerical simulations

4.1. Fitting procedure

For the fitting process, let us consider the Euler scheme:

$$\begin{aligned}
 a. \quad S^{k+1} &= S^k + h \left(-c\delta_0 \mathcal{I}^k S^k - \left[(1 - \delta_0)\delta^{ci^k} \mathcal{I}^k + \delta^x \mathcal{I}^k + \delta^e E^k \right] S^k \right. \\
 &\quad \left. + (1 - \delta^{ci^k}/c)/14 C^k + s_1 \mathcal{I}^k + \alpha_1 E^k \right), \\
 b. \quad E^{k+1} &= E^k + h \left(\left[(1 - \delta_0)\delta^{ci^k} \mathcal{I}^k + \delta^x \mathcal{I}^k + \delta^e E^k \right] S^k \right. \\
 &\quad \left. + ra\delta^{ci^k}/(3c) C^k - (\alpha_1 + \alpha_2 + \alpha_3) E^k \right), \\
 c. \quad C^{k+1} &= C^k + h \left(c\delta_0 \mathcal{I}^k S^k - ((1 - \delta^{ci^k}/c)/14 + \delta^{ci^k}/(3c)) C^k \right), \\
 d. \quad \mathcal{I}^{k+1} &= \mathcal{I}^k + h \left((-r_1 - s_1 - \mu_1) \mathcal{I}^k + \alpha_2 E^k + (1 - ra)\delta^{ci^k}/(3c) C^k \right), \\
 e. \quad \mathcal{I}^{k+1} &= \mathcal{I}^k + h \left(B \left(1 - \frac{\mathcal{I}^k}{B_s \mathcal{I}^k + B_a E^k} \right) [\beta_s \mathcal{I}^k + \beta_a E^k] - \mu_3 \mathcal{I}^k \right), \\
 f. \quad R^{k+1} &= R^k + h (r_1 \mathcal{I}^k + \alpha_3 E^k).
 \end{aligned} \tag{8}$$

The fitting algorithm we use in this work consists on utilizing the available daily data about the people under treatment, \mathcal{I}_f , and the quarantined individuals, C_f . Indeed, we impose to the latter to be solution of (8), from day 15th to day 77th. This fitting allows to estimate the social and epidemiological data such as the average number of contacts per person per day and the probability to be infected when in contact with an infected person. Assume that the identification rate δ_0 is

known. Then, with the infectivity rate denoted by $\delta^{ci} = c\beta^{ci}$, the fitting system is

$$\begin{aligned}
 a. \quad S^{k+1} &= S^k + h \left(-c^k \delta_0 \mathcal{I}_f^k S^k - \left[(1 - \delta_0) \delta^{ci^k} \mathcal{I}_f^k + \delta^x \mathcal{I}x^k + \delta^e E^k \right] S^k \right) \\
 &\quad + h \left((1 - \beta^{ci^k}) / 14 C_f^k + s_1 \mathcal{I}_f^k + \alpha_1 E^k \right) \\
 b. \quad E^{k+1} &= E^k + h \left(\left[(1 - \delta_0) \delta^{ci^k} \mathcal{I}_f^k + \delta^x \mathcal{I}x^k + \delta^e E^k \right] S^k \right) \\
 &\quad + h \left(ra / 3 \beta^{ci^k} C_f^k - (\alpha_1 + \alpha_2 + \alpha_3) E^k \right) \\
 c. \quad C_f^{k+1} &= C_f^k + h \left(c^k \delta_0 \mathcal{I}_f^k S^k - ((1 - \beta^{ci^k}) / 14 + \beta^{ci^k} / 3) C_f^k \right) \\
 d. \quad \mathcal{I}_f^{k+1} &= \mathcal{I}_f^k + h \left((-r_1 - s_1 - \mu_1) \mathcal{I}_f^k + \alpha_2 E^k + (1 - ra) / 3 \beta^{ci^k} C_f^k \right) \\
 e. \quad \mathcal{I}x^{k+1} &= \mathcal{I}x^k + h \left(B \left(1 - \frac{\mathcal{I}x^k}{B_s \mathcal{I}^k + B_a E^k} \right) [\beta_s \mathcal{I}_f^k + \beta_a E^k] - \mu_3 \mathcal{I}x^k \right) \\
 f. \quad R^{k+1} &= R^k + h \left(r_1 \mathcal{I}_f^k + \alpha_3 E^k \right),
 \end{aligned} \tag{9}$$

where c^k and β^{ci^k} are calculated from equations (10)_{c,d}, say

$$c^k = \frac{1}{\delta_0^k \mathcal{I}_f^k S^k} \left(\frac{C_f^{k+1} - C_f^k}{h} + ((1 - \beta^{ci^k}) / 14 + \beta^{ci^k} / 3) C_f^k \right), \tag{10}$$

$$\beta^{ci^k} = \frac{3}{(1 - ra) C_f^k} \left(\frac{\mathcal{I}_f^{k+1} - \mathcal{I}_f^k}{h} + (r_1 + s_1 + \mu_1) \mathcal{I}_f^k - \alpha_2 E^k \right). \tag{11}$$

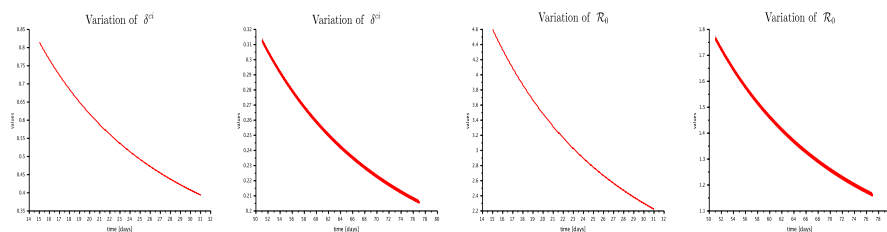
The fitting results and the simulation being tightly related to senegalese authorities strategies to fight the CoVid-19 pandemic, let us present an overview of the public decisions made by them during the first three months of the pandemic.

Senegalese Authorities responses to the COVID-19 Pandemic

Date	Measures
22 January 2020	<ul style="list-style-type: none"> - First communiqué of the Ministry of Health <ul style="list-style-type: none"> o Reinforcement of epidemiological surveillance throughout the national territory, with particular emphasis on the country's gateways; o Dissemination of information for health workers (sheets on the disease to medical regions, health districts and all public health facilities); o Starting Public information and awareness.
14 March 2020	<ul style="list-style-type: none"> - Closure of Schools and universities for three weeks - Cancellation of public events - Restriction of gatherings
18 March 2020	- Restrictions for a period of one month, of international flights coming from France, Spain, Belgium, Italy and Portugal, Algeria and Tunisia
20 March 2020	- Closure of international borders including all international and domestic flights
24 March 2020	<ul style="list-style-type: none"> - State of Health emergency <ul style="list-style-type: none"> o Curfew from 8 p.m. to 6 a.m. o Intercity Movement restriction o Closure of schools and universities extended until further notice o Limited number of passengers in Public transport o Stay at home recommended o Income support to workers (at least 70% of salary to be paid) o Creation of the Response and Solidarity Fund against the effects of the coronavirus (FORCE-COVID-19)
11 April 2020	- Starting Food distribution Programme for 1 million people to address the socio-economic effects of Covid-19
17 April 2020	- Mandatory wearing of protective mask in public places
20 April week 2020	- Starting Tests for both symptomatic and asymptomatic
04 May 2020	- Restriction of the shops and markets opening to 2 days and 3 days respectively
11 May 2020	<ul style="list-style-type: none"> - Relaxation of State of Health emergency <ul style="list-style-type: none"> o Curfew from 9 p.m. to 5 a.m. o Reopening of places of worship o Opening of the shops and markets extended to six days o Opening of schools for students in exam classes planned for 2 June 2020
1 June 2020	- Opening of schools for students in exam classes postponed

Note: First confirmed case: 2 March 2020

The infectivity rate δ^{ci} is calculated from the public data released by senegalese health authorities. As presented in the graphs below, that critical epidemic parameter has decreased by 51.65% and 34.1% in Senegal in the two periods March 14 - April 1 and April 21 - May 17, respectively. And, in accordance with the infectivity rate, the basic reproduction number \mathcal{R}_0 follows the same trend.

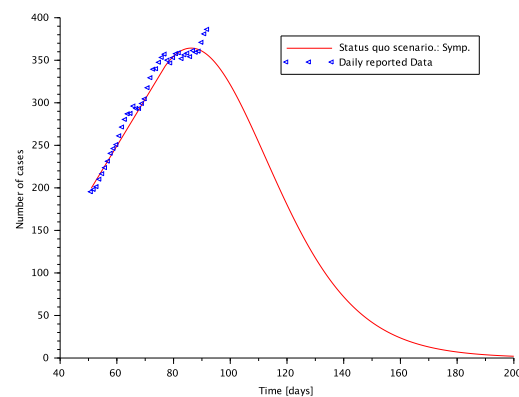


The decrease of the infectivity rate in the period March 14 - April 1, must be related to the decisions made by the authorities on March 14, mainly the closure of schools and universities. And the continuation of the decline in the period April 21 - May 17 is quite possibly due to the new measures taken, on the 20th and the 24th of March, notably the closure of senegalese international borders and the declaration of a state of public health emergency.

Recall that the infectivity rate $\delta^{ci} = \beta^{ci}c$, where the parameter c is the rate of contacts per person per day, and β^{ci} the probability to be infected from a contact. Therefore, these are two parameters which depend on the decisions made by public authorities and the degree of compliance of the

population with them. In order to highlight the consequences of the definition and application of such measures, we simulate various scenarios.

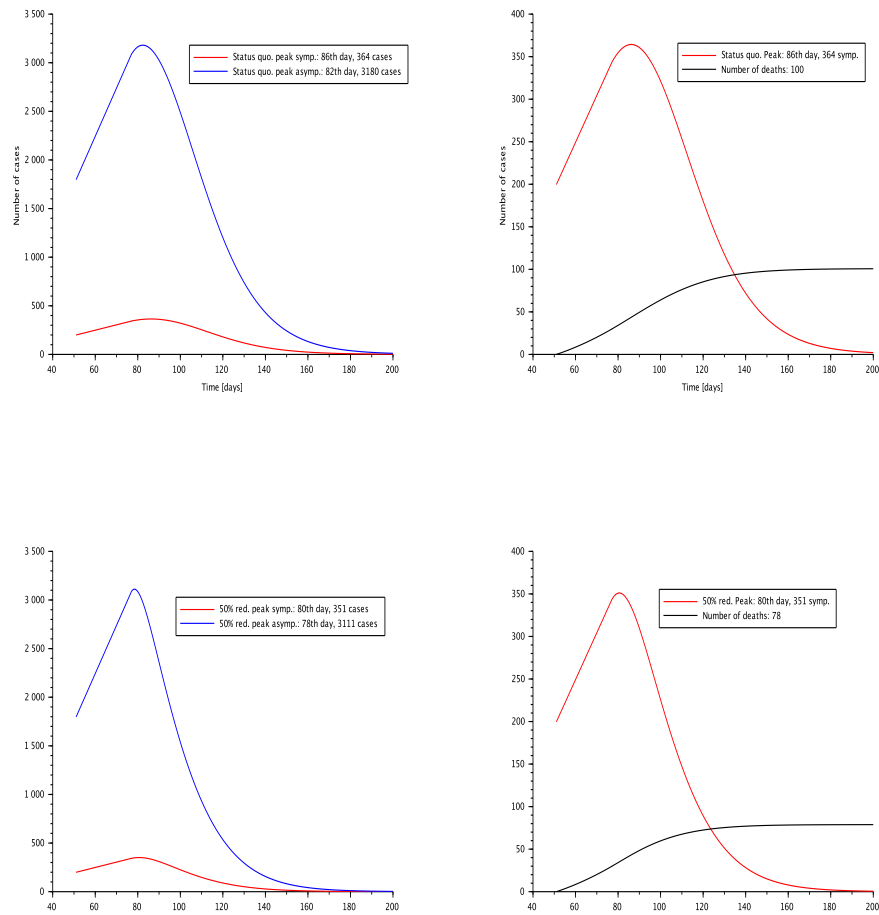
In the figure below, the curve represents the status quo relative to the trend between April 21 and May 17. Notice that the data keep on fitting the curve until about May 27. Beyond that date, all of a sudden, the status quo is broken because the data become much greater than the corresponding value of the curve. This phenomenon might be a delayed effect of the relaxation of the state of health emergency decided by the senegalese authorities on May 11 (71st day after the beginning of the pandemic in Senegal).



4.2. Post-fitting scenarios

Optimistic scenario: status quo and 50% reduction of the contamination rate in one month

Regarding the figures below, we mention that if the trend at the 77th day of the pandemic were maintained, the peak would be reached around the 86th day. One can also notice that reducing the infectivity rate by 50% would only bring the peak forward by one week, and reduce the number of deaths by 20%.



Note that the status quo is to stick to the tendency for the infectivity rate to exponentially decrease. In the sequel, we present simulations in which we gradually assume much less important decrease of the infectivity rate. In simulating these scenarios, we consider two strategies: 1. protecting the people at risk, 2. not protecting the people at risk.

Deviations from the status quo tendency

The simulations we have performed give an evidence that the less important the decrease of the infectivity rate is, the higher and the earlier the peak is. However, the price to pay is that the number of the deaths can be dramatically high. For instance, in case the infectivity rate is constant from the 77th day, the number of deaths is around 77000. This is in line with the worst scenario for Africa characterized by no interventions and no awareness of african people presented by the United Nations Economic Commission for Africa in its report in April (see [14]). This report exploiting the Imperial College COVID-19 Response Team simulations [4] predicts 3.3 million deaths in Africa in 2020, for this scenario. It's clear that Africa is not heading to this scenario, so far. Indeed, 3 months after the burst of the pandemic the continent has less than 5 thousand deaths.

Such a favourable situation is possibly due to the youthfulness of the african population. In the case of Senegal, the median age of the population is 19.4 years whereas the world's is 30.6 years. Indeed, the senegalese population is relatively young compared to most of other countries'. See the age pyramids below.

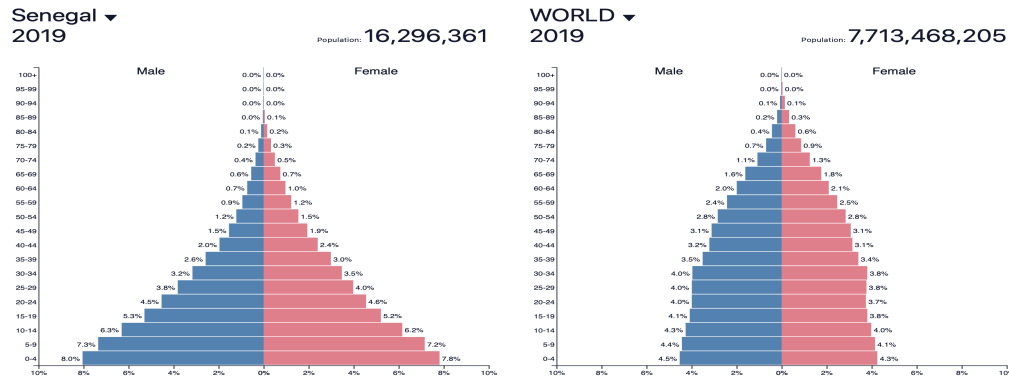


Figure 1: Source: <https://www.populationpyramid.net/world/2019>

Youthfulness as an asset for collective immunity

Considering the youthfulness of the senegalese population brings us to add one more optimistic but realistic scenario to the 4 presented by United Nations Economic Commission for Africa in its report on CoVid-19 [14]. The previsions of this report are inspired by the Imperial College COVID-19 Response Team simulation results [4] summarized in the table below.

26 March 2020

Imperial College COVID-19 Response Team

Table 1: Estimated impact of suppression strategies. The impact on infections and deaths over 250 days for two different suppression strategies triggered according to different thresholds for mortality incidence (0.2 and 1.6 deaths per 100,000 population per week).

	Unmitigated Scenario		Suppression at 0.2 deaths per 100,000 population per week		Suppression at 1.6 deaths per 100,000 population per week	
	Infections	Deaths	Infections	Deaths	Infections	Deaths
East Asia & Pacific	2,117,131,000	15,303,000	92,544,000	442,000	632,619,000	3,315,000
Europe & Central Asia	801,770,000	7,276,000	61,578,000	279,000	257,706,000	1,397,000
Latin America & Caribbean	566,993,000	3,194,000	45,346,000	158,000	186,595,000	729,000
Middle East & North Africa	419,138,000	1,700,000	30,459,000	113,000	152,262,000	594,000
North America	326,079,000	2,981,000	17,730,000	92,000	90,529,000	520,000
South Asia	1,737,766,000	7,687,000	111,703,000	475,000	629,164,000	2,693,000
Sub-Saharan Africa	1,044,858,000	2,483,000	110,164,000	298,000	454,968,000	1,204,000
Total	7,013,734,000	40,624,000	469,523,000	1,858,000	2,403,843,000	10,452,000

Table 1. CoVid-19 world previsions

By the way, regarding the resistance of young people to SARS-CoV-2, and the fact that the median age of Africa's population is only 19.7 years, the continent should define its specific strat-

egy by aiming to collective immunity along with the protection of aged people and other people at risk. It's worth noticing that only 5.3% of Africa's population are over 60 years old (see <https://www.populationpyramid.net/africa/2019/>), and that up to 94% of the deaths from the CoVid-19 pandemic in the world are over 60 years old (see [9]).

The youthful profile of the population makes it easier for Senegal to attain collective immunity because it has less (aged) people at risk. This results, for example, in the following simulations with various levels of protection for people at risk, while the rest of the population are totally exposed. Indeed, considering the fact that people over 60 years age represent 4.8% of the population, with different rates of infectivity, we perform a simulation in which: 1) there is no protection for people at risk ; 2) 90% of them are protected ; 3) 99% of them are protected.

	Infectivity +200%			Infectivity +100%			Infectivity +50%			Infectivity +25%		
Protection	0 %	90 %	99 %	0 %	90 %	99 %	0 %	90 %	99 %	0 %	90 %	99 %
Date of the peak	08-15 July 2020	10-18 July 2020	10-18 July 2020	July 28 - Aug 04 2020	July 31 - Aug 07 2020	July 31 - Aug 07 2020	16-24 Aug 2020	20-28 Aug 2020	20-28 Aug 2020	03-10 Sept 2020	08-15 Sept 2020	07-15 Sept 2020
Asymptomatics peak value	4.61 millions	4.49 millions	4.49 millions	3.10 millions	3 millions	3.01 millions	2.24 millions	2.16 millions	2.16 millions	1.77 millions	1.70 millions	1.70 millions
Symptomatics peak value	468955	471451	472449	341131	341077	343438	257548	256148	26993	208074	205979	206550
Average R0	4.259	4.270	4.267	2.940	2.944	2.958	2.27	2.27	2.27	1.919	1.920	1.920
Number of deaths	93590	9644	993	91122	9372	967	87247	8954	924	83549	8556	885
Number of immune	15.59 millions	15.65 millions	15.66 millions	14.86 millions	14.88 millions	14.91 millions	14.06 millions	14.06 millions	14.07 millions	13.39 millions	13.35 millions	13.37 millions

Table 2. CoVid-19 senegalese previsions: collective immunity

	Infectivity 00%			Infectivity -25%			Infectivity -50%			Infectivity -85%		
Protection	0 %	90 %	99 %	0 %	90 %	99 %	0 %	90 %	99 %	0 %	90 %	99 %
Date of the peak	01- 08 Oct 2020	08-09 Oct 2020	08-15 Oct 2020	27 Nov- 04 Dec 2020	08-16 Dec 2020	08-15 Dec 2020	12-18 June 2021	21-27 July 2021	19-26 July 2021	17-19 May 2020	17-19 May 2020	17-19 May 2020
Asymptomatics peak value	1.28 millions	1.22 millions	1.22 millions	771768	722359	723173	246063	211936	213033	3102	3095	3095
Symptomatics peak value	153472	150703	151179	93635	90293	90657	30185	26776	26993	348	351	351
Average R0	1.559	1.560	1.560	1.189	1.190	1.190	0.807	0.807	0.807	0.334	0.336	0.336
Number of deaths	77478	7904	819	66556	6732	702	42662	4152	444	79	34	30
Number of immune	12.35 millions	12.27 millions	12.28 millions	10.55 millions	10.39 millions	10.40 millions	6.73 millions	6.36 millions	6.38 millions	12280	12137	12148

Table 3. CoVid-19 senegalese previsions: reinforcing barrier measures

Simulation for various disinfection rates

The influence of the Infected objects on the system can be brought out by tuning their infectivity rate and the disinfection parameter. Recall that the infectivity rate of objects is tightly related to hand washing and face touching. Below are some simulations related to three levels of disinfection, in the context of a 10% increase of the infectivity rate.

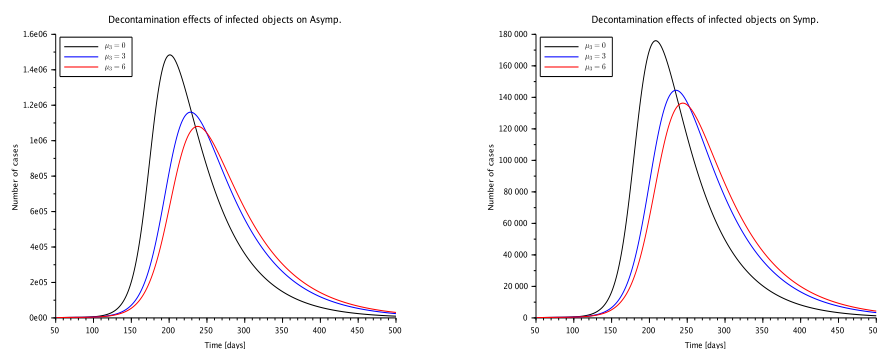


Figure 2: Number of infected objects

Note that infected objects affect the magnitude of the peak and its date of occurrence. This is an evidence of the relevancy of considering the objects compartment.

5. Conclusion

In order to comply with senegalese public health authorities response to the CoVid-19 pandemic, we have built a 6-compartment model including a quarantined people compartment. An infected objects compartment is considered not to neglect their impact in the propagation of the disease. The equation governing the number of objects variable has received a special treatment by imposing the decrease of contamination parameters as the number of infected objects increases. This condition ensures that recontamination of infected objects is not counted, which is a trick used to compensate for the lack of a susceptible objects compartment. The latter could hardly make sense since the number of all objects is infinity.

Then, in terms of analysis of the ODE system, using the barrier theorem, one easily proves the boundedness of solutions, and by the classical decomposition of the system, the basic reproduction number \mathcal{R}_0 is formally calculated.

And the last part of this work is devoted to numerical simulation of different scenarios, from the most optimistic to the most pessimistic in terms of day of occurrence and magnitude of the peak, and number of deaths. Based on the official daily data released by senegalese health authorities, from the 15th to the 77th day after the burst of the pandemic, by fitting, we estimate epidemiologic parameters such as the infection and contact rates. Then, from the 77th day of the pandemic, we have simulated the different possible scenarios depending on the behavior of the population.

The scenarios presented allow to quantify the impact of the application of the barrier measures taken by public authorities. For instance, the observation of physical distancing measures, the wearing of masks, hand washing and objects disinfection are variation factors of the infectivity rate. Here are a few observations we have made:

- If the status quo were maintained from May 17 on, say the infectivity rate decreases with the same velocity as the period of time April 21 - May 17 2020, the peak would be attained about May 26.
- There is a deviation from the trend of the period April 21 - May 17 leading to a higher and later peak.
- The gradual increase of the infectivity rate allows to bring out advantages and drawbacks of collective immunity. The observation of the different scenarios shows that increasing the infectivity rate without accompanying measures: brings the peak earlier; puts the peak higher; causes more deaths (see Tables 2 and 3).
- The ideal scenario in terms of number of deaths, early occurrence of the peak, and collective immunity, consists on:
 - taking measures for protecting only the aged people and all the people with comorbidity, from the virus.
 - cancelling all other barrier measures for attaining collective immunity,
- The decontamination of objects affects significantly the magnitude and the date of occurrence of the peak.

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