

S1 Appendix for

Covid-19 in Africa: underreporting, demographic effect, chaotic dynamics, and mitigation strategies impact

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Appendix S1 - Estimating the contact number

A. Averaged number of contact estimated from the new cases

The governing equations of the observed dynamics are generally unknown. Here, we use a simplified formulation of the epidemic to explain how to reconstruct the efficacy of the non pharmaceutical measures from a single variable commonly observed under real conditions: the number of new cases per day $I(t)$. It will be later applied to the dynamics of the SEi²RD model (see S2 Appendix) in order to investigate if, although based on a simple formulation, this formulation can apply to a dynamics of higher complexity.

The SEIRD model is commonly used in epidemiology where S stands for Susceptible, E for Exposed and i for infected, R for recovered and D for dead people at time t :

$$\begin{cases} \frac{dS_t}{dt} = -\frac{\beta}{N_t} S_t i_t \\ \frac{dE_t}{dt} = +\frac{\beta}{N_t} S_t i_t - \alpha E_t \\ \frac{di_t}{dt} = +\alpha S_t i_t - \nu i_t - m i_t \\ \frac{dR_t}{dt} = +\nu i_t \\ \frac{dD_t}{dt} = +m i_t, \end{cases} \quad (1)$$

and $N_t = S_t + E_t + i_t + R_t$ the total population at time t . Considering that D is only a by product of the SEIR model and that the variations of N are very small (and also slow) in comparison to the other variations, this dynamics can be reduced to the three equations

$$\begin{cases} \frac{dS_t}{dt} = -\frac{\beta}{N_t} S_t i_t \\ \frac{dE_t}{dt} = +\frac{\beta}{N_t} S_t i_t - \alpha E_t \\ \frac{di_t}{dt} = +\alpha S_t i_t - \nu i_t - m i_t, \end{cases} \quad (2)$$

and then, assuming that the exposure plays as a simple delay between the infection stage and infected stage, the system can be rewritten in two equations

$$\begin{cases} \frac{dS_t}{dt} = -\frac{\beta}{N} S_t i_t \\ \frac{di_t}{dt} = +\frac{\beta}{N} S_{t-\tau} i_{t-\tau} - \nu i_t - m i_t, \end{cases} \quad (3)$$

with τ the average time delay before an exposed people becomes infected (be he symptomatic or not).

The number of daily new cases $I_t^{(1)}$ (where the number in bracket denotes the first derivative) is thus given by

$$I_t^{(1)} = + \frac{\beta}{N} S_{t-\tau} i_{t-\tau}, \quad (4)$$

and its cumulative number by

$$I_t^{(0)} = \int + \frac{\beta}{N} S_{t-\tau} i_{t-\tau} dt. \quad (5)$$

Rewriting the first equation of Eqs. (3) into

$$S_{t-\tau} = - \frac{N}{\beta i_{t-\tau}} \frac{dS_{t-\tau}}{dt},$$

we get

$$S_{t-\tau} = S_{t=0} - I_t^{(0)}, \quad (6)$$

by integrating Eq. (5). Replacing Eq. (6) in Eq. (4) we get

$$I_t^{(1)} = + \frac{\beta}{N} (S_{t=0} - I_t^{(0)}) i_{t-\tau}. \quad (7)$$

The exposure ratio can thus be written

$$\beta = \frac{N I_t^{(1)}}{(S_{t=0} - I_t^{(0)}) i_{t-\tau}}. \quad (8)$$

This formula remains valid locally if β is varying with time. As i_t is not always available, it can be estimated from $I_t^{(1)}$ such as

$$\hat{i}_t = \int_{t-T}^t I_t^{(1)} dt, \quad (9)$$

with T the characteristic time of the disease, so that $\hat{\beta}(t)$ can be entirely deduced from the single variable $I_t^{(1)}$. Taking into account the number of vaccination V_t at time t (if available), we get

$$\hat{\beta}(t) = \frac{N I_t^{(1)}}{(S_{t=0} - I_t^{(0)} - V_t) i_{t-\tau}}, \quad (10)$$

with $i_{t-\tau} \neq 0$. Note that i_t is composed of both symptomatic and asymptomatic compartment here ($i_t = i_t^S + i_t^A$) and may thus be difficult to estimate practically since asymptomatic people are by definition difficult to count. However, this limitation will only have a marginal effect on the estimate of $\hat{\beta}(t)$ because it will act similarly onto $I_t^{(1)}$. To avoid singularities, $\hat{\beta}(t)$ is estimated only if $i_{t-\tau} \geq 100$. This formulation was tested on synthetic cases and revealed a systematic overestimation (in the range 1.28 to 1.42) when varying the infected period T_i from 5 to 7.5 days (see Table A in the present S1 Appendix). A correction coefficient $\xi = 0.75$ such as

$$\hat{\beta}_t = \xi \beta_t \tag{11}$$

was thus applied to get more realistic estimates of the average number of contact per person per time.

Table A:

T_i	4.5	4.75	5	5.5	6	6.5	7	7.25	7.5
τ	3.2	3.4	3.6	3.9	4.3	4.6	5.0	5.2	5.4
ξ	0.63	0.67	0.70	0.73	0.78	0.78	0.78	0.75	0.70
$1/\xi$	1.59	1.50	1.42	1.37	1.28	1.28	1.28	1.33	1.42

B. Recovery and Mortality rates

The recovery rate ν and the mortality rate m can also be deduced from i_t (deduced from $I_t^{(1)}$ using Eq. 9) and $R_t^{(1)}$ (to be estimated numerically from R_t)

$$\hat{\nu}(t) = \frac{R_t^{(1)}}{i_t}, \quad (12)$$

or $D_t^{(1)}$ (to be estimated numerically from D_t)

$$\hat{m}(t) = \frac{D_t^{(1)}}{i_t}, \quad (13)$$

Here also, the contribution of the asymptomatic cases are compensated.

C. Effective reproduction number

Since the original dynamics is a priori unknown, three formulations of the basic reproduction number \mathbf{R}_0 were investigated, based on SIR model

$$\mathbf{R}_0^{\text{SIR}} = \frac{\beta}{\nu}, \quad (14)$$

SIRD model

$$\mathbf{R}_0^{\text{SIRD}} = \frac{\beta}{\nu + m}, \quad (15)$$

and SEi2RD model hypotheses:

$$\mathbf{R}_0^{\text{SEi2RD}} = \frac{\beta}{\nu + m} + \frac{(1-p)m}{\nu + m}. \quad (16)$$

The three formulations were tested to estimate the effective reproduction number

$$\mathbf{R}_t = \mathbf{R}_0 \times \frac{(S_{t=0} - I_t^{(0)} - V_t)}{N_t}, \quad (17)$$

the SIRD formulation was found the more realistic when applied to the synthetic data. The following formulation

$$\hat{\mathbf{R}}_t = \frac{\beta_t}{\nu_t + m_t} \times \frac{(S_{t=0} - I_t^{(0)} - V_t)}{N_t}, \quad (18)$$

was thus preferred – in practice – to estimate the effective reproduction number. Other formulations were kept to estimate the error resulting from this lack of knowledge.