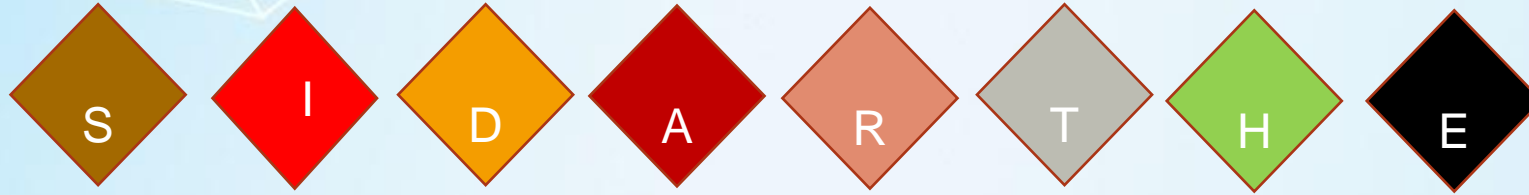


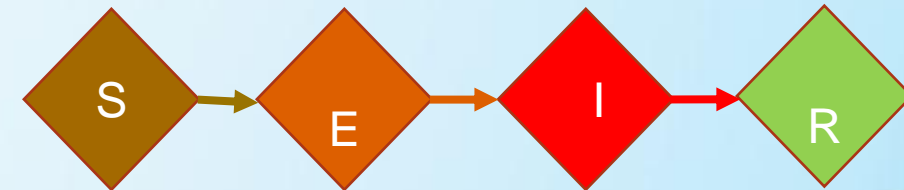
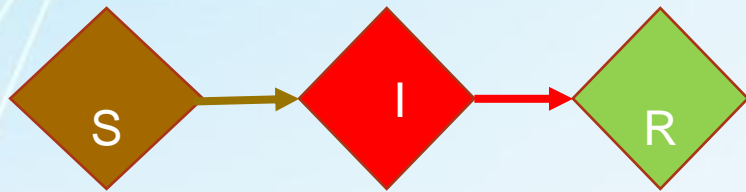
# A STUDY OF COVID-19 DATA FROM AFRICAN COUNTRIES



**Presenter: Somiealo Azote**

20/07/2021

**Project initiated and supervised by**



**ASP 2021**

# My bio

- 2018: PhD in theoretical Biophysics from Stellenbosch University, S. Africa
- 12/2020-Now: On going postdoctoral studies, Ca Foscari University of Venice
- Interest in Computing, Simulation, Data Analysis/Science
- Passionate in mentoring young women in science (Member of AIMSWSIS Mentoring pr)
- The After ASP: stay connected



## ASP 2021

# Outlines

## ➤ Introduction

- Global COVID-19 Situation
- COVID-19 Situation in Africa
- Impact on the African School of Physics (ASP)

## ➤ ASP COVID-19 Study

## ➤ Results










## ➤ Conclusions



# Global COVID-19 Situation

➤ COVID-19 started in China and has spread to the entire world within a few months

[WHO Coronavirus \(COVID-19\) Dashboard](#) (15 July 2021)

Name	Cases - cumulative total	Cases - newly reported in last 24 hours	Deaths - cumulative total	Deaths - newly reported in last 24 hours	Transmission Classification
Global	188,128,952	516,528	4,059,339	8,547	
Americas	74,239,667	147,561	1,949,777	3,790	
Europe	57,472,767	130,355	1,200,672	1,068	
South-East Asia	36,400,628	126,002	520,095	2,086	
 United States o...	33,604,822	32,107	602,820	411	Community transmission
 India	30,987,880	41,806	411,989	581	Clusters of cases
 France	5,714,131	8,747	110,446	6	Community transmission
 Turkey	5,500,151	6,907	50,367	43	Community transmission
 The United Kin...	5,233,211	41,748	128,530	49	Community transmission
 Argentina	4,682,960	20,023	99,640	385	Community transmission
 Colombia	4,548,142	17,532	113,839	504	Community transmission
Africa	4,495,778	30,155	105,354	683	
 Italy	4,275,846	2,153	127,831	23	Clusters of cases
 Spain	4,041,474	14,929	81,043	0	Community transmission

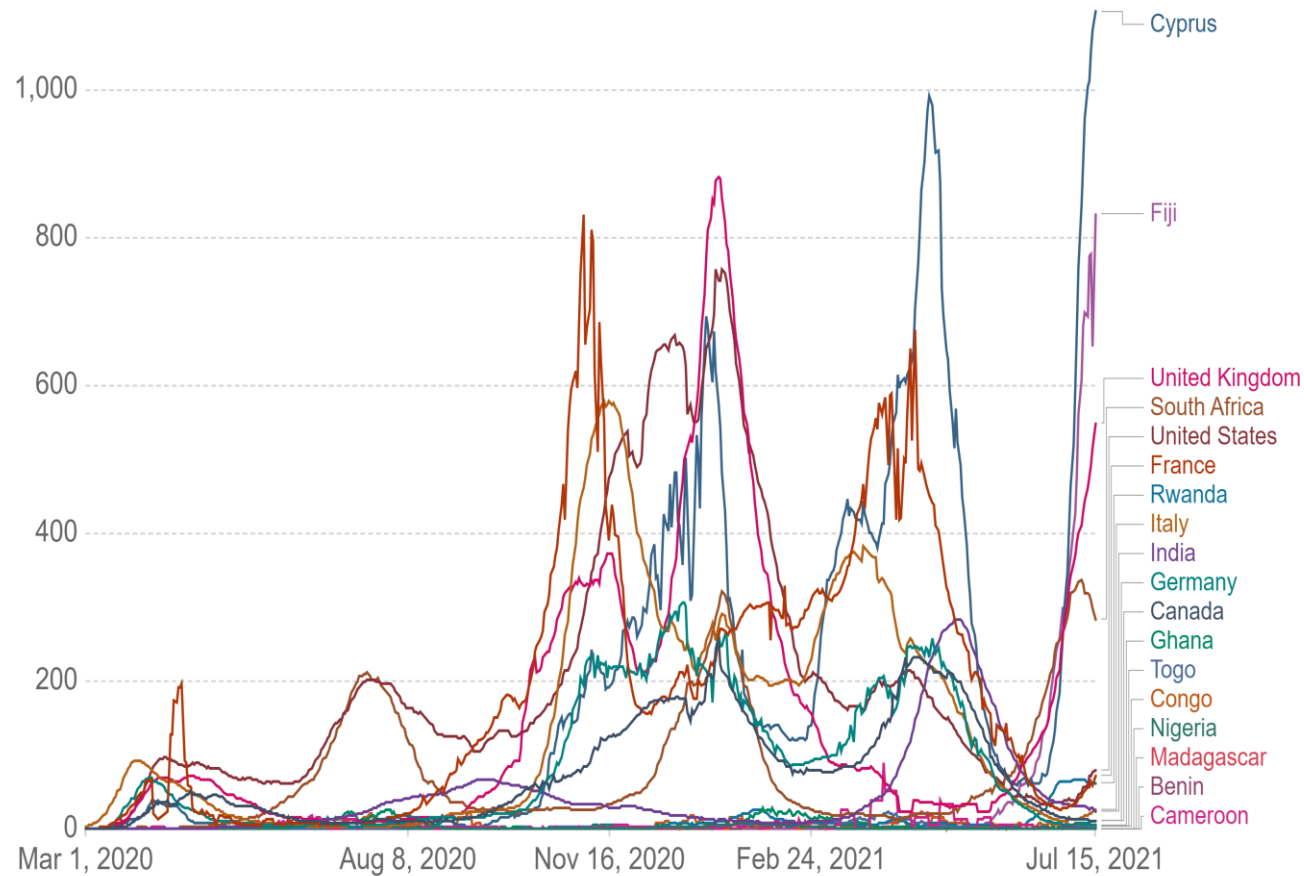
➤ Weekly epidemiological update on COVID-19 by WHO

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# Comparison with Africa

## Daily new confirmed COVID-19 cases per million people

Shown is the rolling 7-day average. The number of confirmed cases is lower than the number of actual cases; the main reason for that is limited testing.



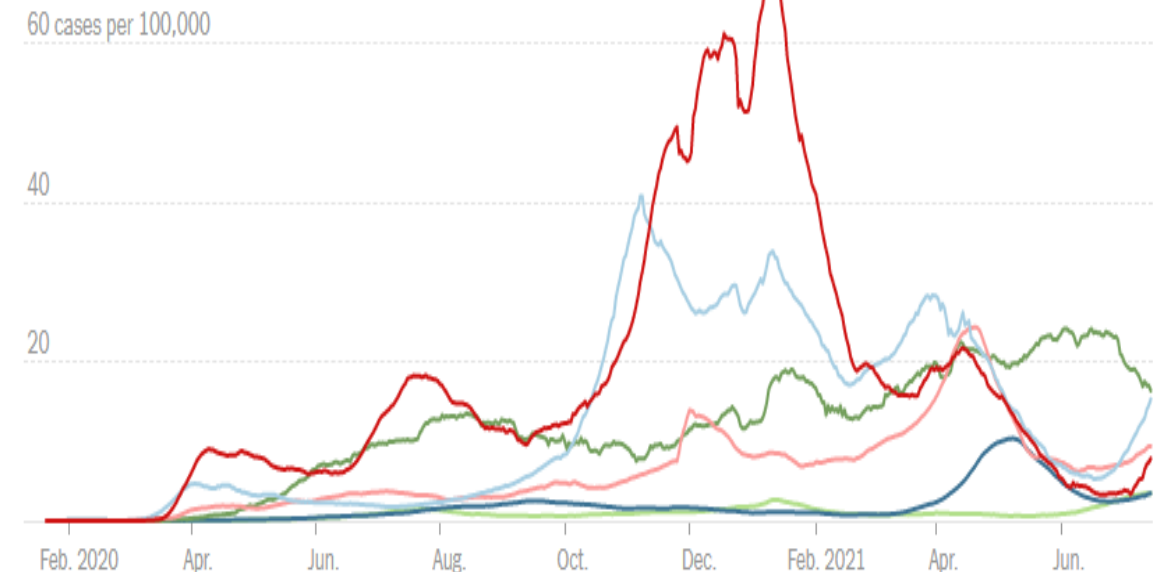
Source: Johns Hopkins University CSSE COVID-19 Data

Our World  
in Data

## Cases by region

This chart shows how cases per capita have changed in different parts of the world.

Africa Asia-Pacific Europe Latin America Middle East U.S. and Canada



About this data:

Sources: Center for Systems Science and Engineering at Johns Hopkins University and state and local health agencies (cases); World Bank and U.S. Census Bureau (population data).

CC BY

# COVID-19 Situation in Africa

- WHO prediction:
  - 29-44m Africans could be infected
  - 83-190000 Africans could die
  - Suggests Africa has lower transmission rates
  - Prolong outbreak for several years
  - Pressure on economic resources
  - Containment measures challenge: 60% living below poverty line
- Current Situation:
  - 4.5m infected
  - 106,078 dead
  - Millions rendered jobless, increased poverty & insecurity
  - A number of studies conducted
- ASP COVID-19 Study: Contribute to understanding & containment

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# Impact on African School of Physics

- ASP2020, initially planned for July 2020 in Morocco was postponed and only taking place now online
- In the meantime, ASP set up an online lecture series, since May 2020 – Twice a week
- In March 2020, Dr. Keteve A. Assamagan made a call to all the ASP alumni for Covid 19 exercise: to study and help understand Covid 19 data from African countries
- A number of ASP alumni responded to the call with the permission of our supervisors
- From there started the study – Alumni were in charge to collect Covid data from their countries – the study was guided mentored by senior researcher and leaders of ASP

# Research Team

## Team Leader



**Dr. Kétévi A. Assamagan**

Togo & USA

**Experimental particle physicist at Brookhaven National Laboratory**

**Fellow, African Academy for Sciences**

**Research Interests: Search for new physics beyond the Standard Model of particle physics**

**Alma Mater: University of Lomé & University of Virginia**

**Co-founder, ASP, ASFAP**

**Passionate: Physics Education, Communication & Outreach**

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# Research Team



**Prof. Simon Connell**

**South Africa**

**Professor of Physics**

**University of Johannesburg**

**Research Interests: nuclear & particle physics**

**Alma Mater: University of Witwatersrand**

**Passionate: Technology & Innovation**



**Prof. Azwinndini Muronga**

**South Africa**

**Executive Dean of Science**

**Nelson Mandela University**

**Interests: Research, Training & Innovation in Science**



# Research Team



**Dr. Dephney Mathebula**  
**South Africa**  
**Senior Lecturer,**  
**Biomathematics**  
**University of**  
**Venda**



**Dr. Laza Rakotondravohitra**  
**Madagascar**  
**Resident**  
**Duke University**  
**Medical Center,**  
**USA**



**Aluwani Guga**  
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**MSc Physics**  
**Nelson Mandela**  
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**Kondwani C. Mwale**  
**Malawi**  
**Masters of Physics**  
**University of**  
**Rwanda, Rwanda**  
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**Dr. Kossi Amouzouvi**  
**Togo**  
**Lecturer**  
**KNUST, Ghana**  
**APS2014**



**Essossolim C. Haliya**  
**Togo**  
**PhD, University of**  
**Abomey-Calavi,**  
**Benin**  
**ASP2018**



**Ebode Fabien Onyie**  
**Cameroon**  
**PhD**  
**University of**  
**Yaounde**  
**ASP2016**



**George Zimba**  
**Zambia**  
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**University of**  
**Jyväskylä, Finland**  
**ASP2016**



**Dr. Fenoso Fanomezana**  
**Madagascar**  
**PostDoc**  
**University of**  
**Antananarivo**

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# Research Team



**Ann Njeri,**  
**Kenya**  
**3<sup>rd</sup> PhD (Astrophysics) Student**  
**Jodrell Bank Centre for**  
**Astrophysics**  
**University of Manchester, UK**



**Toivo S. Mabote**  
**Mozambique**  
**Undergrad Student**  
**(BSc)**  
**Eduardo Modlane**  
**University**  
**ASP2020**



**Dr. Somiealo Azote**  
**Togo**  
**Posdoc**  
**Ca Foscari University of Venice,**  
**Italy**  
**ASP 2016**



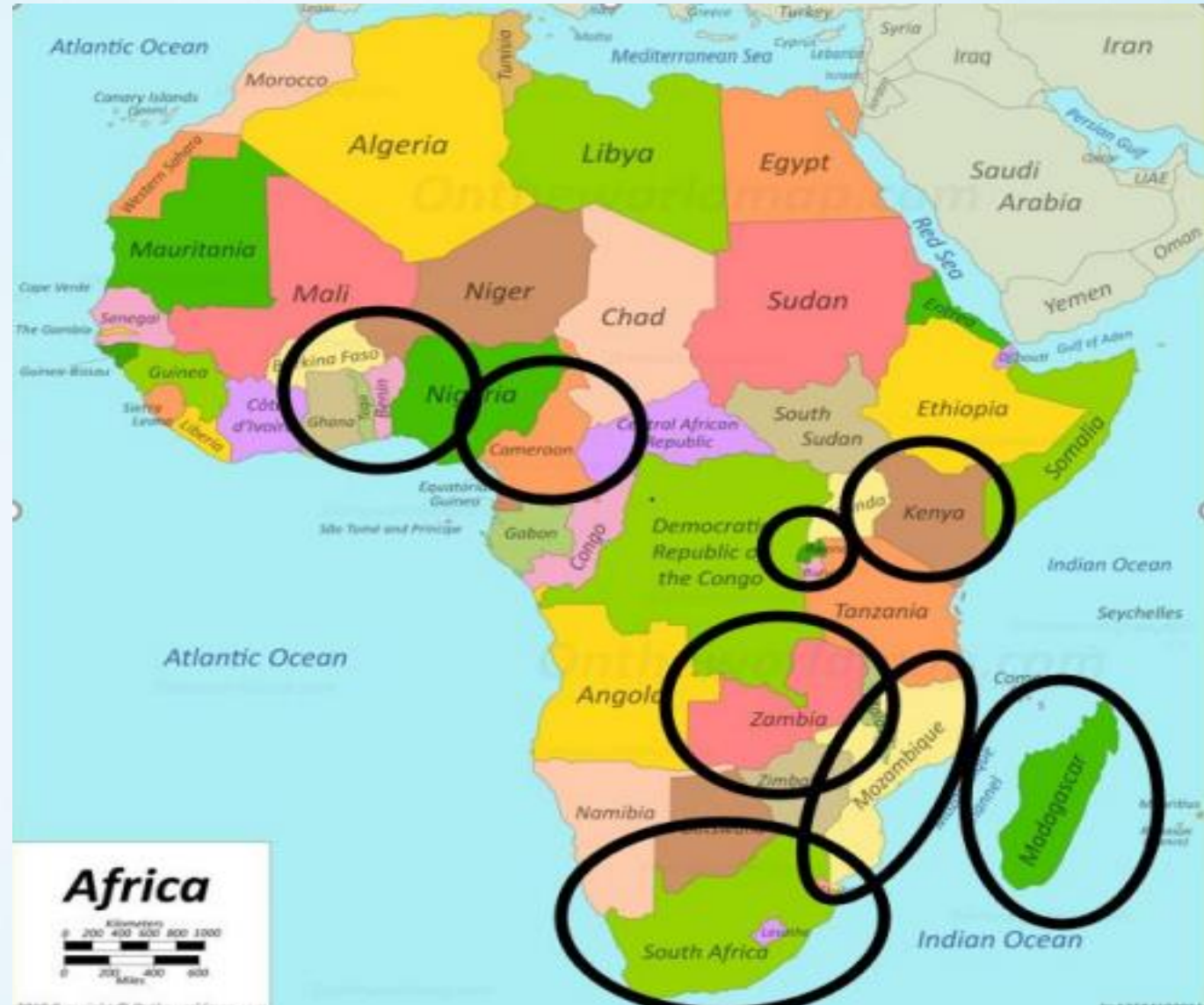
**Francisco Macucule**  
**Mozambique**  
**Undergrad Student**  
**(BSc)**  
**Eduardo Mondlane**  
**University**



# Countries we studied

➤ The countries that we studied are:

- Benin
- Cameroon
- Ghana
- Kenya
- Madagascar
- Mozambique
- Rwanda
- South Africa
- Togo
- Zambia



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# Epidemiological models

- ❖ Deterministic models also known as compartmental models are simpler and yet efficient tool to predict and control the spread of the disease through individuals.

Mathematical  
(epidemic) modeling

- ❖ Stochastic modelling forecasts the probability of various outcomes under different conditions, using random variables (Convenient for small infection

Deterministic  
Discrete or  
continuous time

Stochastic  
Discrete or  
continuous time and  
space

*Which one of the deterministic  
model is best for the analysis  
of the Covid-19 data from  
African countries?*

SIR

Susceptible-Infected-  
Recovered

SEIR

Exposed before getting infected

SEI

Have infectious force in the  
latent period

SEIRS

Recovered people can become  
infected again, Immunity not  
guaranteed

SIDARTHE

8 stages compartmental model,  
more complex.

DTMC

Discrete Time Markov Chains

CTMC

Continuous Time Markov Chains

SDE

Stochastic Differential equation

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# Deterministic (compartmental) models

- These models are dynamic and can be used to :
  - make predictions such as the duration of an epidemic or pandemic.
  - how the disease spread over time
  - Estimate various epidemiological parameters such as the reproduction number
  - Help estimates the fraction of population that need to be vaccinated
  - Show how different public health or government interventions may affect the outcome of the pandemic
- Susceptible-Infected-Recovered (SIR) is the simplest compartmental model
  - It has 3 groups S-I-R and each of them represent specific stage of the epidemic and fluctuates over time
  - Individuals may progress between compartments
  - It serve as the basis for all other compartmental models

# SIR model without vital dynamic

- For specific disease,  $S = S(t)$ ,  $I = I(t)$ ,  $R = R(t)$  are number of susceptible, Infected, recovered in a population  $N$  at time  $t$
- The number of Susceptible going from  $S$  to  $I$  category is:  $\frac{\beta SI}{N}$  i.e on average, each infected person generates  $\beta s(t)$  new infected persons.  $\beta$  quantifies the number of contacts between susceptible and infected people
- $\gamma I$  is the rate at which infected person recover. If an individual is infectious for an average time period  $D$ , then  $\gamma = 1/D$
- Normalization:  $s(t) = \frac{S(t)}{N}$ ,  $i(t) = \frac{I(t)}{N}$  and  $r(t) = \frac{R(t)}{N}$  such that  $s(t) + i(t) + r(t) = 1$
- SIR Model is then described by the following set of equations:

- **The rate of change in the susceptible population**

1)  $\frac{ds}{dt} = -\beta s(t)i(t)$

- **The rate of change in the infected population**

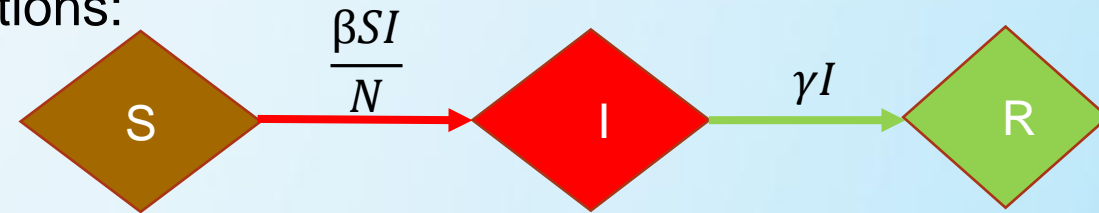
2)  $\frac{di}{dt} = \beta s(t)i(t) - \gamma i(t)$

- **The rate of change in the recovered population**

3)  $\frac{dr}{dt} = \gamma i(t)$

**Check that**  $\frac{ds}{dt} + \frac{dr}{dt} + \frac{di}{dt} = 0$

- Initial conditions:  $s(0) = s_0$ ,  $i(0) = i_0$ ,  $r(0) = r_0$  and  $i + s + r = 1$  (conservation)



- Equations (1,2,3) are non-linear coupled ordinary differential equations.
- Can thus be solved only numerically

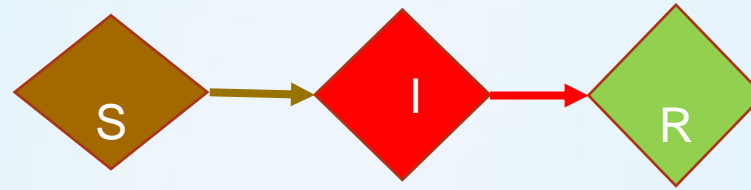
# Reproduction number

- The disease becomes a pandemic when the infection keeps on increasing
- meaning  $\frac{di}{dt} = \beta s(t)i(t) - \gamma i(t) > 0$  equivalently  $\frac{\beta}{\gamma} s(t) > 1$ .
- Since  $s(t) < 1$ , for a pandemic,  $\frac{\beta}{\gamma} > 1$ .
- The ratio  $R_o = \frac{\beta}{\gamma}$ , dimensionless, is called the reproduction number.
- So higher the reproduction number faster the disease going to spread.
- If  $R_o < 1$  then the disease will die out soon.
- Physical meaning:  $R_0$  represent the total number of secondary infections produced by a single infection in a completely susceptible population.

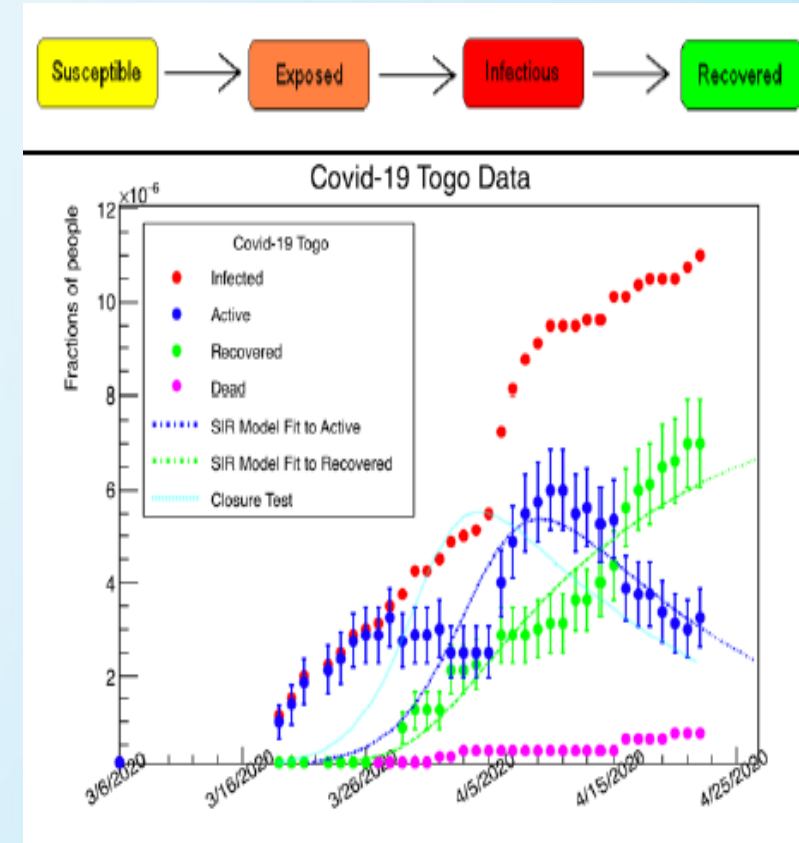
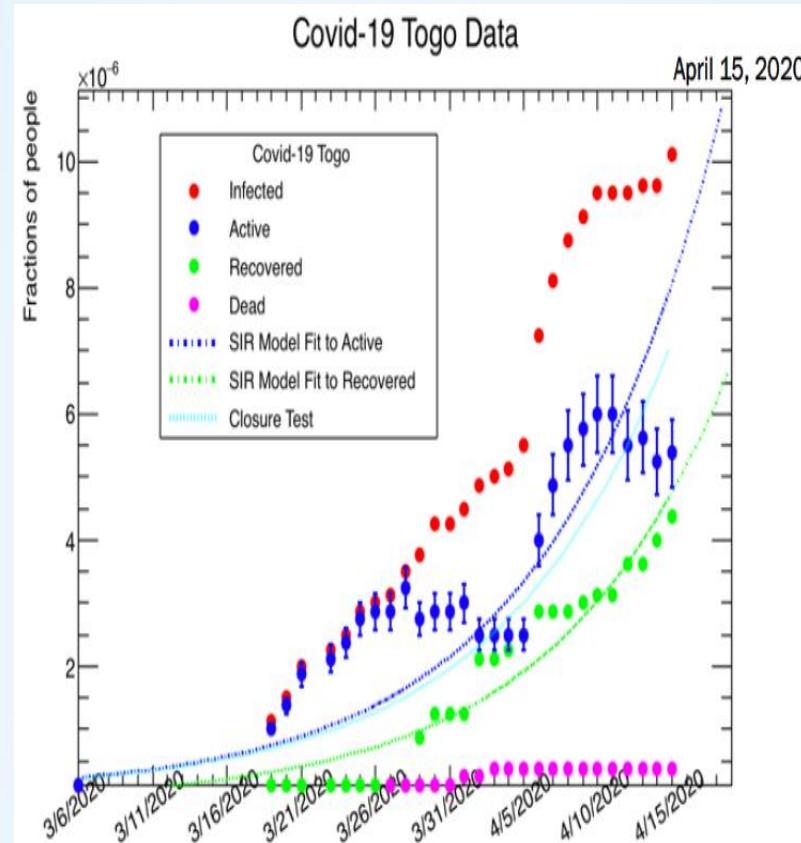


# Fitting SIR and SEIR models to Covid-19 data from Togo

- Fitting the recovered and the active independently also gives 2 different model parameters from fits to 2 independent datasets.



- Closure Test: using the 2 model parameters obtained by fitting the recovered to test for the active dataset, gives light blue curve, instead of the blue curve.
- Fitting only 1 dataset does not result in a good modeling on the other dataset.

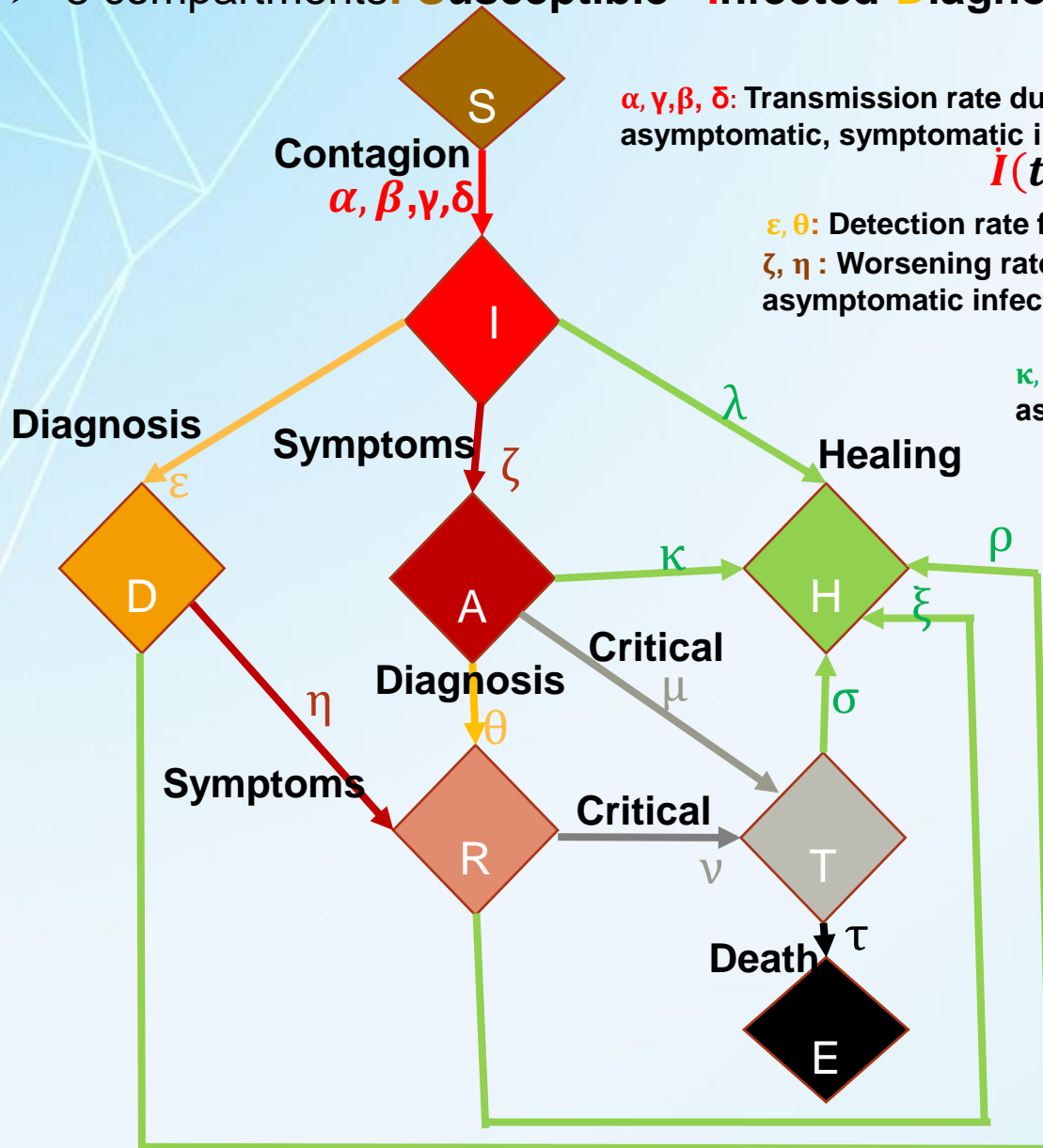


- Models unsuccessful in fitting the 3 curves simultaneously

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# SIDARTHE model formulation: 8 ODEs and 16 parameters

- 8 compartments: **S**usceptible – **I**nfected – **D**iagnosed – **A**iling – **R**ecognized – **T**hreatened – **H**ealed – **E**xtinct



$\alpha, \gamma, \beta, \delta$ : Transmission rate due to contact with UNDETECTED asymptomatic, symptomatic infected, DETECTED asymptomatic, symptomatic infected respectively.

$$\dot{S}(t) = -S(t)(\alpha I(t) + \beta D(t) + \gamma A(t) + \delta R(t)) \quad (1)$$

$$\dot{I}(t) = S(t)(\alpha I(t) + \beta D(t) + \gamma A(t) + \delta R(t)) - (\varepsilon + \zeta + \lambda)I(t) \quad (2)$$

$\varepsilon, \theta$ : Detection rate for ASYMPTOMATIC, SYMPTOMATIC respectively

$\zeta, \eta$ : Worsening rate, UNDETECTED asymptomatic infected becomes symptomatic ; DETECTED asymptomatic infected becomes Symptomatic

$$\dot{D}(t) = \varepsilon I(t) - (\eta + \rho)D(t) \quad (3)$$

$\kappa, \lambda, \xi, \rho$ : Recovery rate for asymptomatic infected, undetected symptomatic, detected asymptomatic, symptomatic infected respectively.

$$\dot{A}(t) = \zeta I(t) - (\mu + \kappa + \theta)A(t) \quad (4)$$

$\mu$ : Worsening rate, UNDETECTED symptomatic infected develop life-threatening symptoms.

$$\dot{R}(t) = \eta D(t) + \theta A(t) - (\nu + \xi)R(t) \quad (5)$$

$\nu$ : Worsening rate, DETECTED symptomatic infected develop life-threatening symptoms.

$$\dot{T}(t) = \mu A(t) + \nu R(t) - (\sigma + \tau)T(t) \quad (6)$$

$\tau$ : Mortality rate for infected with life-threatening symptoms

$\sigma$  Recovery rate life-threatening symptomatic infected

$$\dot{H}(t) = \lambda I(t) + \sigma T(t) + \kappa A(t) + \rho D(t) + \xi R(t) \quad (7)$$

$$\dot{E}(t) = \tau T(t) \quad (8)$$

<https://www.nature.com/articles/s41591-020-0883-7.pdf>

# Basic Reproduction number

- The model is initially developed to study the dynamics of Covid-19 in Italy.
- The model parameters are updated over time as the intervention measures or restriction increase. And the basic reproduction number computed
- The basic reproduction number from SIDARTHE model

$$R_0 = \frac{\alpha}{r_1} + \frac{\beta * \varepsilon}{r_1 * r_2} + \frac{\zeta * \gamma}{r_1 * r_3} + \frac{\eta * \delta * \varepsilon}{r_1 * r_2 * r_4} + \frac{\delta * \zeta * \theta}{r_1 * r_3 * r_4} \quad \text{with} \quad \begin{aligned} r_1 &= \varepsilon + \zeta + \lambda \\ r_2 &= \eta + \rho \\ r_3 &= \mu + \theta + \kappa \\ r_4 &= \xi + \nu \end{aligned}$$

- $R_0 > 1$ : epidemic continues
- $R_0 < 1$ : outbreak ends
- $R_0$  can be used to estimate the fraction of people to vaccinate
- $R_0$  must be applied with caution due to model dependency

# Analysis of Covid-19 data from 10 African countries

- We adapted the SIDARTHE model to consider the containment measures ( social distancing, mask wearing, curfew, Gatherings and recreational activities, hand sanitizing) taken by African countries
- For each country, we modeled the three datasets of active, recovered and death cases by finding the SIDARTHE model parameters that best match the time-evolution of the data
- The best-matched parameters may change over time, thus providing an evolution of the basic reproduction number as a function of time
- We propagated to the  $R_0$  estimates the statistical uncertainties related to the numbers of tests done and cases identified



# Statistical and Systematic uncertainties

- We also applied a systematic uncertainty to the  $R_0$  estimates based on the fraction of the infected but unaffected population

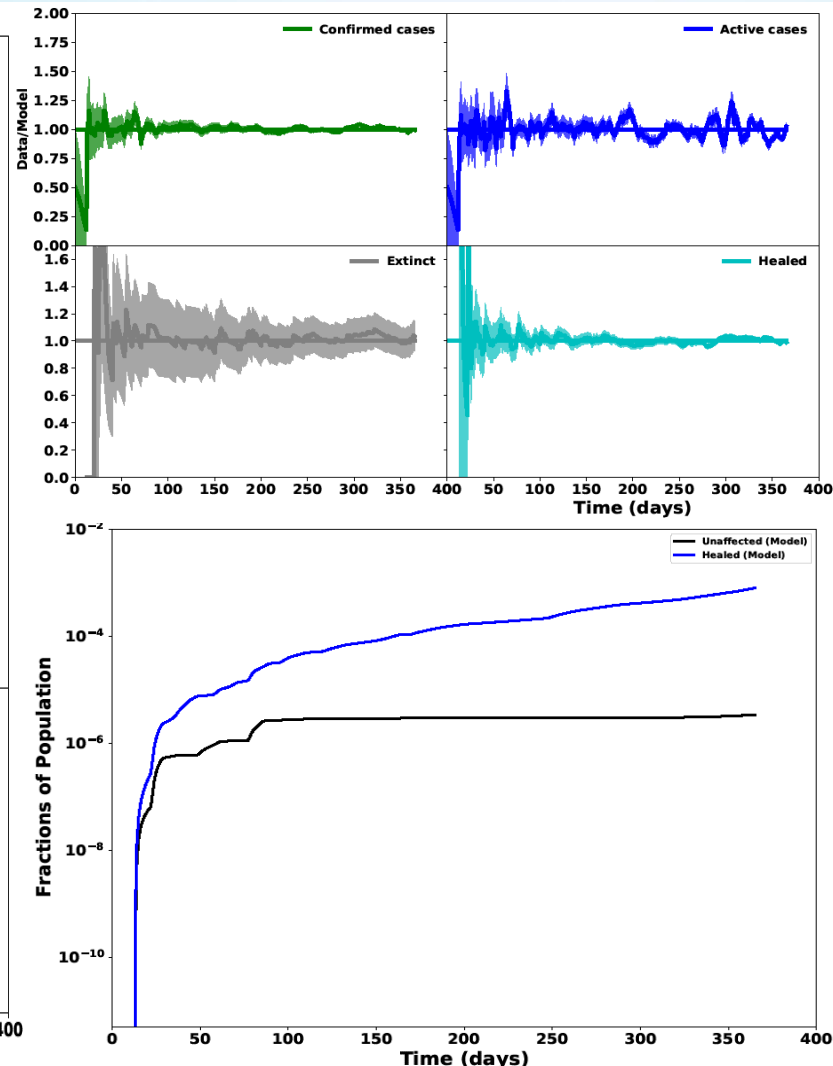
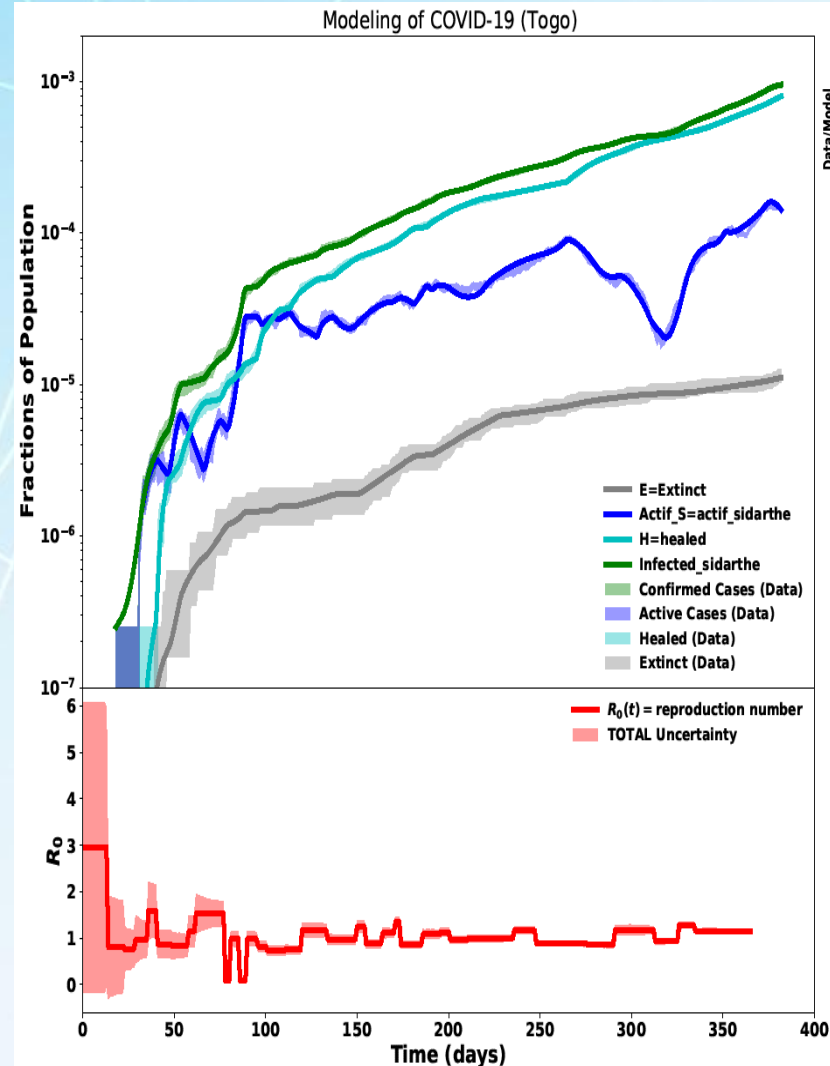
- SIDARTHE Model: rate of change in the recovered population

$$\dot{H}(t) = \lambda I(t) + \sigma T(t) + \kappa A(t) + \rho D(t) + \xi R(t)$$

The number of undetected asymptomatic persons who are healed are not included in the data, yet it is included in the model → Systematic error on the modeling  
Systematic Uncertainties on  $R_0$ .

# Analysis of Covid 19 data from Togo

## ➤ 1st Case: 6 March: Case travelled to Germany, France, Turkey & Benin

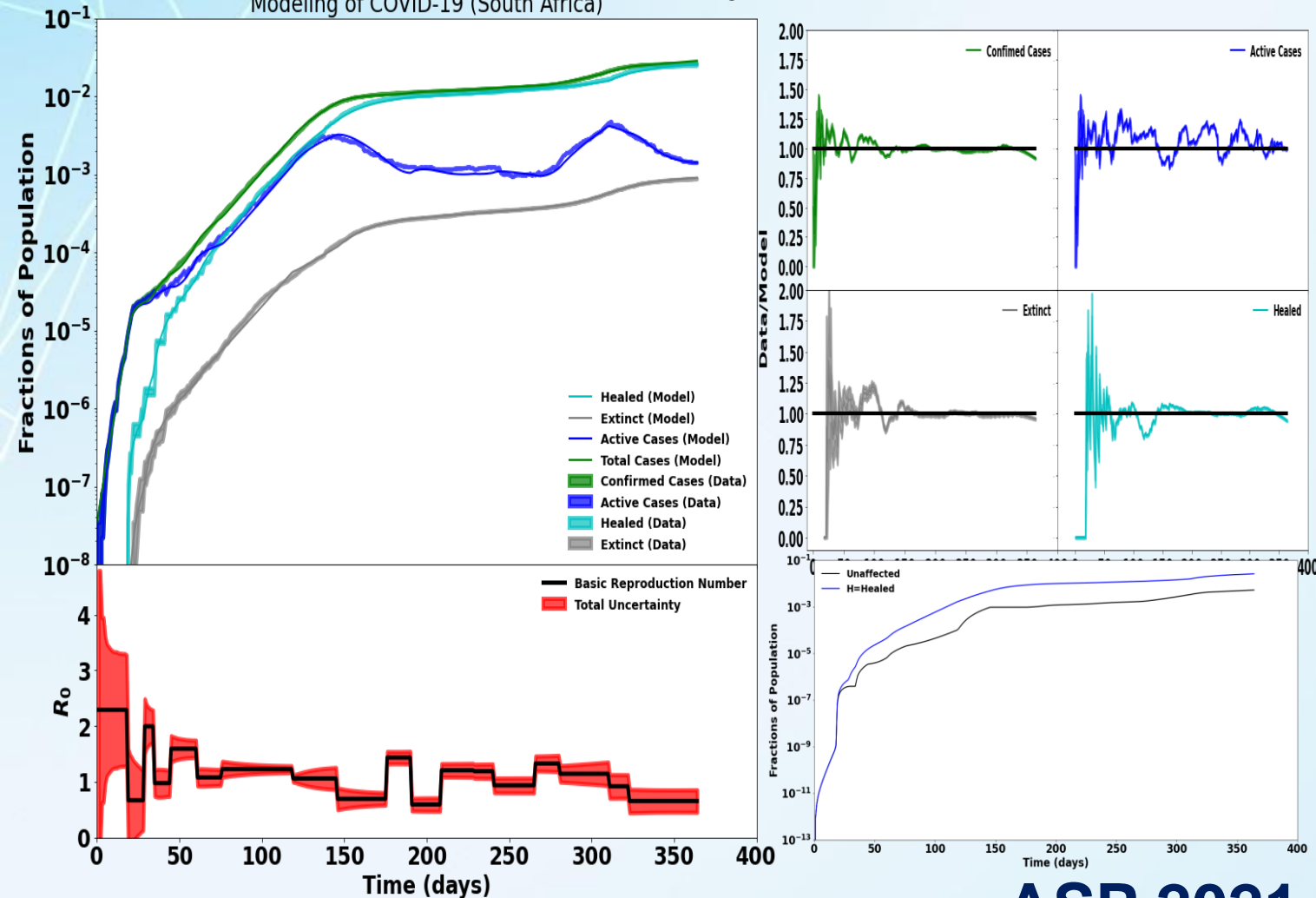


- Surveillance at PoEs
  - Flights to Europe suspended ~10 days
  - Borders closed 2 weeks later
  - Curfew introduced
  - Public places e.g schools closed
  - Social gatherings restricted
  - Massive test drive from 07 April
  - 5-20 May cases increased:
- Neighboring countries reopened borders-Togolese returned home
- 9 June 2020, curfew lifted, school resumed
  - 1st August flight, schools resumed
  - End of year holiday season, despite the curfew and all the preventive measures, the number of cases increased substantially.
  - Violations of the preventive and social distancing measures

# Analysis of Covid 19 data from South Africa

- The first Covid 19 case on March 5, 2020, which arose due to a returning traveler from Italy
- 5 different levels of lockdown regulations : level 1 most relaxed and alert level 5 the strictest
- There is a correlation between  $R_0$  and the implementation of the various lockdown levels

Modeling of COVID-19 (South Africa)



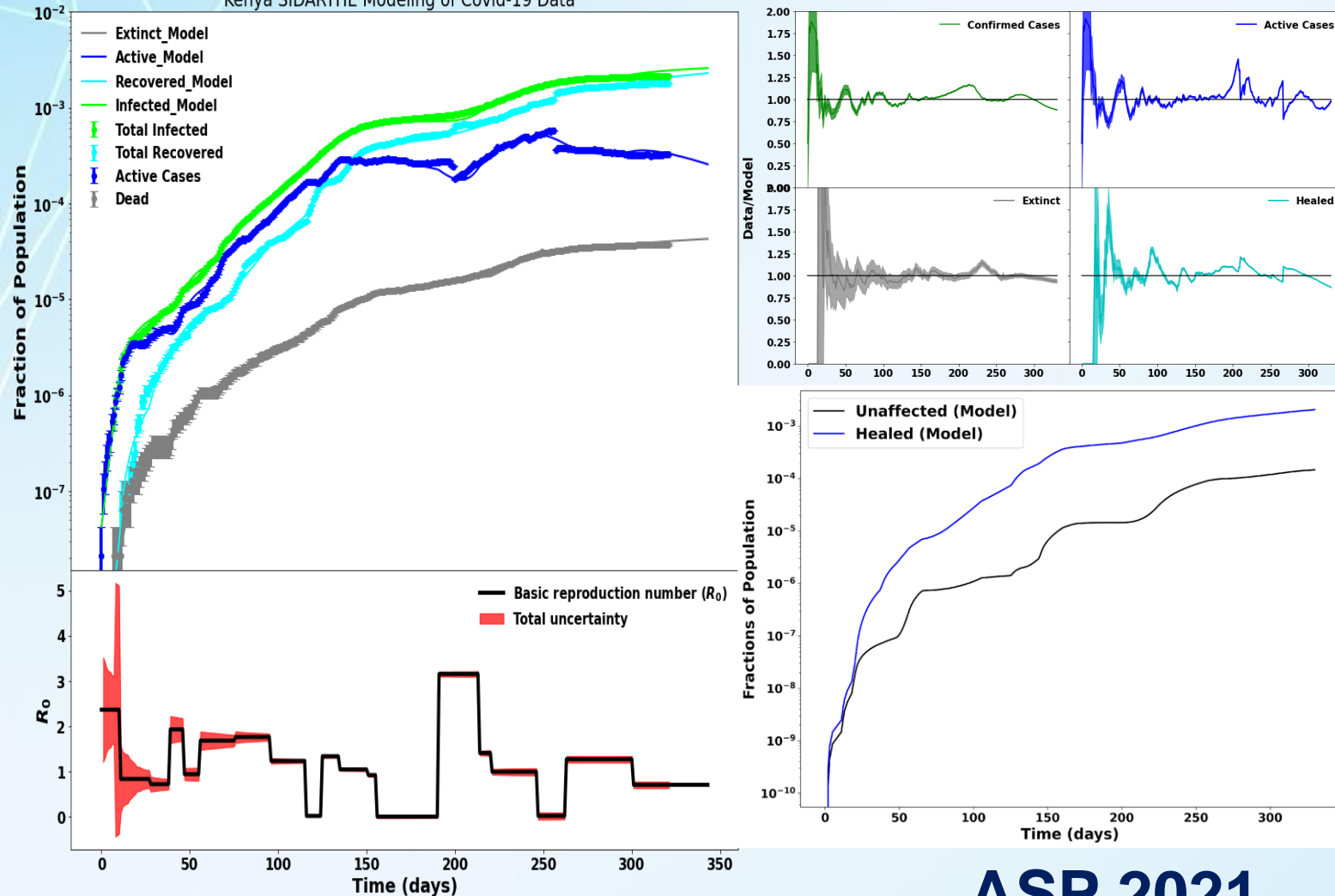
Level	Period	Summary
Adjusted alert level 3	Effect from 29 December 2020 to 28 February 2021	Gatherings and recreational activities severely restricted. Curfew imposed and restrictions on alcohol sales and consumption. Restricted cross border travels were allowed.
Alert level 1	Effect from 21 September to 28 December 2020	International travel allowed with restrictions, domestic travel open, all economic sectors open. For this and all lockdown levels, restrictions or closure on sporting events, religious gatherings, public entertainment, restaurants and similar.
Alert level 2	Effect from 18 August to 20 September 2020	Domestic air and road travel restored. Further economic sectors opened
Alert level 3	Effect from 1 June to 17 August 2020	Economy more open than Level 4, but still restrictions, example, no restaurants, restricted inter-provincial travel.
Alert level 4	Effect from 1 to 31 May 2020	Some non-essential services operate, with restrictions, eg: agriculture, mining, communications, business travel. Local travel within curfew, restricted provincial travel.
Alert level 5	Effect from midnight 26 March to 30 April 2020	Only essential services, transport and movement restrictions

$R_0 \sim 2.25$  at start , level 5 was effective and the infection rates decreased as  $R_0 \sim 1$

# Analysis of Covid 19 data from Kenya

- Gradual decline in the  $R_0$  implied that the control measures introduced were effective in earlier days.
- However  $R_0$  periodically rose and fell due to a number of reasons such as the increased testing capacity and relapses in adherence to the control measures.

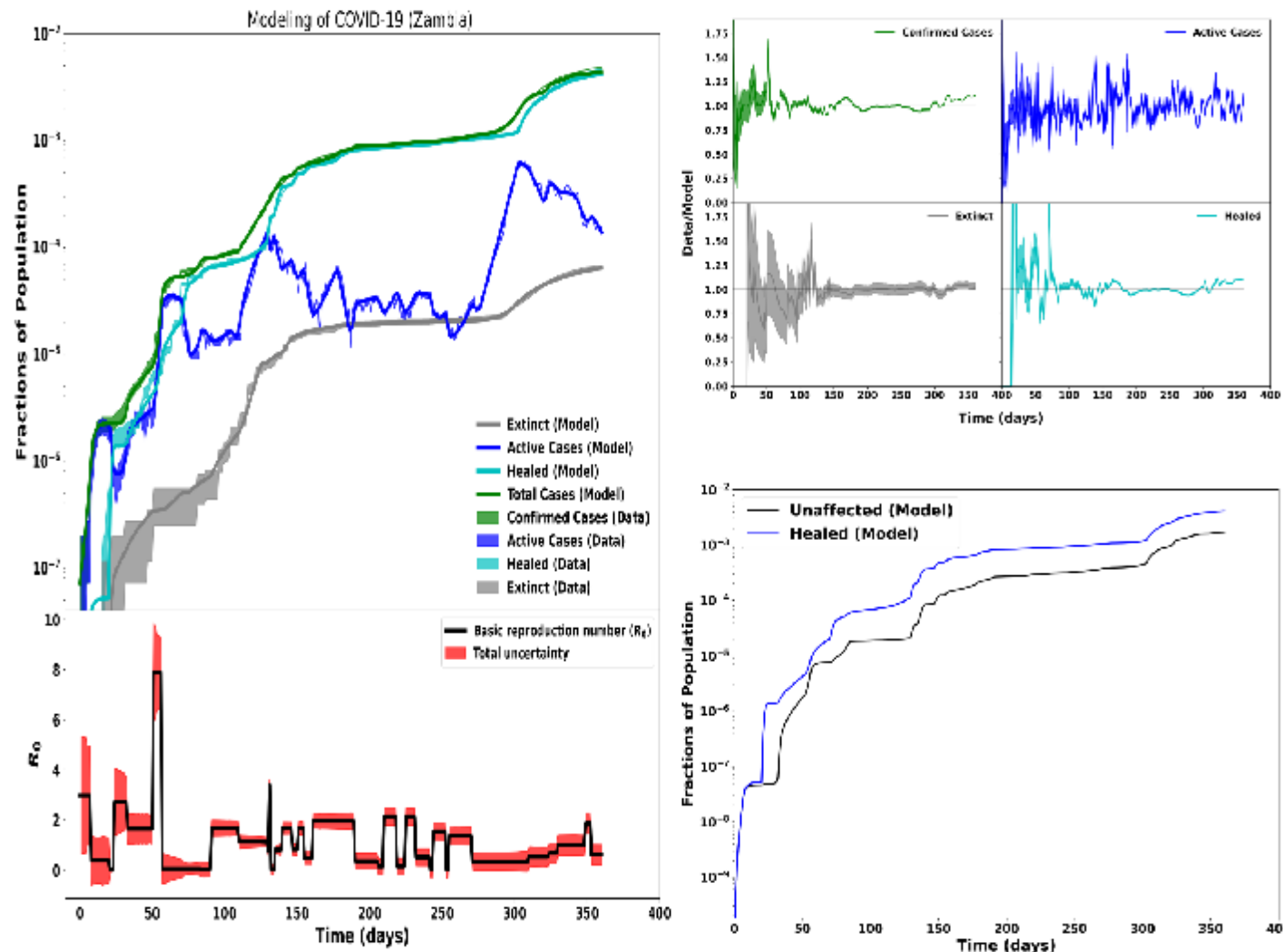
Kenya SIDARTHE Modeling of Covid-19 Data



- 1<sup>st</sup> Case: 13<sup>th</sup> March
- Travelled from US via London
- Attended party
- 2<sup>nd</sup> case friend of 1<sup>st</sup> case
- Testing capacity increased ~April
- Cessation of Movement in 2 major cities: Nairobi & Mombasa
- Flights suspended
- Partial country lockdown
- Night Curfews
- Social distancing in restaurants
- 50 people maximum
- Curve slowed down(1-4)
- Flattening of a curve(Jul)
- Fatality rate ~1.57%

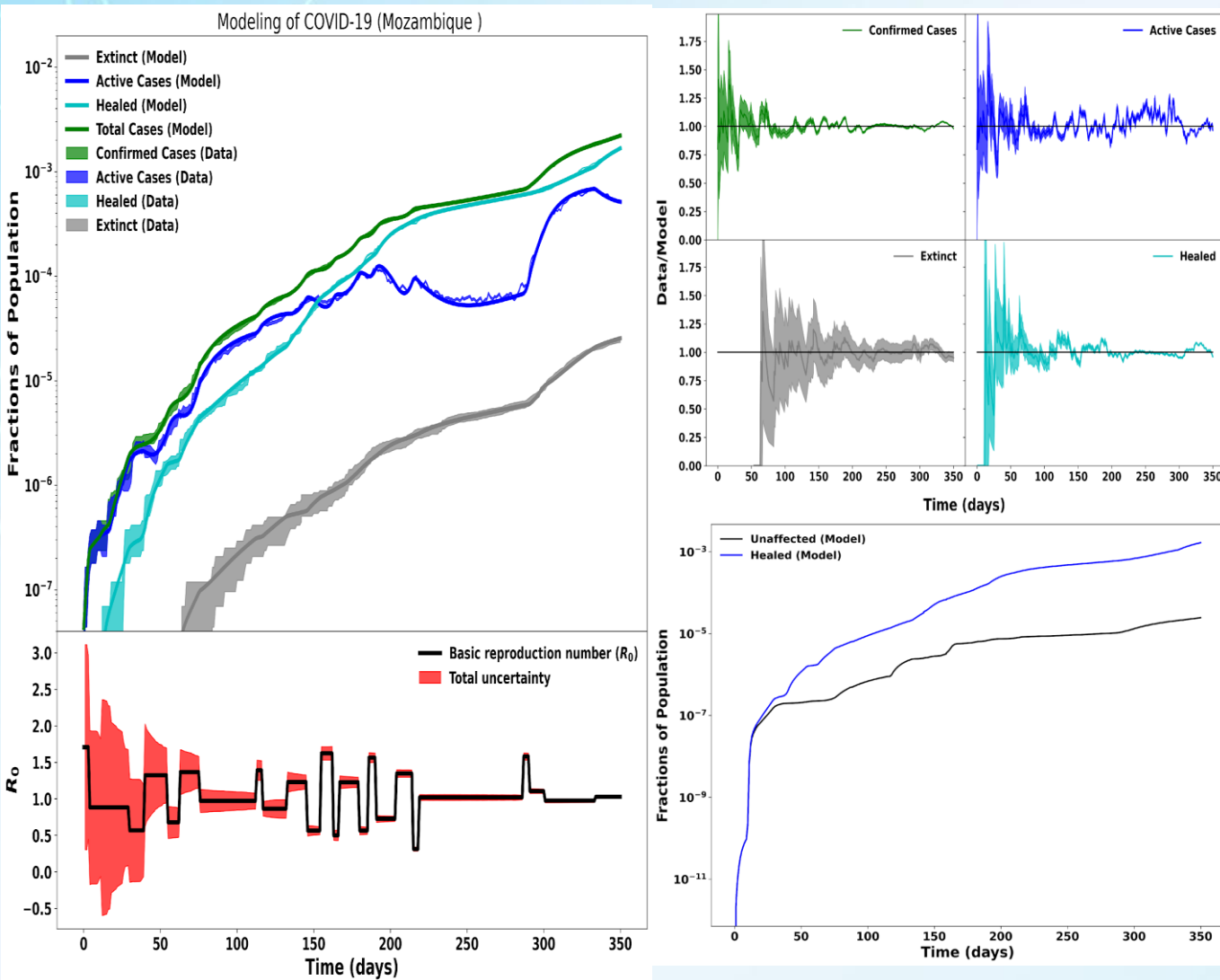


# Analysis of Covid 19 data from Zambia



- **1<sup>st</sup>Case: 18<sup>th</sup>March**
- **Presidential Directive -26/03: Measures to control spread of COVID-19**
- **Peak around day 60 truck driver crossing over from Tanzania in the northern region**
- $R_0$  value around day 90 reflected a gradual relaxation of physical distancing measures in May and June 2020.
- $R_0$  remained below 2 for all the dates, apart from around day 130 due to an increase in testing.
- **Cyclic rises and falls in the  $R_0$  as a function of time**
- **No obvious correlations between the  $R_0$  and the control measures can be made,**
- **However, difficulties to adhere to government directives or porous regional borders—might account for the observed patterns.**

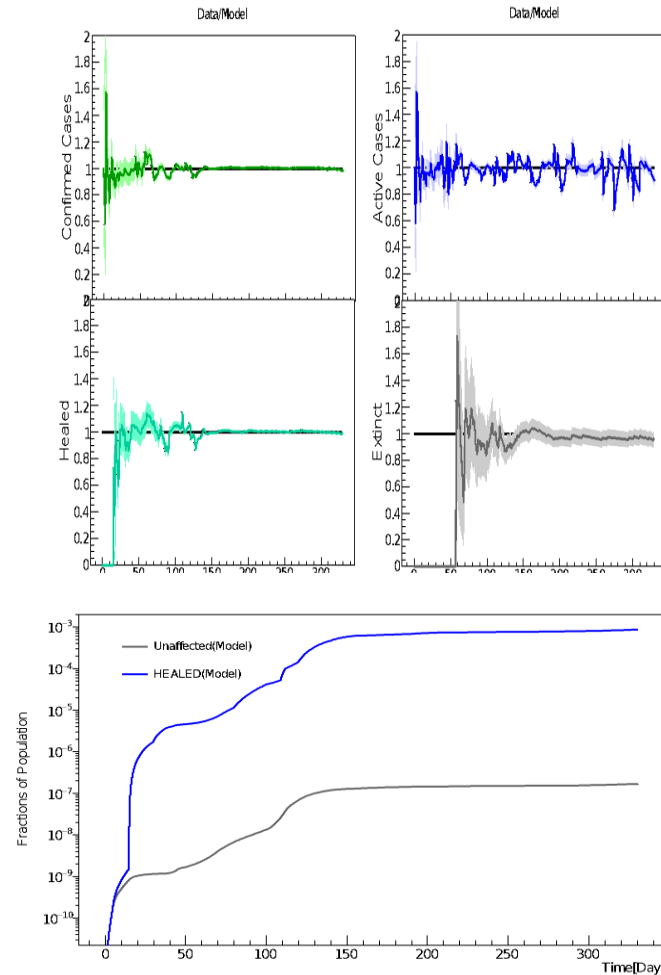
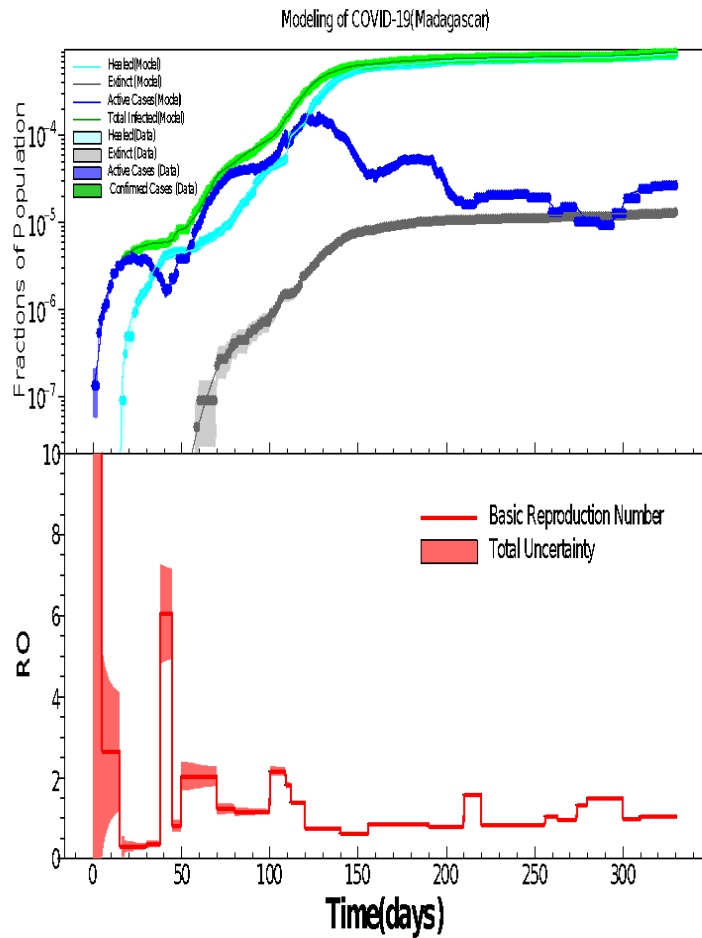
# Analysis of Covid 19 data from Mozambique



- **1<sup>st</sup>Case : 22<sup>nd</sup>March**
- **Case had travelled from the UK**
- **Schools closed immediately**
- **Social gatherings restricted**
- **State of emergency declared 01/04-29/07**
- **Flights suspended on 12/05**
- **No strict country lockdown**
- **Country slowly reopening with social distancing measures in place**
- **End of year 2020 (around day 280 ):many violations of the containment measures during the festive**

# Analysis of Covid 19 data from Madagascar

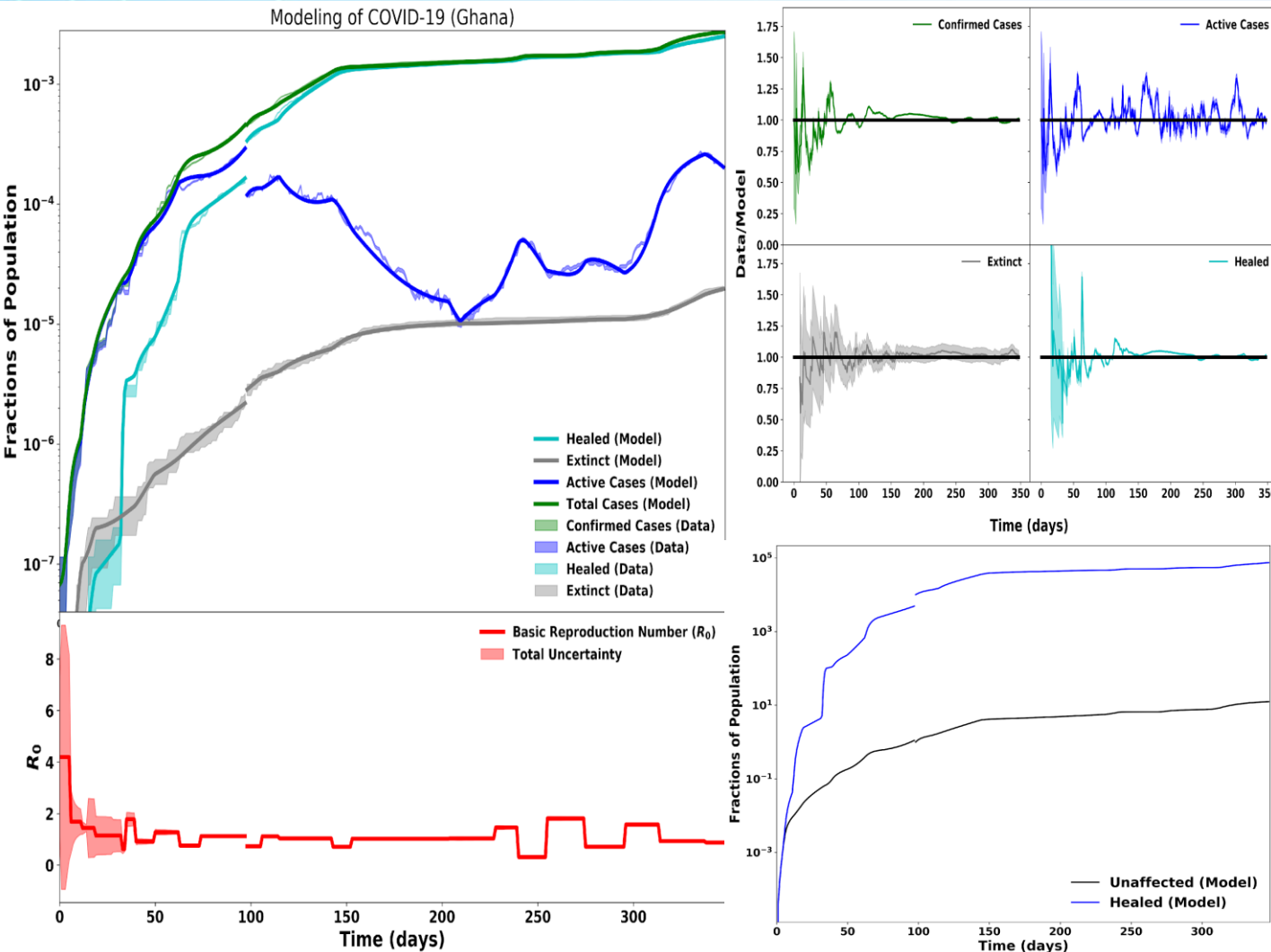
## ➤ 1<sup>st</sup> 3 Cases: 20 March



- 57 cases at end of March
- High recovery end of April & dip in Active cases
- Increased testing capacity
- Infection/Active peaked from May-July. Maximum peak in July
- Infection slowed down from August
- Flattening of the curve towards end of August
- Country is currently closed off to international flights
- 25-45 cases per week

# Analysis of Covid 19 data from Ghana

➤  $R_0$  and implemented measures are correlated

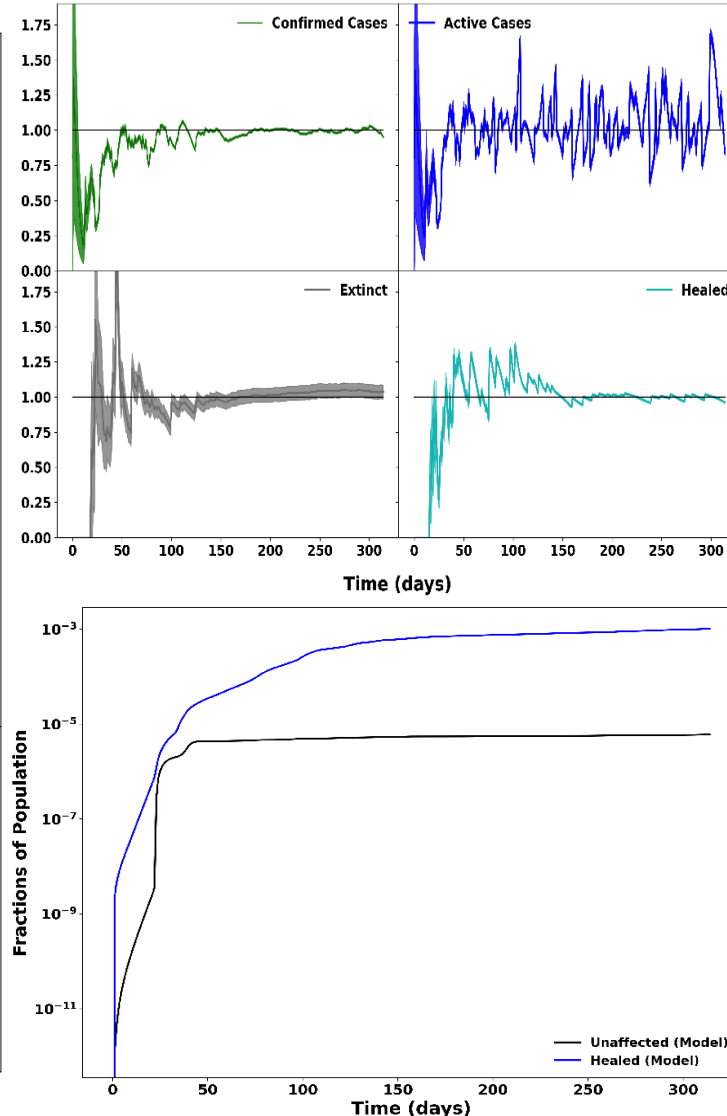
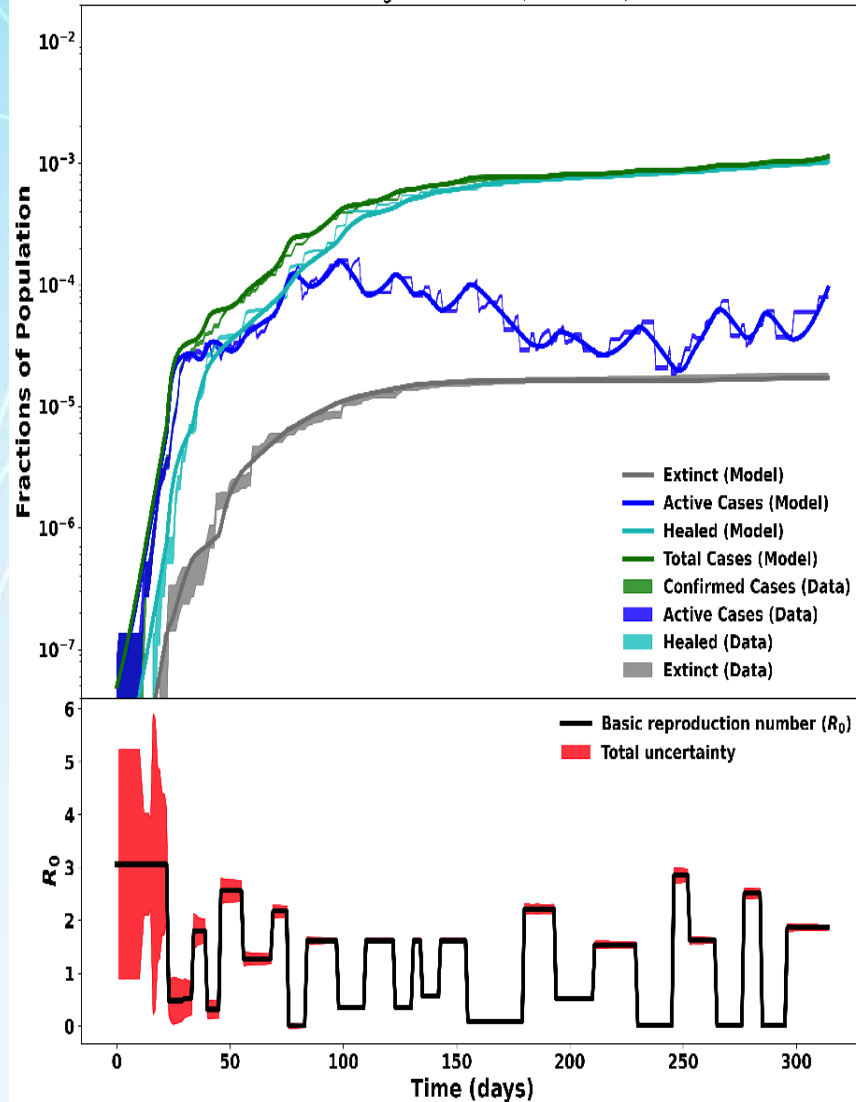


- **1<sup>st</sup> Case: 12 March**
- **Day 6: Ban on all public gatherings**
- **Day 12: Borders & beaches closed**
- **Day 18: Partial lockdown**
- **Day 38: Lockdown lifted**
- **Day 47: Mandatory use of masks**
- **Day 60: Reopening of Hotels, bars and restaurants**
- **Day 98: Discharge policy is reviewed**
- **Instead of following the two negative tests protocols, the infected who did not show any symptoms or whose symptoms disappeared during treatment**
- **Day 138: Public transport is allowed**
- **Day 173: Airports are opened.**
- **Day 250: Borders and schools are opened.**



# Analysis of Covid 19 data from Cameroon

Modeling of COVID-19 (Cameroon)

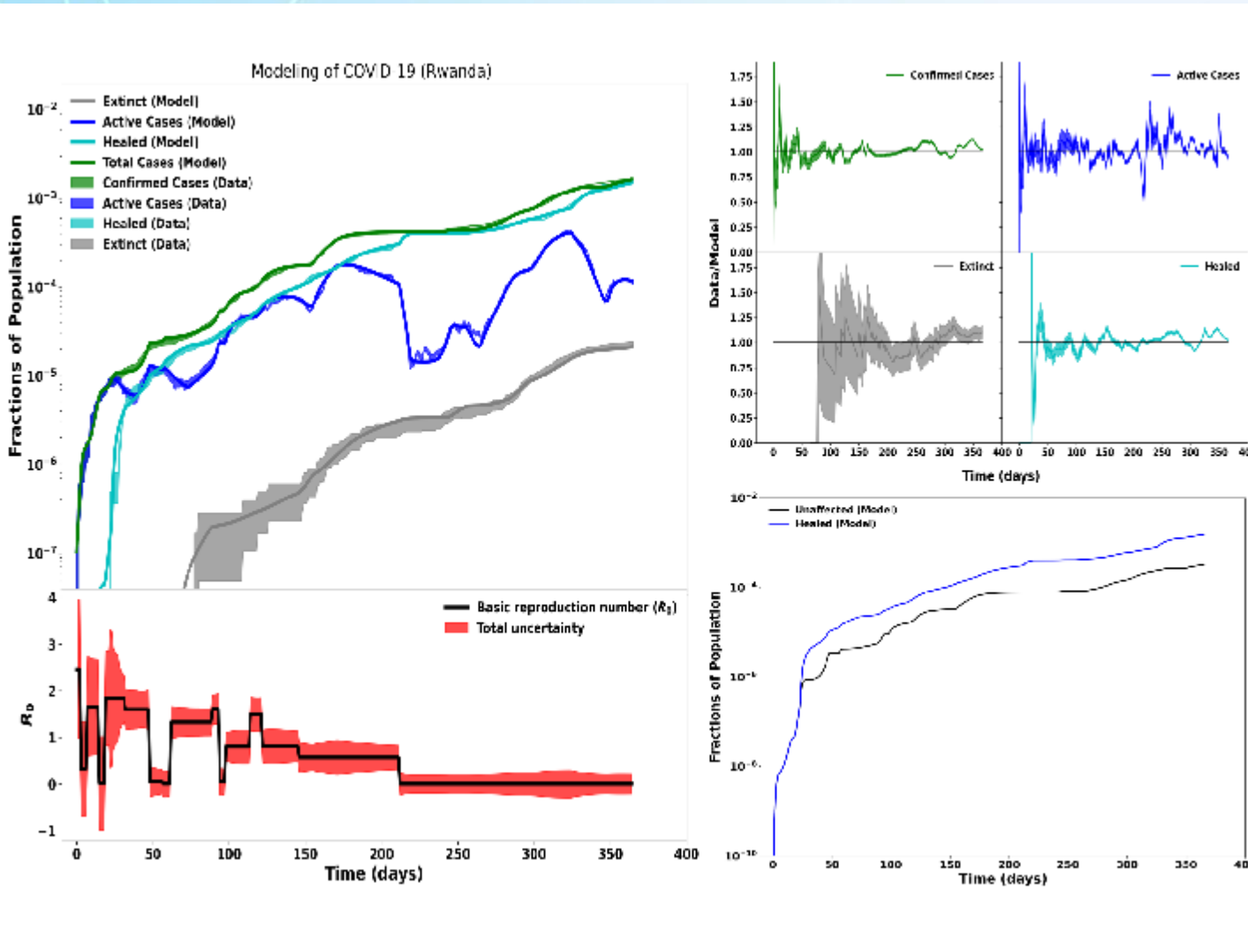


- **1<sup>st</sup>Case : 06 March**
- **Closed borders on 18/03**
- **Screening & quarantine measures for returnees**
- **Schools closed**
- **6pm curfew**
- **Social gatherings restricted**

# Analysis of Covid 19 data from Rwanda

➤ **1st Case : 14th March , foreigner**

- **Schools closed immediately**
- **Testing of symptomatic cases & contact tracing**
- **21/03 National lockdown**
- **Contact tracing one of the most effective measure used by the Rwandan Govt to test asymptomatic individuals who came into contact with the infected persons.**
- **Lowers the active cases after just the 1st month of pandemic outbreak**



# CONCLUSION

- We analyzed about one year of COVID-19 data of Cameroon, Ghana, Kenya, Madagascar, Mozambique, Rwanda, South Africa, Togo and Zambia.
- For each country, we estimated the time-dependent basic reproduction number,  $R_0$ . At the onset of the pandemic,  $R_0$  was above one in all the cases studied.
- Over time, the basic reproduction numbers followed the fluctuation patterns reflected in the data. The fluctuations were correlated with the control measures imposed and the emergence of new cases.
- Approximately eleven months since the first cases were detected in the country's studies, all the  $R_0$  were about or below one, suggesting that the pandemic had slowed in these countries.
- However, because the virus mutates and new waves are most likely to occur, we suggest to maintain the control measures until enough vaccines have been administered to reach herd immunity.
- Our studies also estimated the fractions of the population that were infected and not diagnosed but recovered without symptoms; in general, we find that these fractions are between 1-10% of the recovered cases. Paper submitted for publication: <https://arxiv.org/pdf/2104.09675.pdf>



**THANK  
YOU**

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