



MCMC I for Infectious Diseases

Lab & Activity

Resources

Presentation Activity



PollEv.com/amandableichrodt759

Toolbox Code



https://github.com/gchowell/paramEstimation_forecasting_ODEmodels

Toolbox Tutorial



<https://pubmed.ncbi.nlm.nih.gov/38378161/>

Emory's Apporto

<https://rsphemory.apporto.com/>





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https://rsphemory.apporto.com

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APP STORE

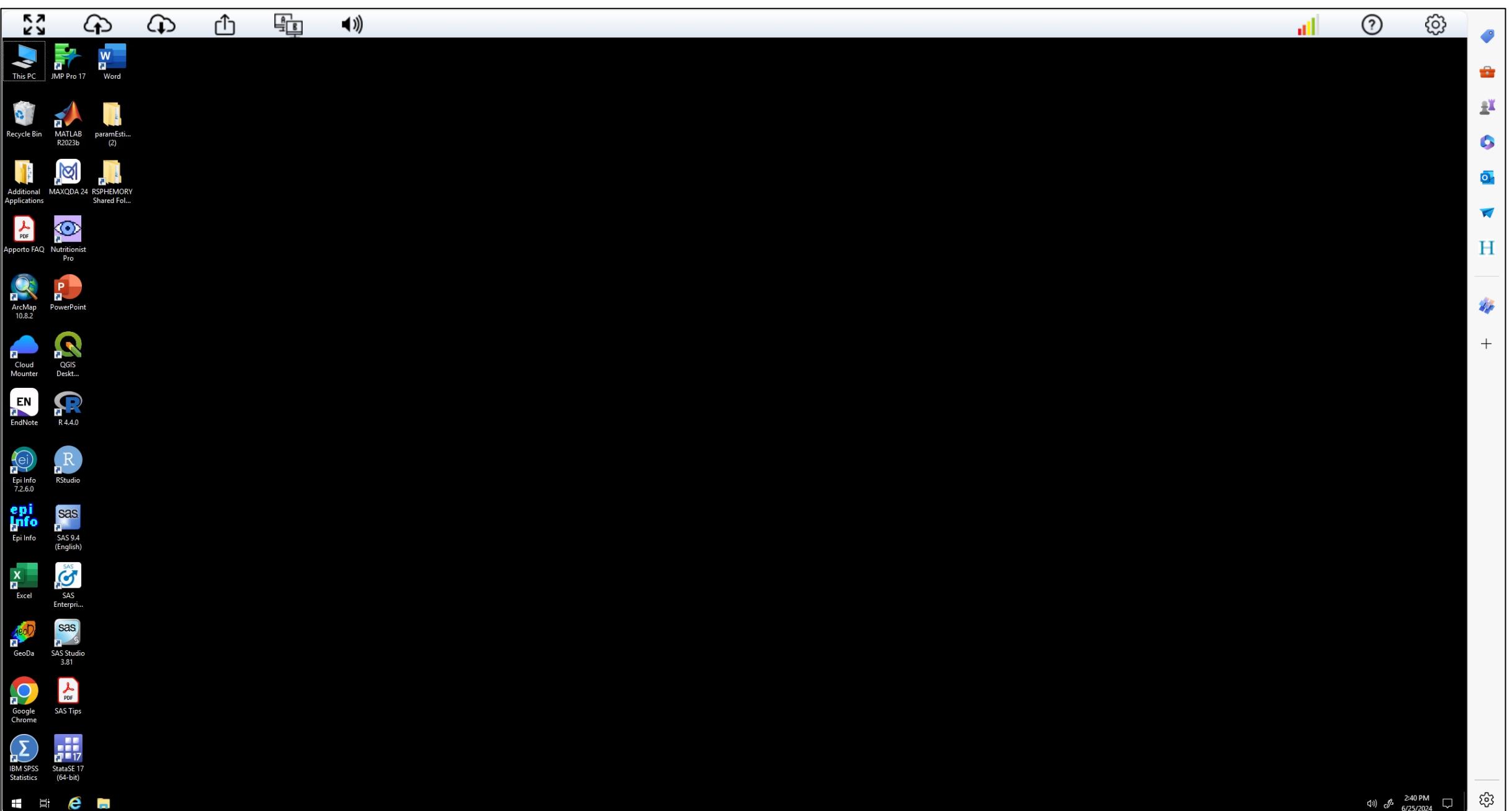
RSPH Desktop
Apporto
Streamed App

Launch

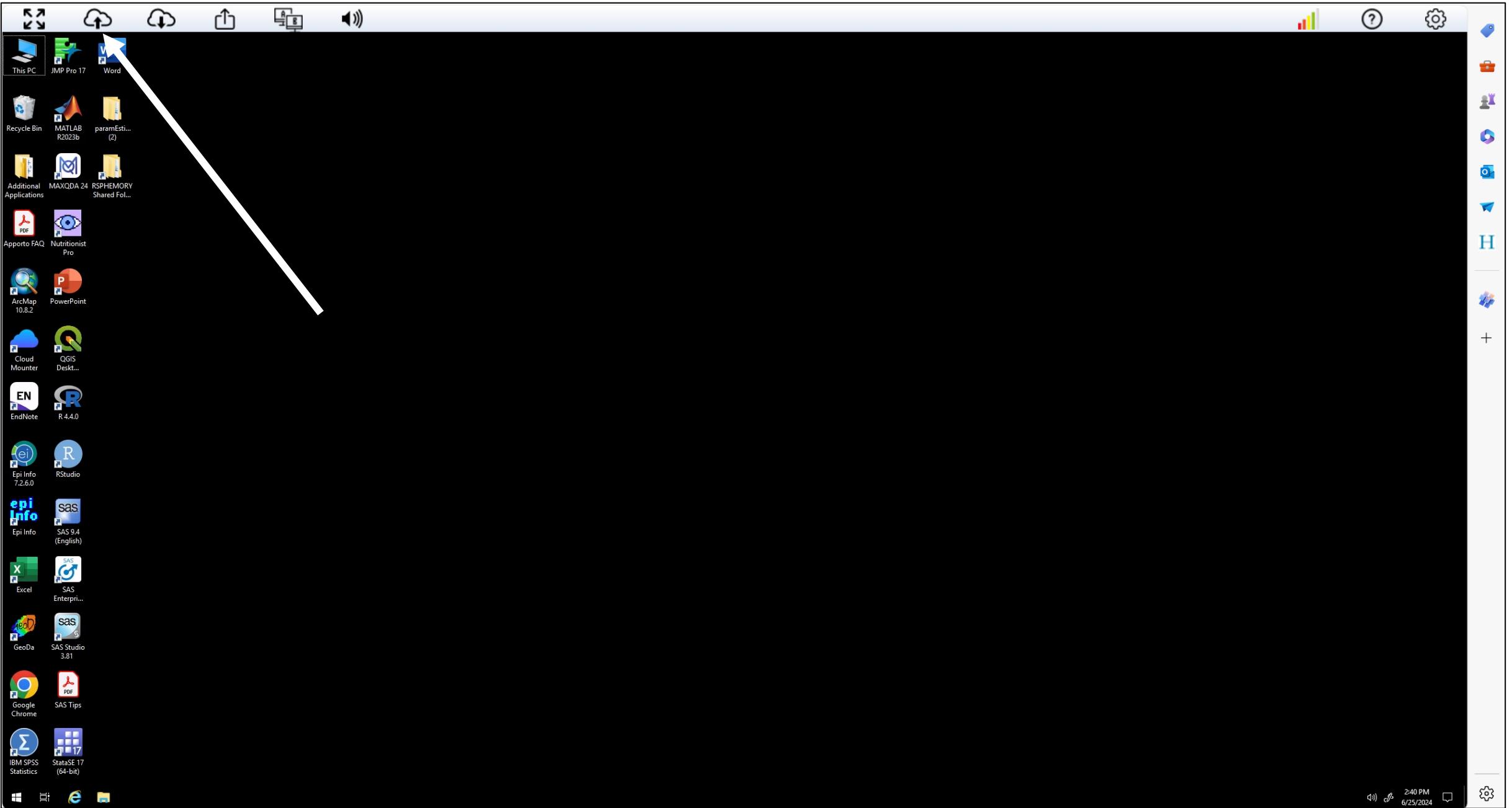
ANNOUNCEMENTS

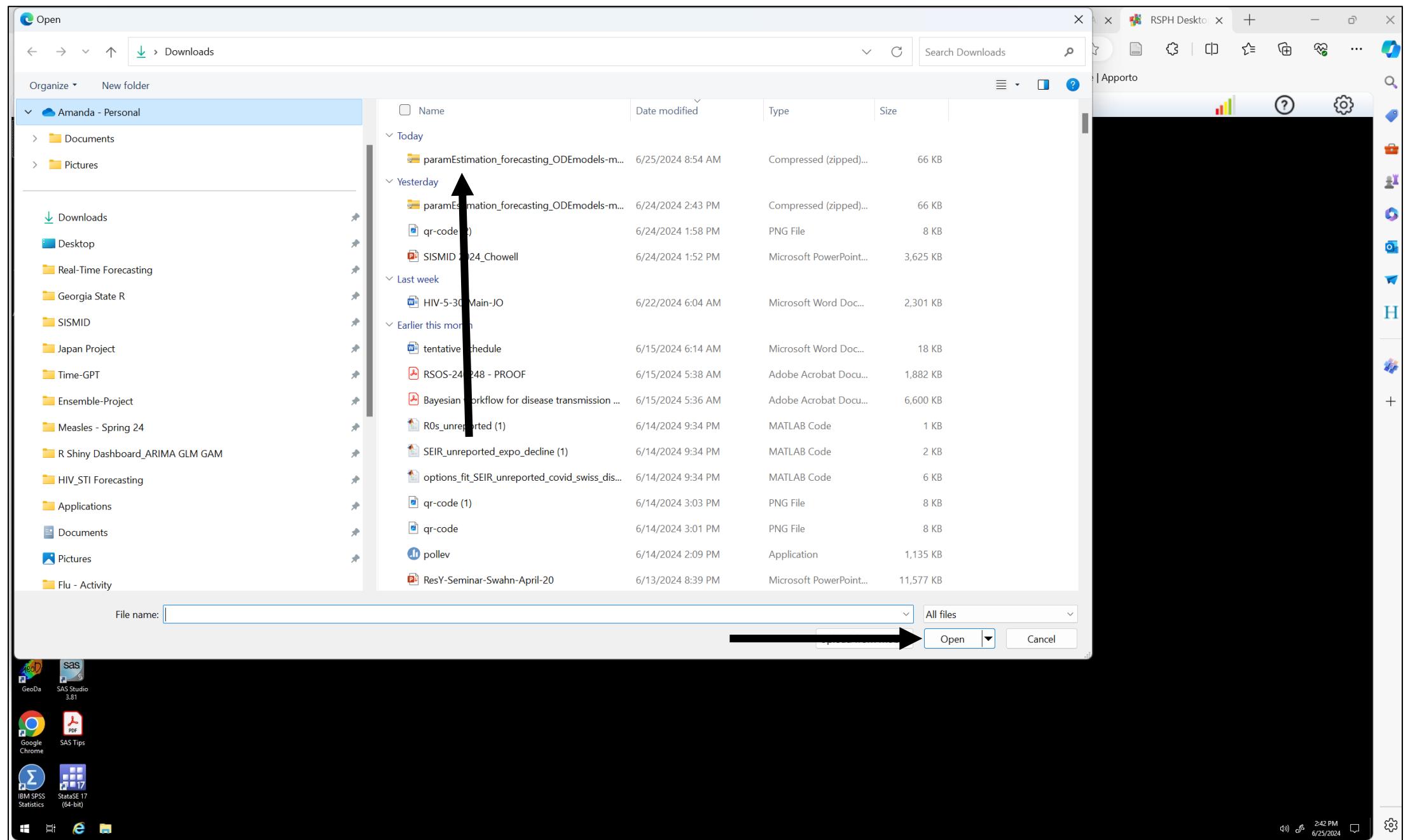
Please Read -- MAXQDA users
Please Read -- Connecting to Network Shares

A screenshot of the Apporto web interface. At the top, there's a navigation bar with various links like 'Lenovo Support', 'YouTube', 'Maps', etc. Below the navigation is the Apporto logo and the Emory Rollins School of Public Health branding. The main area is divided into sections: 'APP STORE' on the left containing an item for 'RSPH Desktop' which is described as an 'Apporto Streamed App' and has a 'Launch' button; and 'ANNOUNCEMENTS' on the right listing 'Please Read -- MAXQDA users' and 'Please Read -- Connecting to Network Shares'. On the far right, there's a vertical sidebar with icons for file operations like search, copy, paste, and help, along with a user profile icon.

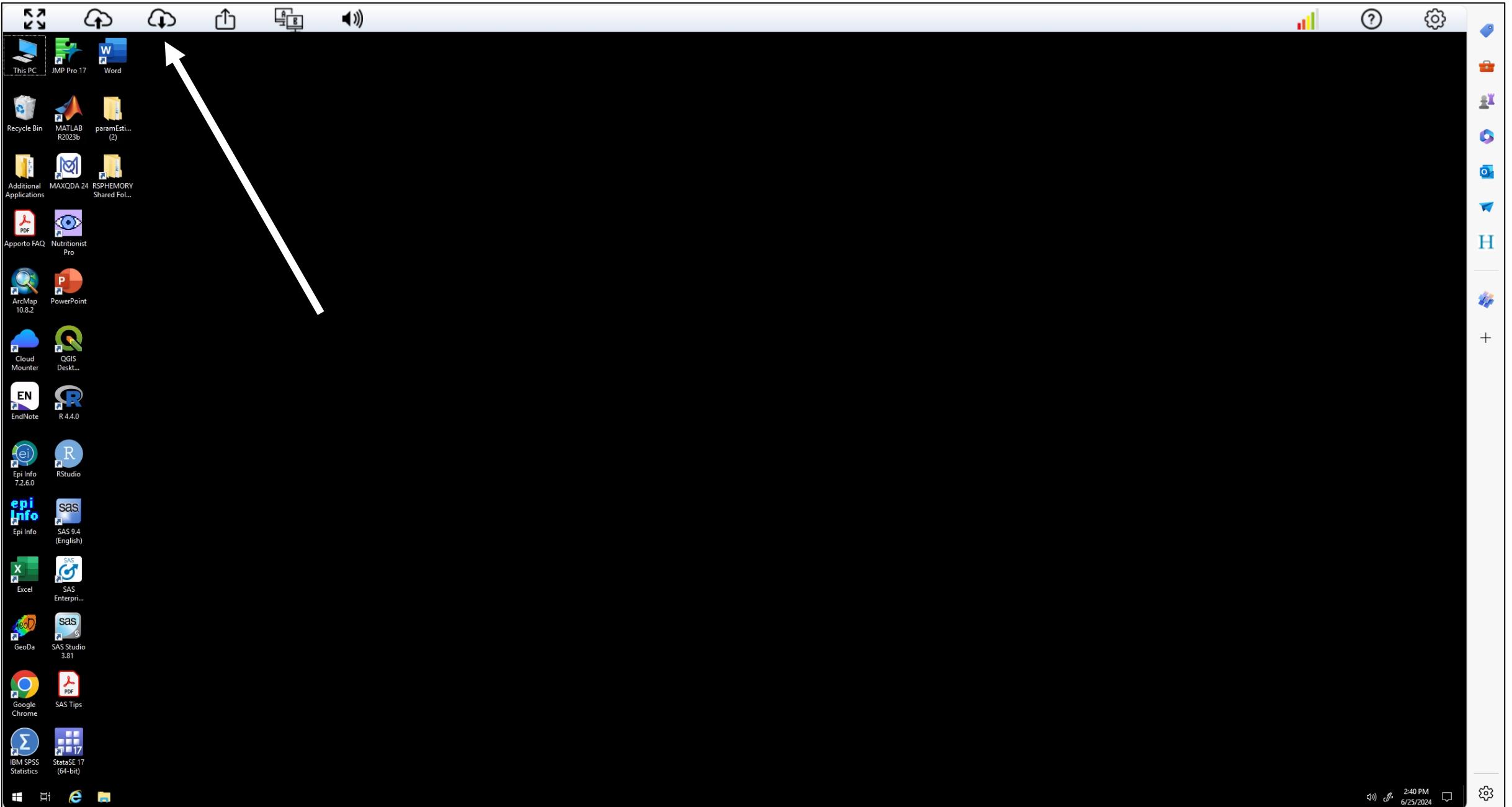


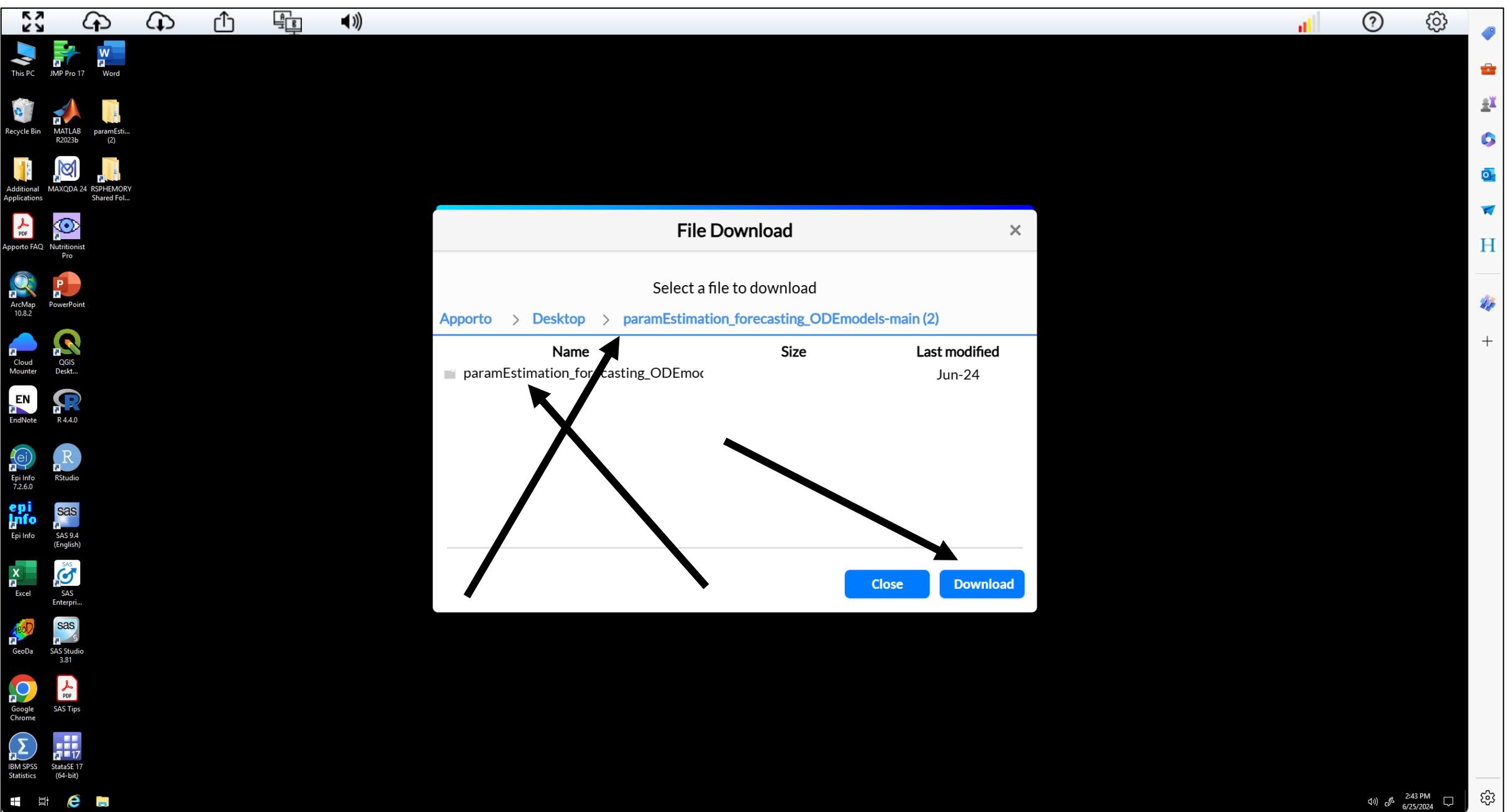
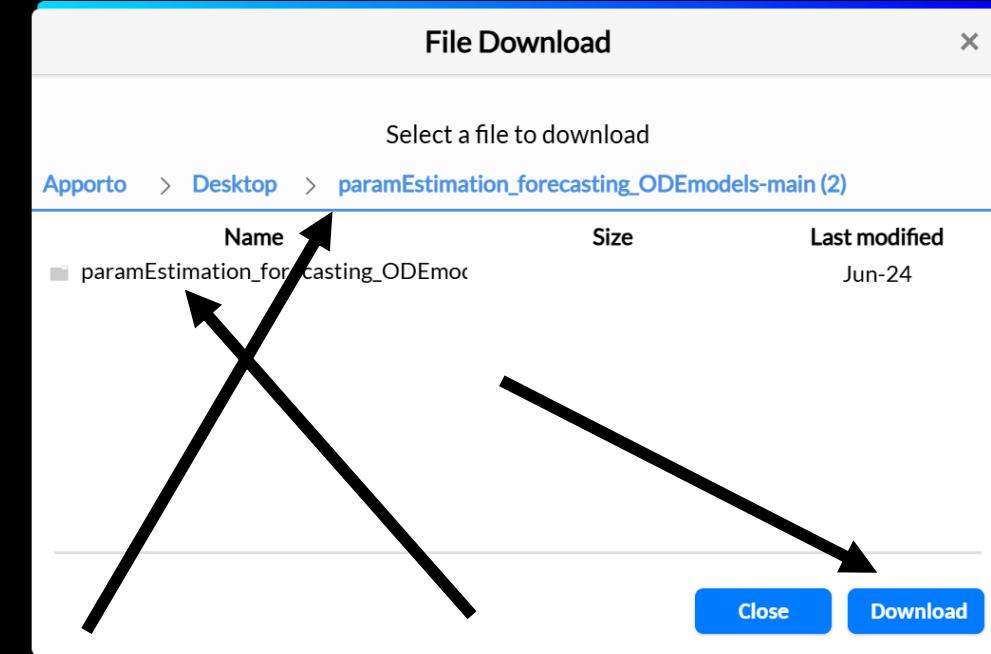
Uploading Materials



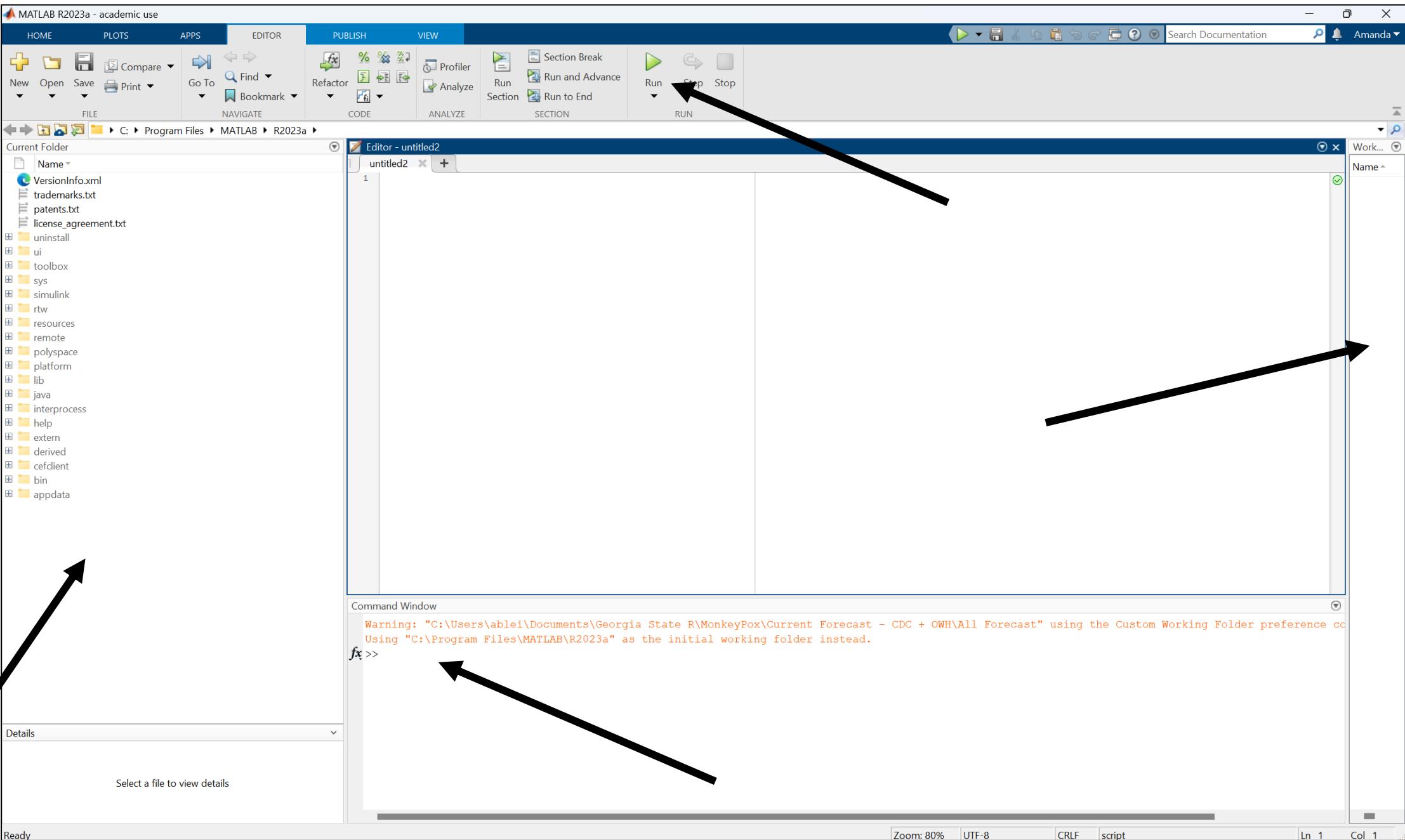


Downloading Data





Exploring MATLAB



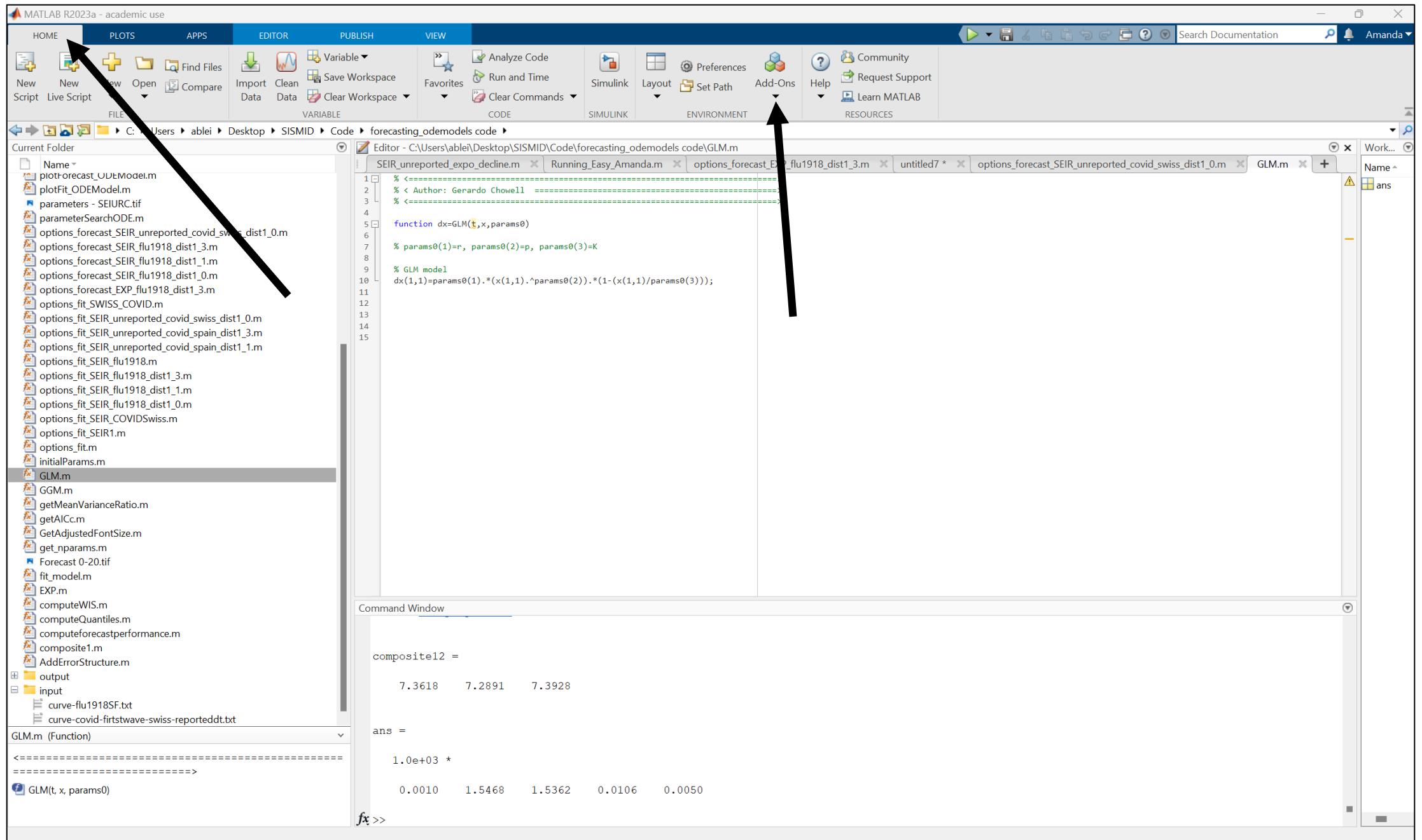
Needed Settings

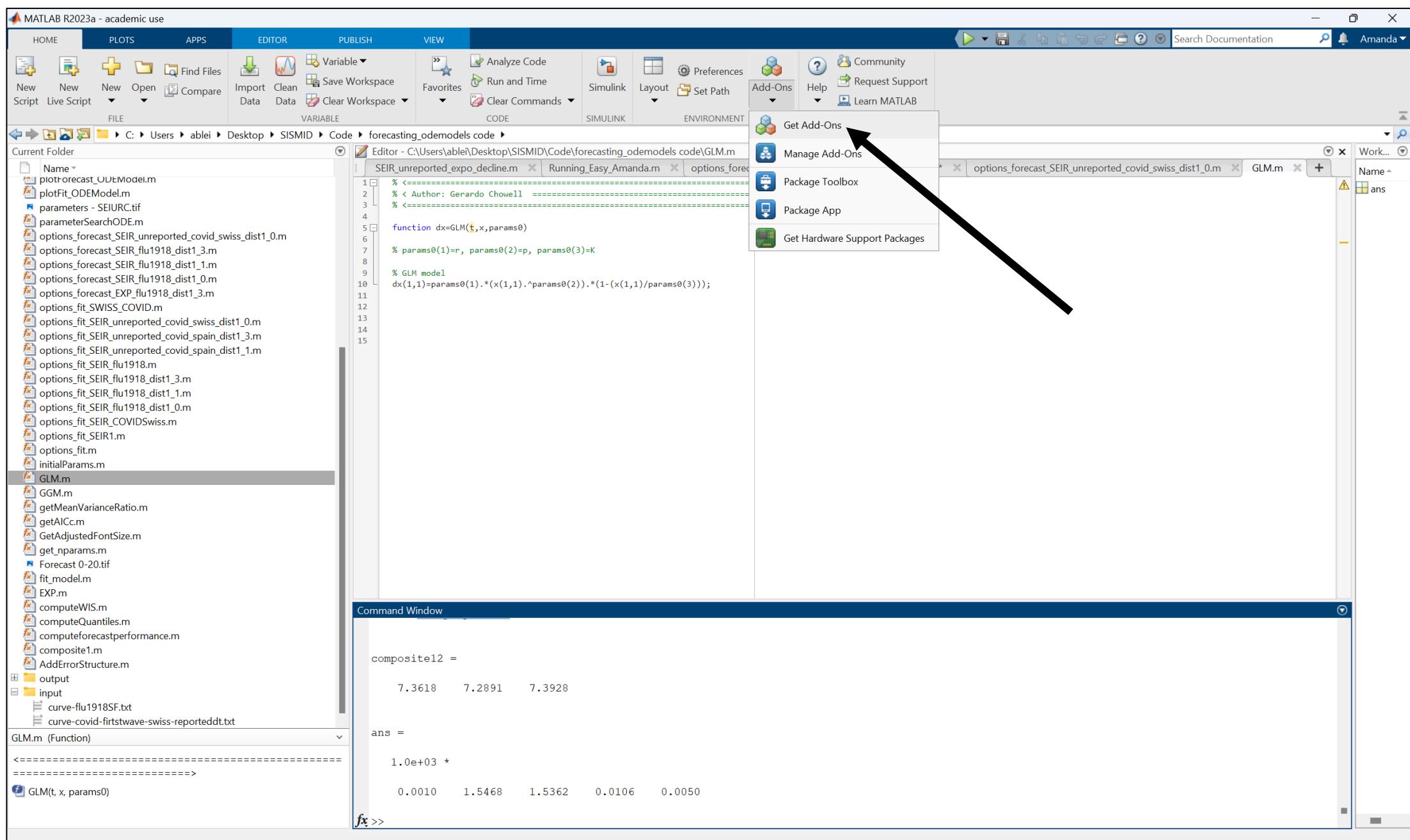
Needed MATLAB Functions
Curve Fitting Toolbox
Global Optimization Toolbox
Optimization Toolbox
Signal Processing Toolbox
Simulink
Statistics and Machine Learning Toolbox

Needed Folders
<i>input</i>
<i>output</i>

Adding Packages

**Not needed if using Emory's Virtual Lab*





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Workflows

Parallel Computing 189
Reporting and Database Access 205
Systems Engineering 27
Code Generation 334
Application Deployment 124
Verification, Validation, and Test 172
Cloud Capabilities 70
Teaching and Learning 3

Applications

Bioinformatics Toolbox Computer Vision Toolbox Deep Learning Toolbox GPU Coder

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Geoid Data for Aerospace Toolbox

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Audio Toolbox

Design and analyze speech, acoustic, and audio processing...

Circuit Level Analog / Mixed Signal Examples

Collection of Circuit Level Sims Models using Simulink and...

ECG simulation using MATLAB

This code generates all possible forms of ECG signals with the...

Google Earth Toolbox

Various plotting/drawing functions that can be saved as...

iPod Scramble Solution

Solves the Scramble game (like Boggle) by Zynga on my iPod.

Show All 124

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GPU Coder

GLM.m

Current Folder

Name

plotForecast_ODEModel.m
plotFit_ODEModel.m
parameters - SEIURC.tif
parameterSearchODE.m
options_forecast_SEIR_unreported
options_forecast_SEIR_flu1918_d
options_forecast_SEIR_flu1918_d
options_forecast_SEIR_flu1918_d
options_forecast_EXP_flu1918_d
options_fit_SWISS_COVID.m
options_fit_SEIR_unreported_cov
options_fit_SEIR_unreported_cov
options_fit_SEIR_unreported_cov
options_fit_SEIR_flu1918.m
options_fit_SEIR_flu1918_dist1_3
options_fit_SEIR_flu1918_dist1_1
options_fit_SEIR_flu1918_dist1_0
options_fit_SEIR_COVIDSwiss.m
options_fit_SEIR1.m
options_fit.m
initialParams.m
GLM.m
GGM.m
getMeanVarianceRatio.m
getAICc.m
GetAdjustedFontSize.m
get_nparams.m
Forecast 0-20.tif
fit_model.m
EXP.m
computeWIS.m
computeQuantiles.m
computeForecastPerformance.m
composite1.m
AddErrorStructure.m
output
input
curve-flu1918SF.txt
curve-covid-firstwave-swiss-re

GLM.m (Function)

GLM(t, x, params0)

fx >>

Zoom: 80% UTF-8 LF GLM Ln 15 Col 1

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Workflows

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- Systems Engineering 27
- Code Generation 334
- Application Deployment 124
- Verification, Validation, and Test 172
- Cloud Capabilities 70
- Teaching and Learning 3

MathWorks Toolboxes and Products

- Bioinformatics Toolbox
- Computer Vision Toolbox
- Deep Learning Toolbox
- GPU Coder

curve fitting

curve fitting

Curve Fitting Toolbox

Fit curves and surfaces to data using regression, interpolation, and smo...

EzyFit 2.44

A free **curve fitting** toolbox for Matlab

gaussian curve fit

gaussian curve fit

Multiple **curve fitting** with common parameters using NLINFIT

Evolutionary **curve fitting**

Suggestions

- curve fittingsmoothen
- fittingmethod
- fittings
- fittinganddesigningexperiments
- fittingby

GLM.m

GLM.m (Function)

GLM(t, x, params0)

fx >

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Curve Fitting Toolbox

by MathWorks

Fit curves and surfaces to data using regression, interpolation, and smoothing

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Fit Plot

Overview Functions Apps Examples

Curve Fitting Toolbox provides an app and functions for fitting curves and surfaces to data. The toolbox lets you perform exploratory data analysis, preprocess and post-process data, compare candidate models, and remove outliers. You can conduct regression analysis using the library of linear and nonlinear models provided or specify your own custom equations. The library provides optimized solver parameters and starting conditions to improve the quality of your fits. The toolbox also supports nonparametric modeling techniques, such as splines, interpolation, and smoothing. After creating a fit, you can apply a variety of post-processing methods for plotting, interpolation, and extrapolation; estimating confidence intervals; and calculating integrals and derivatives.

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Categories

AI, Data Science, and Statistics > Curve Fitting Toolbox > Get Started with Curve Fitting Toolbox

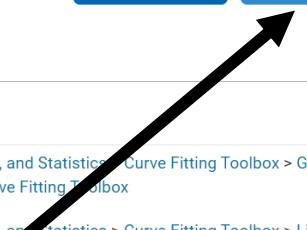
AI, Data Science, and Statistics > Curve Fitting Toolbox > Linear and Nonlinear Regression

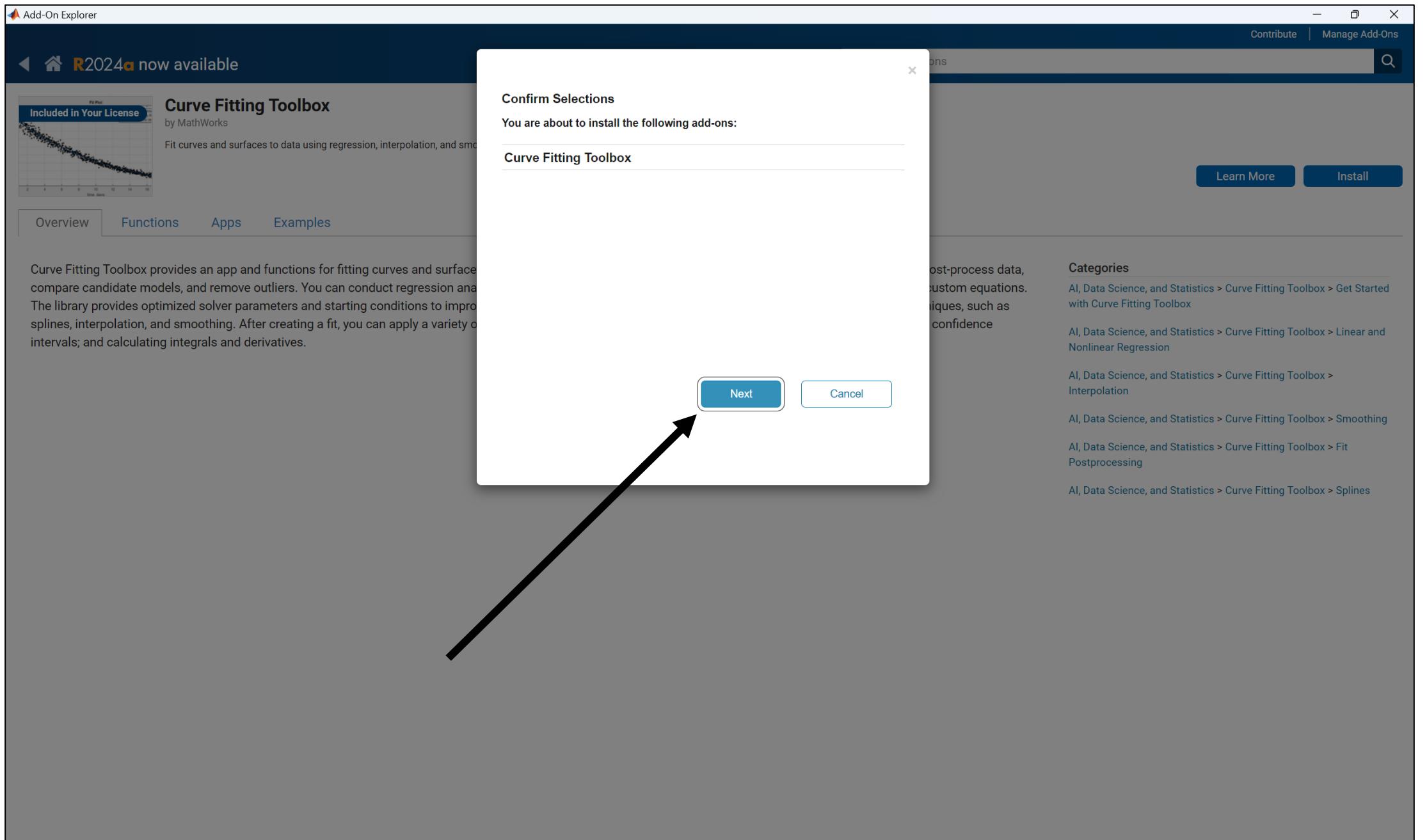
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AI, Data Science, and Statistics > Curve Fitting Toolbox > Smoothing

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Fit curves and surfaces to data using regression, interpolation, and smoo

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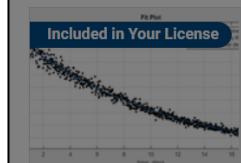
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Overview

Functions

Apps

Examples

Curve Fitting Toolbox provides an app and functions for fitting curves and surfaces to data. You can compare candidate models, and remove outliers. You can conduct regression analysis, and post-process data, such as custom equations. The library provides optimized solver parameters and starting conditions to improve performance. You can also use splines, interpolation, and smoothing. After creating a fit, you can apply a variety of techniques, such as confidence intervals; and calculating integrals and derivatives.

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post-process data, such as custom equations. You can also use techniques, such as confidence intervals; and calculating integrals and derivatives.

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The *QuantDiffForecast* Toolbox

Multiple parameter estimation options

- Estimation: NLSQ & MLE
- Error distributions: Normal, Poisson, and four variations of negative binomial
- Bootstrapping and number of initial guesses

Flexible model specifications

- User specified system of ODEs
 - Initial parameter and state variable conditions
 - Estimation of composite function(s) from the specified model parameters

Evaluation of bootstrapping error, model fit and forecasting performance

- Mean Squared Error (MSE)
- Mean Absolute Error (MAE)
- Weighted Interval Scores (WIS)
- 95% PI Coverage
- Monte Carlo Squared Error (MCSE)

Downloading the Toolbox

https://github.com/gchowell/paramEstimation_forecasting_ODEmodels

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QuantDiffForecast

Parameter estimation, forecasting, and uncertainty quantification for ODE models

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Video tutorial: <https://www.youtube.com/watch?v=eyyX63H12sY&t=41s>

It carries out the following tasks:

- fitting ODE models to time series data,
- estimation of model parameters with quantified uncertainty, Monte Carlo standard errors (MCSES),
- plotting the best fit of the ODE model, calibration performance metrics, and empirical distribution of the parameters
- plotting forecasts from the best-fit model and performance metrics of the forecasts,
- conducts rolling window analyses of parameter estimates for specific periods and window sizes

About

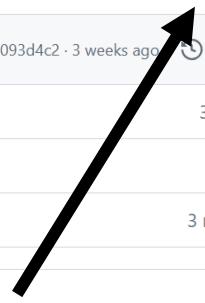
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Releases No releases published

Packages No packages published

Languages MATLAB 100.0%



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HTTPS SSH GitHub CLI

https://github.com/gchowell/paramEstimation_for

Clone using the web URL.

Open with GitHub Desktop

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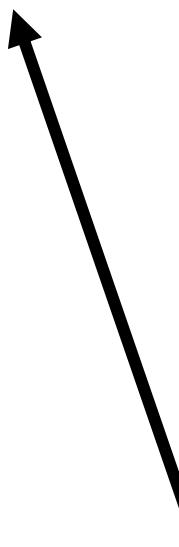
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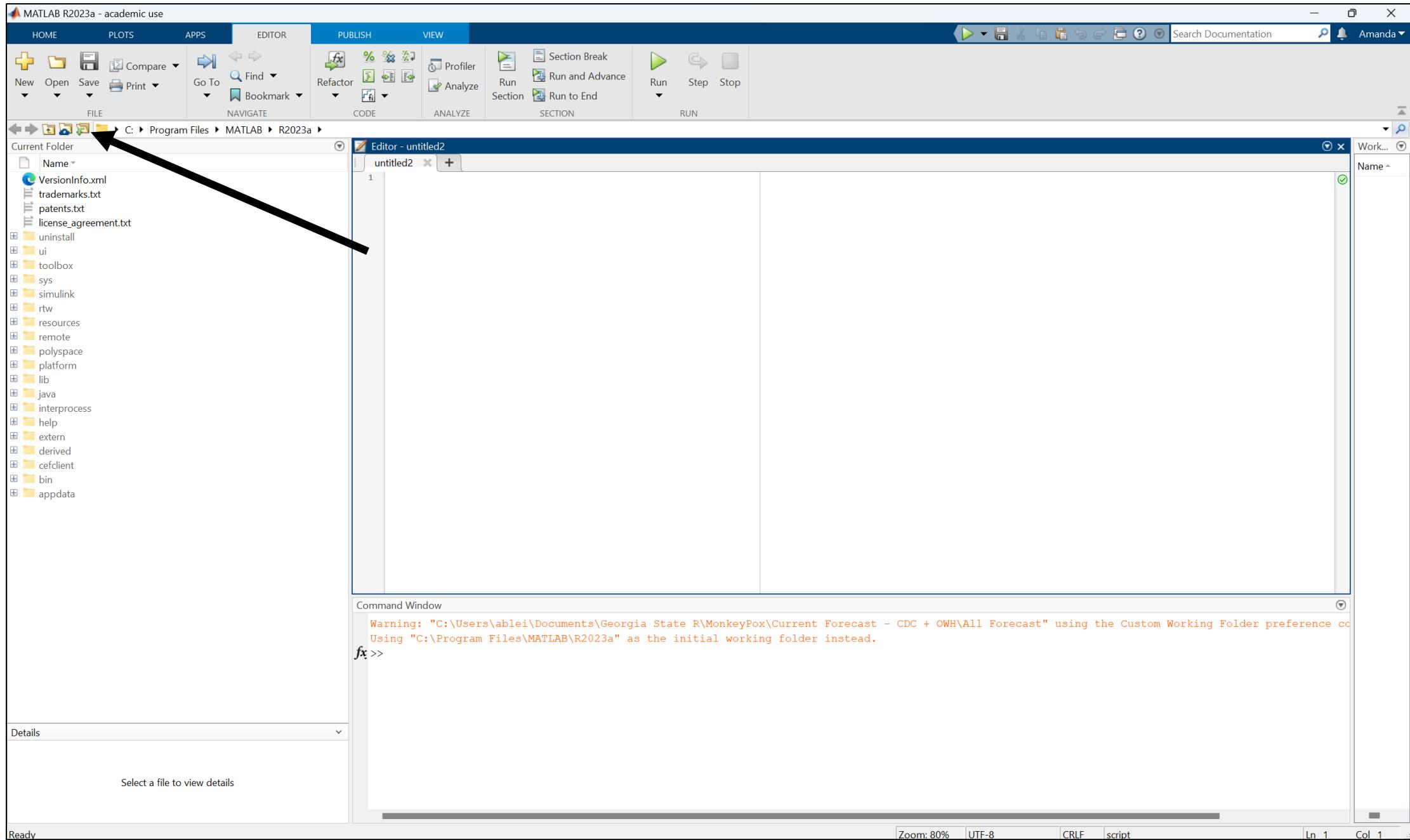


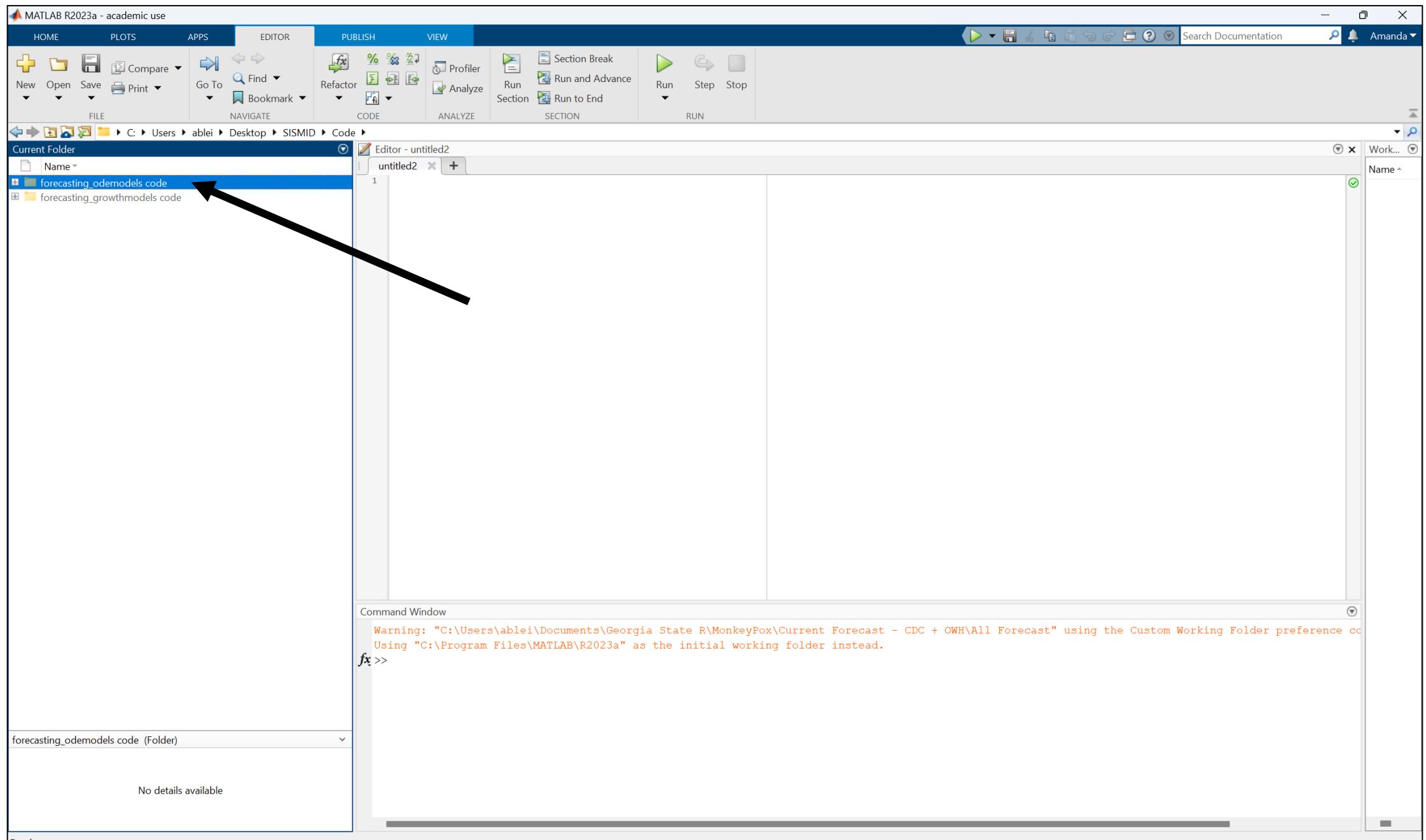
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LICENSE	File	3 KB	No	7 KB	61%	6/6/2024 9:52 AM
README.md	MD File	1 KB	No	2 KB	50%	6/6/2024 9:52 AM

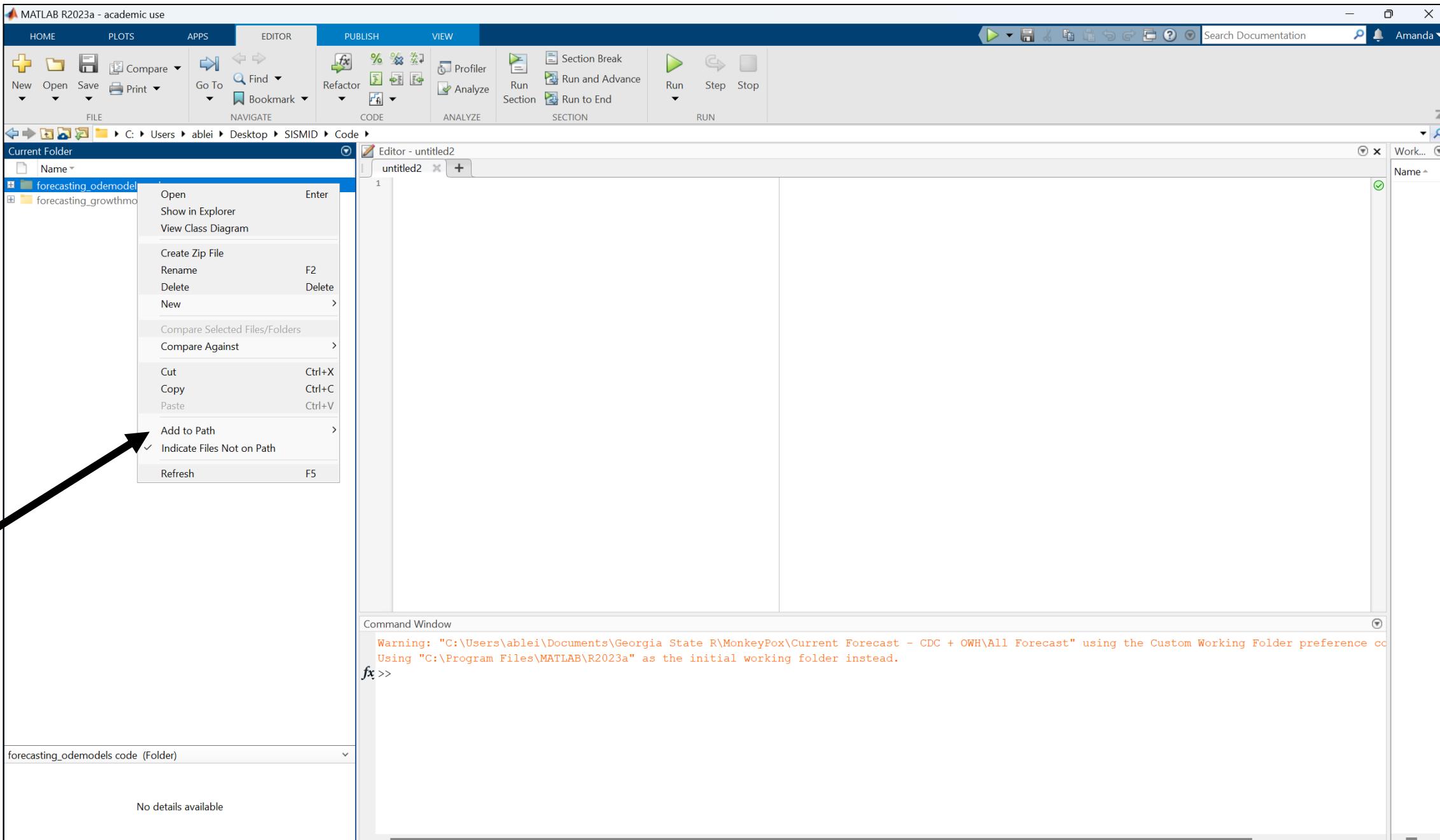


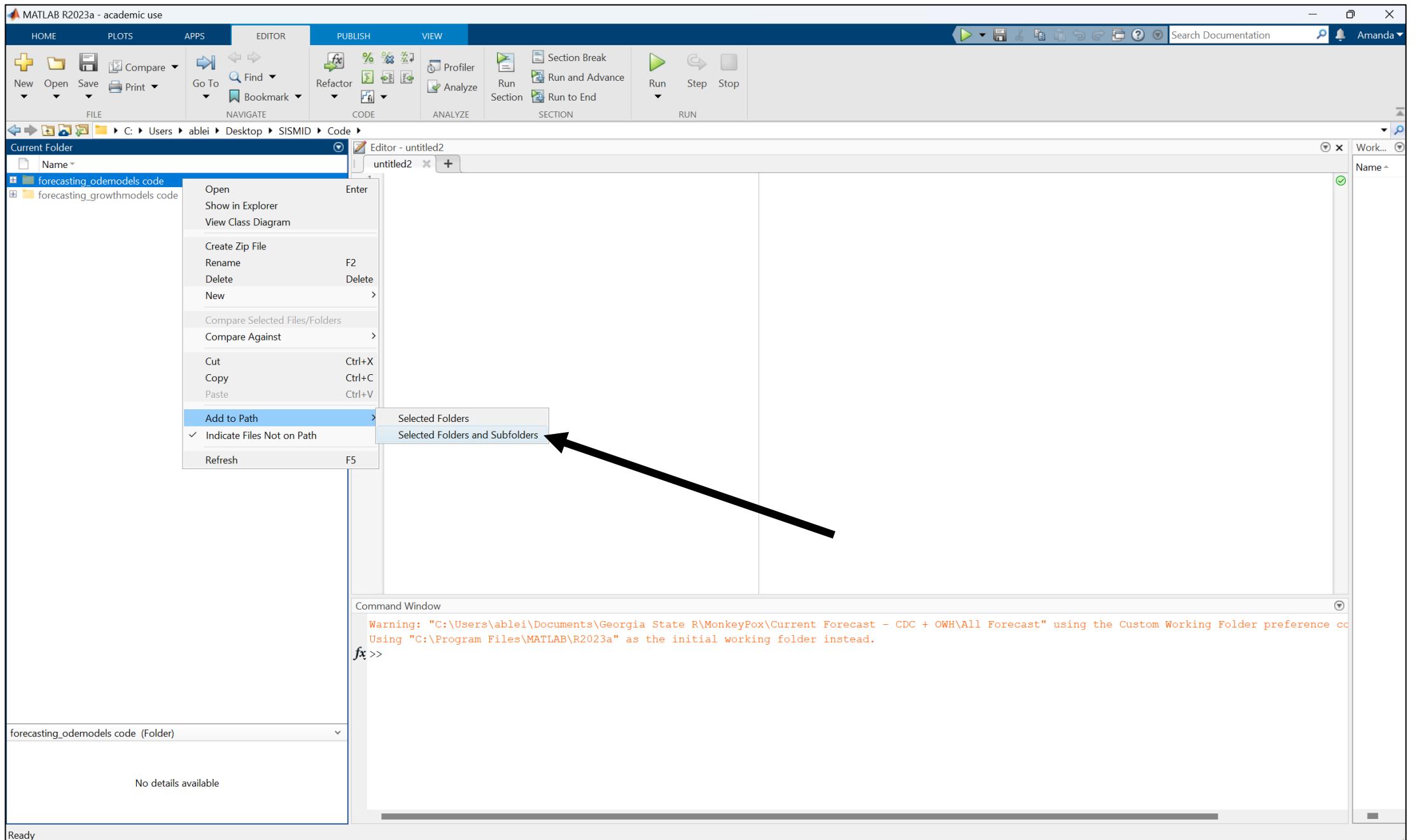
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computeQuantiles	MATLAB Code	1 KB	No	2 KB	60%	6/6/2024 9:52 AM
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getAICc	MATLAB Code	1 KB	No	2 KB	68%	6/6/2024 9:52 AM
getMeanVarianceRatio	MATLAB Code	1 KB	No	1 KB	74%	6/6/2024 9:52 AM
GGM	MATLAB Code	1 KB	No	1 KB	64%	6/6/2024 9:52 AM
GLM	MATLAB Code	1 KB	No	1 KB	66%	6/6/2024 9:52 AM
initialParams	MATLAB Code	1 KB	No	1 KB	71%	6/6/2024 9:52 AM
options_fit	MATLAB Code	2 KB	No	6 KB	71%	6/6/2024 9:52 AM
options_fit_SEIR_flu1918	MATLAB Code	2 KB	No	6 KB	70%	6/6/2024 9:52 AM
options_fit_SEIR_flu1918_dist1_0	MATLAB Code	2 KB	No	6 KB	70%	6/6/2024 9:52 AM
options_fit_SEIR_flu1918_dist1_1	MATLAB Code	2 KB	No	5 KB	70%	6/6/2024 9:52 AM
options_fit_SEIR_flu1918_dist1_3	MATLAB Code	2 KB	No	5 KB	70%	6/6/2024 9:52 AM
options_fit_SEIR_unreported_covid_sp...	MATLAB Code	2 KB	No	6 KB	69%	6/6/2024 9:52 AM
options_fit_SEIR_unreported_covid_sp...	MATLAB Code	2 KB	No	6 KB	69%	6/6/2024 9:52 AM
options_fit_SEIR1	MATLAB Code	2 KB	No	6 KB	70%	6/6/2024 9:52 AM
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options_forecast_SEIR_flu1918_dist1_0	MATLAB Code	2 KB	No	6 KB	71%	6/6/2024 9:52 AM
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Setting up the Working Directory



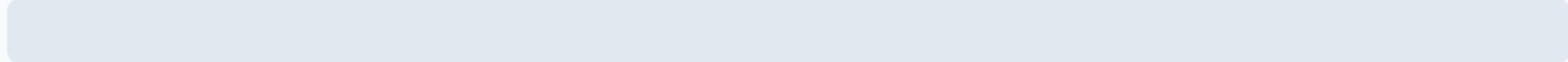






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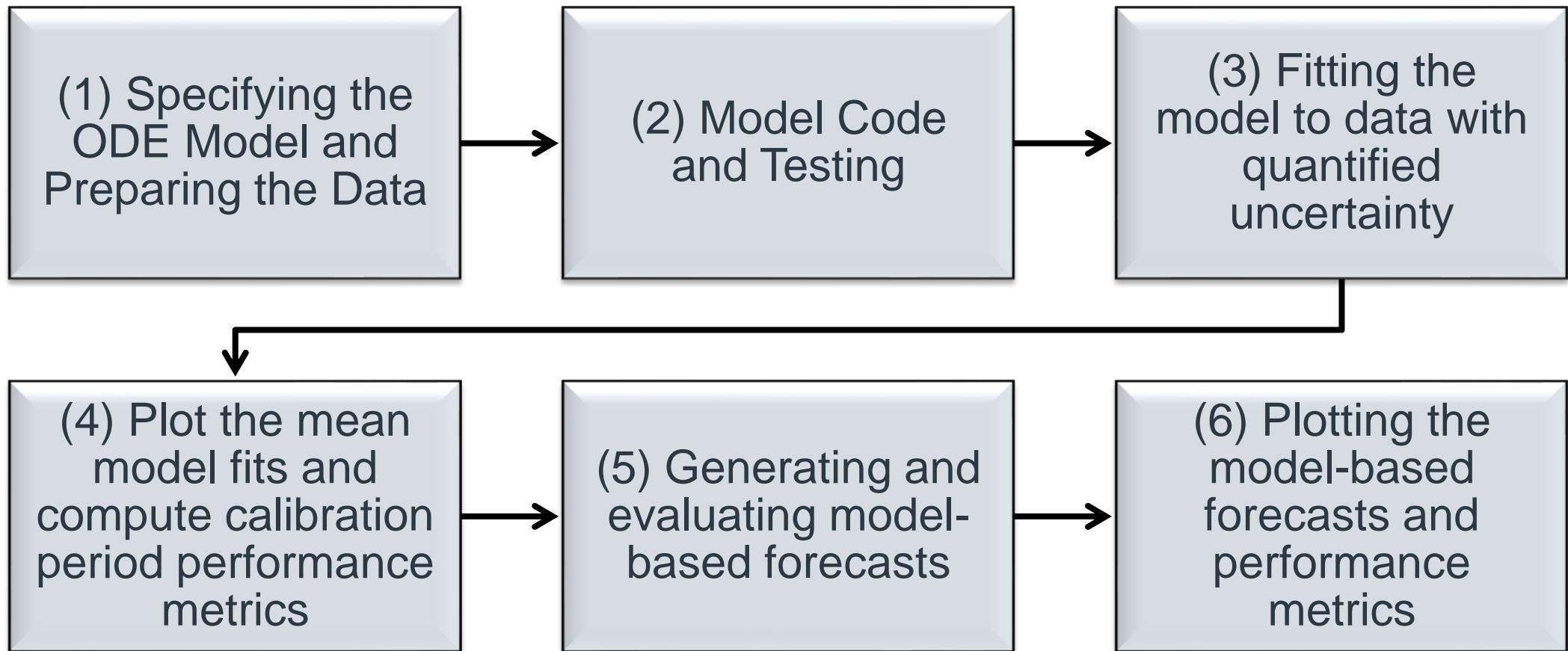


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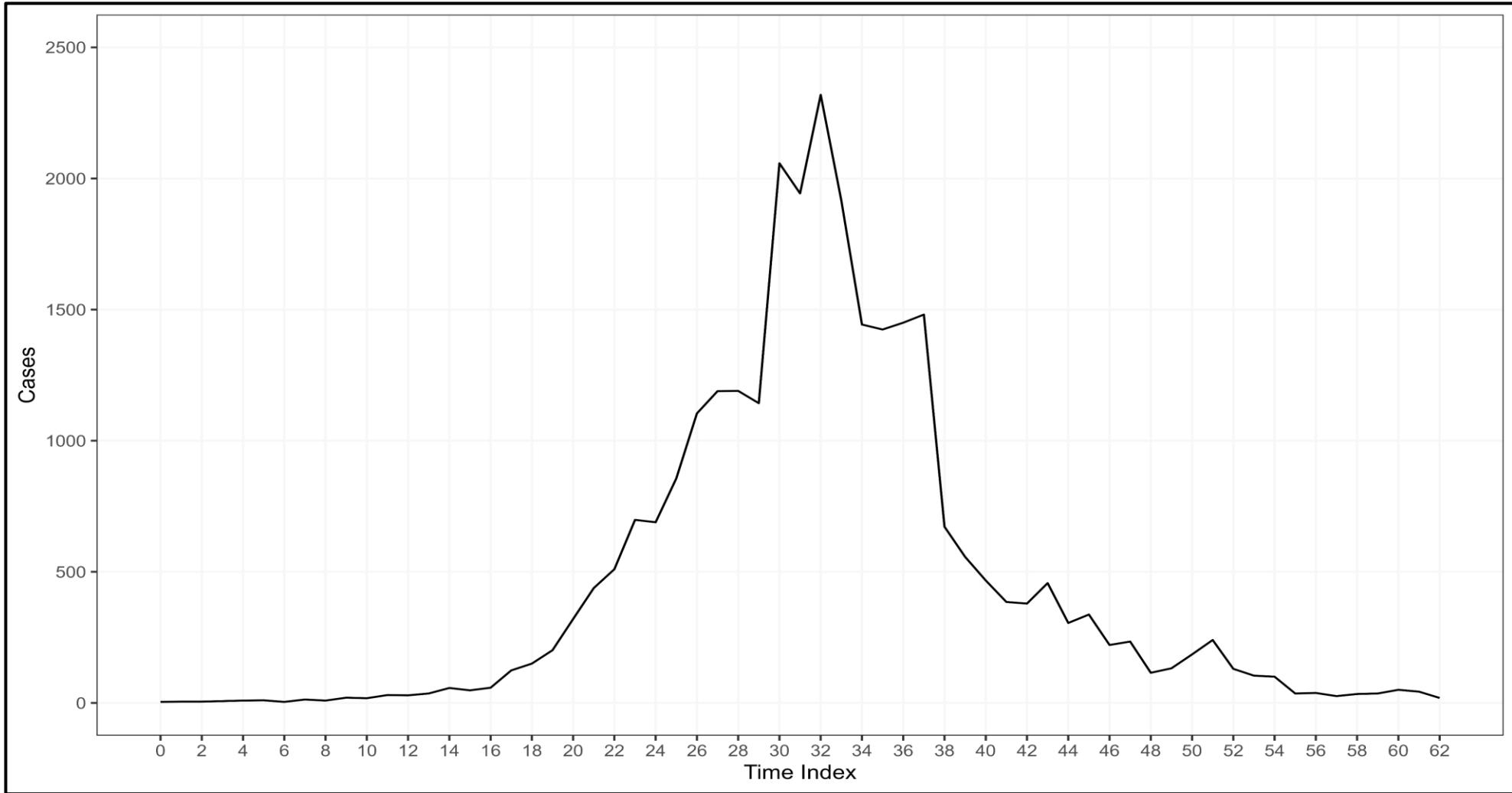
A horizontal progress bar consisting of a grey rectangular bar with rounded ends, positioned above a light grey rectangular area. The progress bar is currently empty, indicating 0% completion.

Preparing for the Tutorial

Workflow



Tutorial #1: 1918 Influenza



Daily incident curve of the fall wave of the 1918 influenza pandemic in San Francisco.

Preparing the data

Set-Up

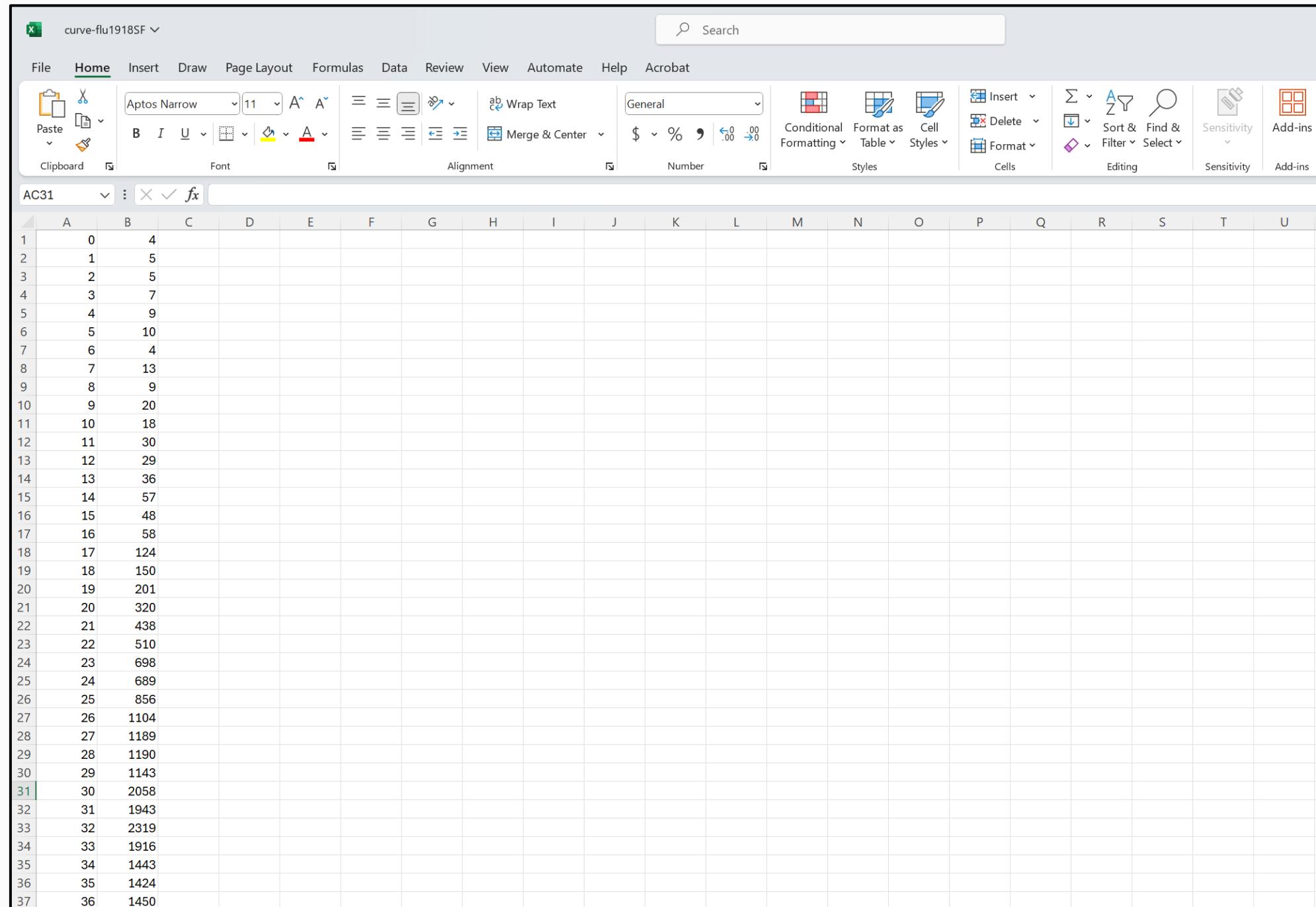
- Two folders are needed in your working directory: (1) input & (2) output
- Data goes in the input folder

File Name

- No formal file naming conventions must be followed IF working with incident data
- If working with CUMULATIVE incidence count data, the name of the time series data file must start with cumulative

File Structure

- 1st Column: Time Index (0, 1, 2...)
- 2nd Column: Temporal Incidence
- Columns should NOT have headers
- Data should be in *.txt* format



Specifying the ODE Model

Specifying the SEIR Model

```
% <=====
% < Author: Gerardo Chowell =====
% <=====

function dx=SEIR1(t,x,params0,extra0)

% params0(1) = beta, params0(2)=k,  params0(3)=gamma, params0(4)=N

dx=zeros(5,1); % define the vector of the state derivatives

dx(1,1)= -params0(1)*x(1,1).*x(3,1)./params0(4); %S
dx(2,1)= params0(1)*x(1,1).*x(3,1)/params0(4) -params0(2)*x(2,1); %E
dx(3,1)= params0(2)*x(2,1) - params0(3)*x(3,1); %I
dx(4,1)= params0(3)*x(3,1); %R
dx(5,1)= params0(2)*x(2,1); %cumulative infections
```

$$\begin{cases} \dot{S} = -\beta S(t) \frac{I(t)}{N} \\ \dot{E} = \beta S(t) \frac{I(t)}{N} - \kappa E(t) \\ \dot{I} = \kappa E(t) - \gamma I(t) \\ \dot{R} = \gamma I(t) \\ \dot{C} = \kappa E(t) \end{cases}$$

Specifying the R0 Function

- In addition to specifying the ODE model, users can choose to specify a composite function using parameters specified in the ODE file
- For this tutorial, we use the following calculation for R0:

```
function composite_val=R0s(params0)  
  
% beta(1), k(2), gamma(3), N(4)  
composite_val=params0(:,1)./params0(:,3);
```

$$R0 = \frac{\beta}{\gamma}$$

Preparing the options_fit.m file

Overview

- Prior to fitting the model to data, multiple parameters must be specified in an `options_fit.m` structured file.
 - Name of `options_fit` file can be customized.
- Required Specification Sections :
 - Dataset Properties
 - Parameter Estimation
 - ODE Model
 - Rolling Window Analysis

Tutorial: We will be using two different `options_fit.m` files for this tutorial:

- (1) `options_fit_SEIR_flu1918_dist1_1`
- (2) `options_fit_SEIR_flu1918_dist1_3`

```

% <===== Declare global variables =====>
% <===== Declare global variables =====>
% <===== Declare global variables =====>

global method1 % Parameter estimation method

% <===== Datasets properties =====>
% <===== Datasets properties =====>
% <===== Datasets properties =====>
% Located in the input folder, the time series data file is a text file with extension *.txt.
% The time series data file contains the incidence curve of the epidemic of interest.
% The first column corresponds to time index: 0,1,2, ... and the second
% column corresponds to the observed time series data.

cadfilename1='curve-flu1918SF'

caddisease='1918 Flu'; % string indicating the name of the disease related to the time series data

datatype='cases'; % string indicating the nature of the data (cases, deaths, hospitalizations, etc)

```

- (1) `cadfilename1`: The name of the data file.
- (2) `caddisease`: Name of the process of interest
- (3) `datatype`: Nature of the data

```

% ===== Parameter estimation =====
method1=1; % Type of estimation method
% Nonlinear least squares (LSQ)=0,
% MLE Poisson=1,
% MLE (Neg Binomial)=3, with VAR=mean+alpha*mean;
% MLE (Neg Binomial)=4, with VAR=mean+alpha*mean^2;
% MLE (Neg Binomial)=5, with VAR=mean+alpha*mean^d;

dist1=1; % Define dist1 which is the type of error structure. See below:

%dist1=0; % Normal distribution to model error structure (method1=0)
%dist1=1; % Poisson error structure (method1=0 OR method1=1)
%dist1=2; % Neg. binomial error structure where var = factor1*mean where
          % factor1 is empirically estimated from the time series
          % data (method1=0)
%dist1=3; % MLE (Neg Binomial) with VAR=mean+alpha*mean (method1=3)
%dist1=4; % MLE (Neg Binomial) with VAR=mean+alpha*mean^2 (method1=4)
%dist1=5; % MLE (Neg Binomial)with VAR=mean+alpha*mean^d (method1=5)

switch method1
  case 1
    dist1=1;
  case 3
    dist1=3;
  case 4
    dist1=4;
  case 5
    dist1=5;
end

numstartpoints=20; % Number of initial guesses for optimization procedure using MultiStart
B=300; % number of bootstrap realizations to characterize parameter uncertainty

```

(1) method1 : The type of estimation method.

- Five options available: NLSQ, MLE Poisson, and three MLE Neg. Binomial options
- The estimation method option should match the appropriate error structure.

(2) dist1: Error structure

- Six options available: Normal, Poisson, and four Neg. Binomial options

(3) numstartpoints: Number of initial guesses for optimization procedure

(4) B: Number of bootstrap realizations

*Tutorial: As we are working with smaller case counts, we will stick with Poisson and Negative Binomial error structures.

Specifying the ODE Model – options_fit.m

```
% <===== ODE model =====>
% <===== ODE model =====>
% <===== ODE model =====>

model.fc=@SEIR1; % name of the model function
model.name='SEIR model'; % string indicating the name of the ODE model

params.label={'\beta','\kappa','\gamma','N'}; % list of symbols to refer to the model parameters
params.LB=[0.01 0.01 0.01 20]; % lower bound values of the parameter estimates
params.UB=[10 2 2 1000000]; % upper bound values of the parameter estimates
params.initial=[0.6 1/1.9 1/4.1 550000]; % initial parameter values/guesses
params.fixed=[0 1 1 1]; % Boolean vector to indicate any parameters that should remain fixed (1) to initial values indicated in params.initial. Otherwise the parameter is estimated.
params.fixI0=1; % Boolean variable indicating if the initial value of the fitting variable is fixed according to the first observation in the time series (1). Otherwise, it is estimated.
params.composite=@R0s; % Estimate a composite function of the individual model parameter estimates otherwise it is left empty.
params.composite_name='R_0'; % Name of the composite parameter
params.extra0=[];

vars.label={'S','E','I','R','C'}; % list of symbols to refer to the variables included in the model
vars.initial=[params.initial(4)-4 0 4 0 4]; % vector of initial conditions for the model variables
vars.fit_index=5; % index of the model's variable that will be fit to the observed time series data
vars.fit_diff=1; % boolean variable to indicate if the derivative of model's fitting variable should be fit to data.
```

```
% <===== Parameters of the rolling window analysis =====>
% <===== Parameters of the rolling window analysis =====>
% <===== Parameters of the rolling window analysis =====>

windowsize1=17; % moving window size

tstart1=1; % time point for the start of rolling window analysis

tend1=1; %time point for the end of the rolling window analysis

printscreen1=1;
```

- (1) windowsize1: The size of the moving window or calibration period
- (2) tstart1: Start of the rolling window analysis
- (3) tend1: End of the rolling window analysis
- (4) printscreen1: Show the plots at the conclusion of running code

Rolling Window Analysis

- A rolling window analysis can be useful to assess the stability of the model parameters and forecasts over
 - Controls the data used to calibration the model (i.e., data that goes into the model)

Example: tstart1 = 1, tend1 = 2, windowsize1 = 5

Time	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
Cases	1	2	5	8	9	1	2	1	1	5	3	6	9	4	8	0

Rolling Window #1

Rolling Window #2

Given $tstart1 = 3$, $tend1 = 3$, and $windowsize1 = 5$, what is the start and end time index of the rolling window analysis?

0

(A) 3, 3

0%

(B) 1, 2

0%

(C) 1, 3

0%

(D) 4, 5

0%

(E) I'm not sure.

0%

Time	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
Cases	1	2	5	8	9	1	2	1	1	5	3	6	9	4	8	0

Generating preliminary model solutions

plotODEModel()

Overview

- It's often helpful to check that the user has correctly specified the model by checking that the model's solutions for the parameter ranges specified by the user correspond to a broad range of expected solutions

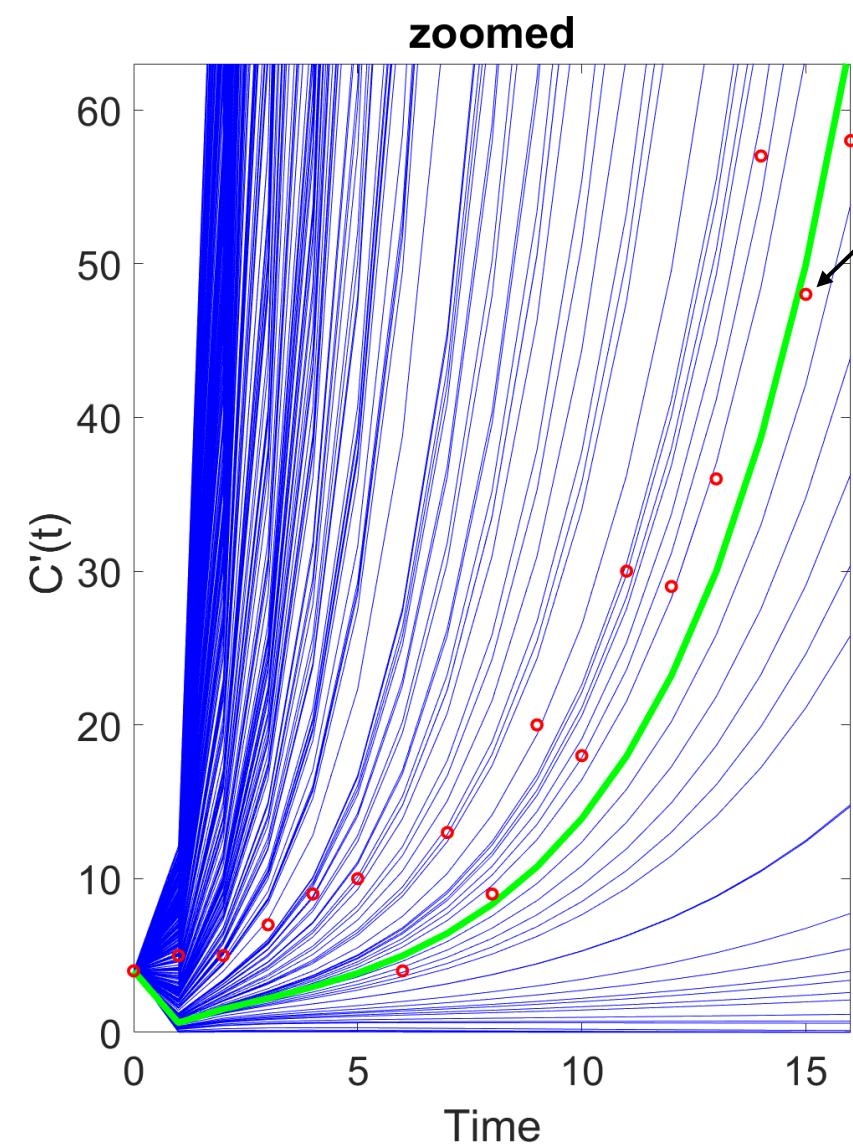
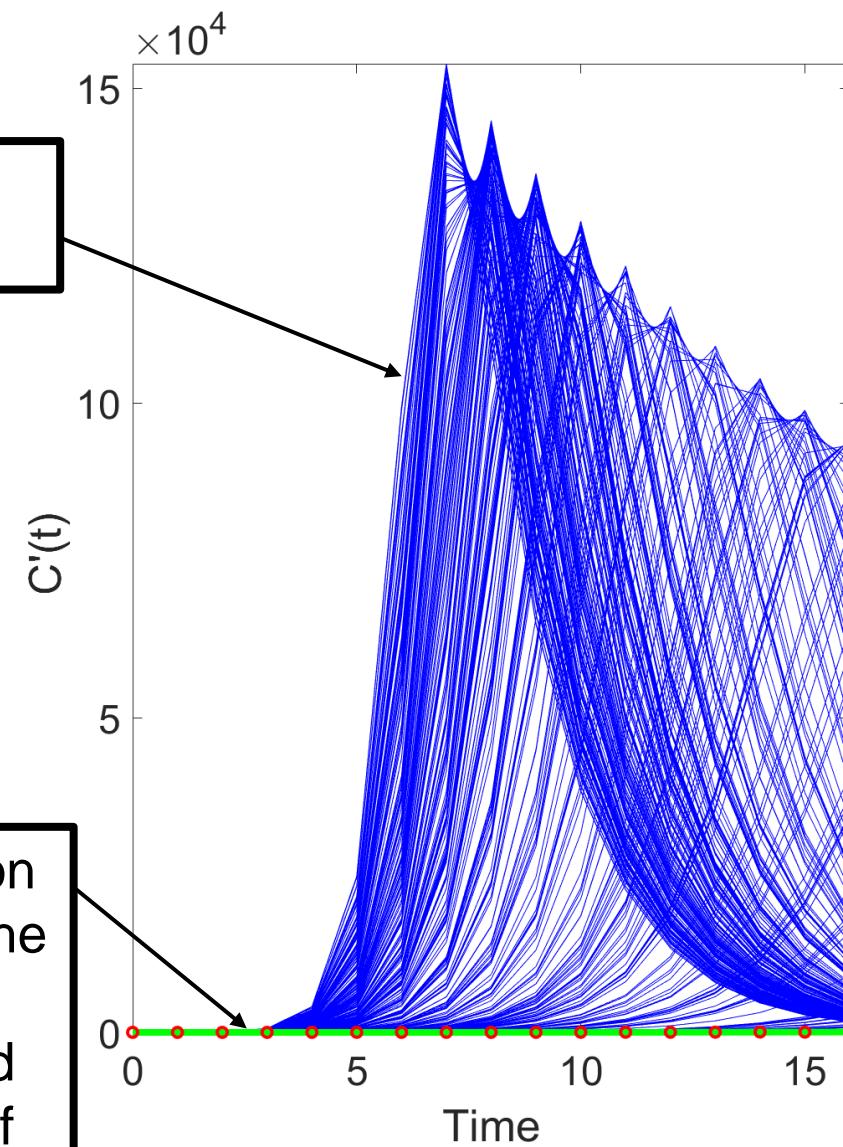
```
plotODEModel (@optionsFitFileName)
```

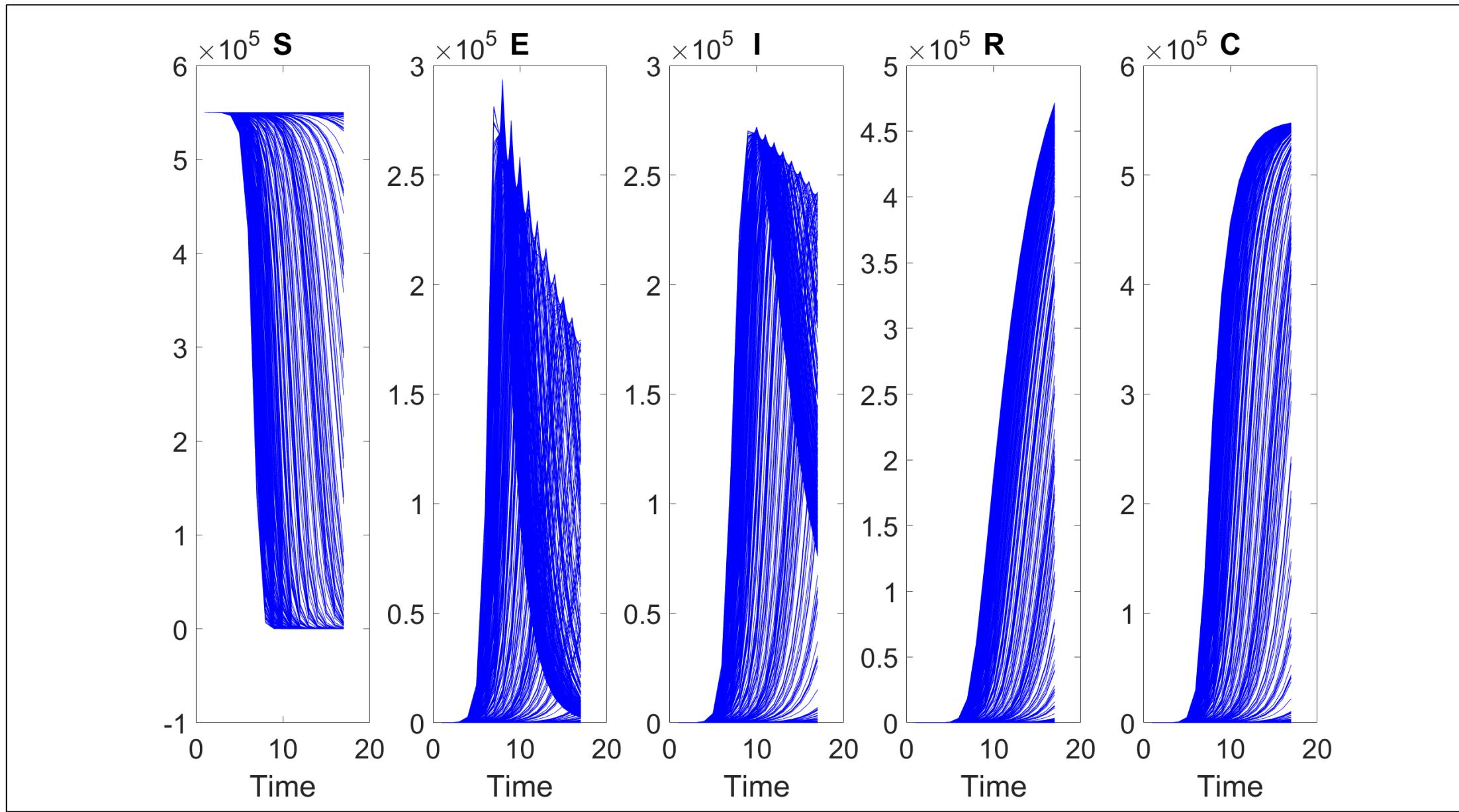
- Produces plots showing model solutions of state variables by generating a random sample of parameter sets from the specified parameter ranges and time-period

Tutorial Code Call

```
plotODEModel (@options_fit_SEIR_flu1918_dist1_1)
```

Time-series
data





Fitting, plotting, and evaluating the model with quantified uncertainty

Poisson and Neg. Binomial Error Structures

Step One: Fitting the Model to Data

- Prior to obtaining model fit files and plotting the model fit, we first must fit the model to data with quantified uncertainty using the following code call:

```
Run_Fit_ODEModel (@OptionsFitFileName, tstart1, tend1, windowsize1)
```

- `tstart1`, `tend1`, and `windowsize1` correspond to the values entered in the rolling window analysis section of the `options_fit.m` file.

Tutorial Code: (1) Poisson & (2) Negative Binomial

```
(1) Run_Fit_ODEModel (@options_fit_SEIR_flu1918_dist1_1, 1, 1, 17)  
(2) Run_Fit_ODEModel (@options_fit_SEIR_flu1918_dist1_3, 1, 1, 17)
```

Step Two: Plotting the model fit

- After obtaining the model fits, we can obtain information related to: (1) model fit, (2) model parameter estimates, (3) Monte Carlo standard errors, (4) AICc values, (5) calibration performance metrics, & (6) composite parameter information for the specified model and data using the following call (.csv files in output)

```
plotFit_ODEModel (@OptionsFitFileName, tstart1, tend1, windowsize1)
```

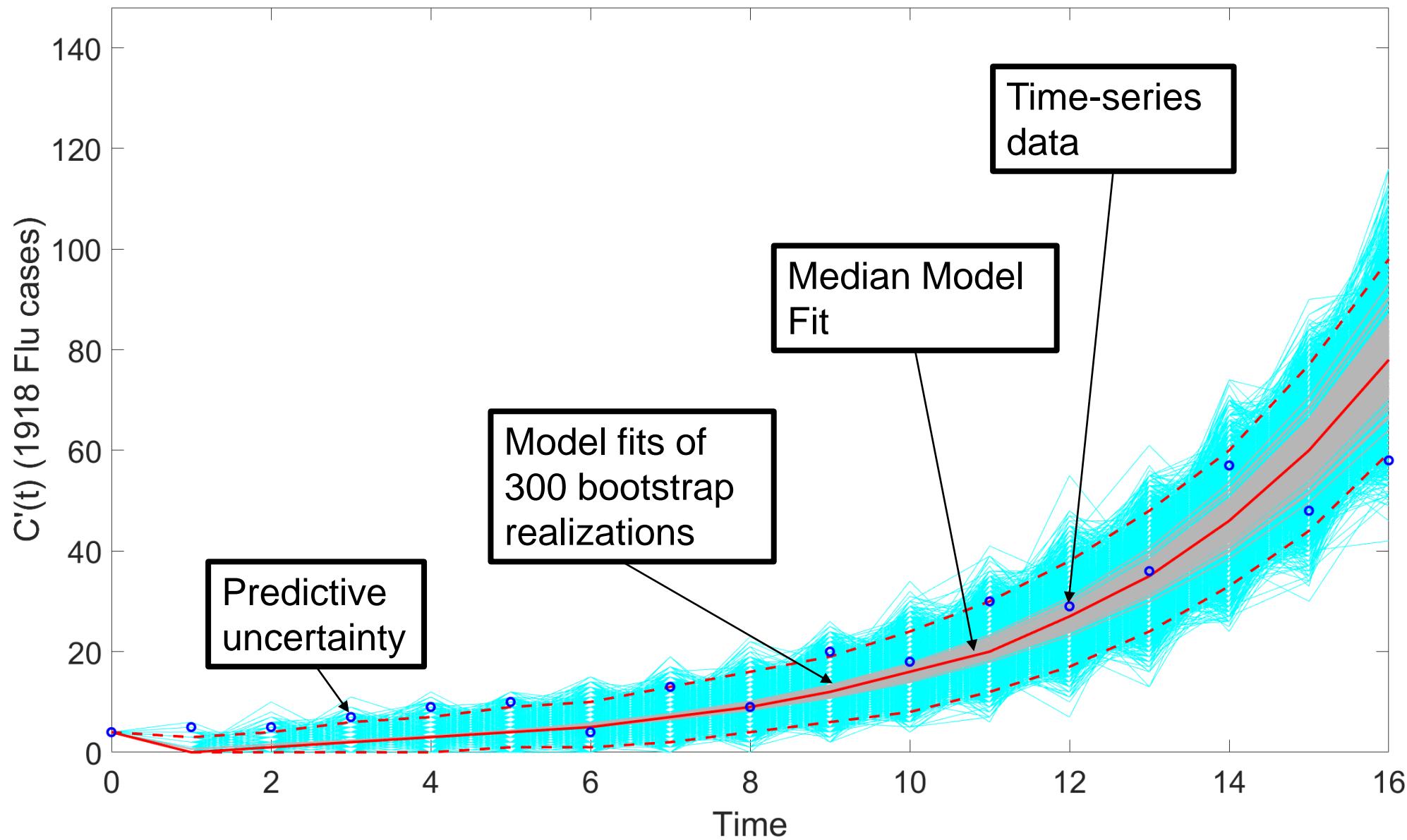
- `tstart1`, `tend1`, and `windowsize1` correspond to the values entered in the rolling window analysis section of the `options_fit.m` file.

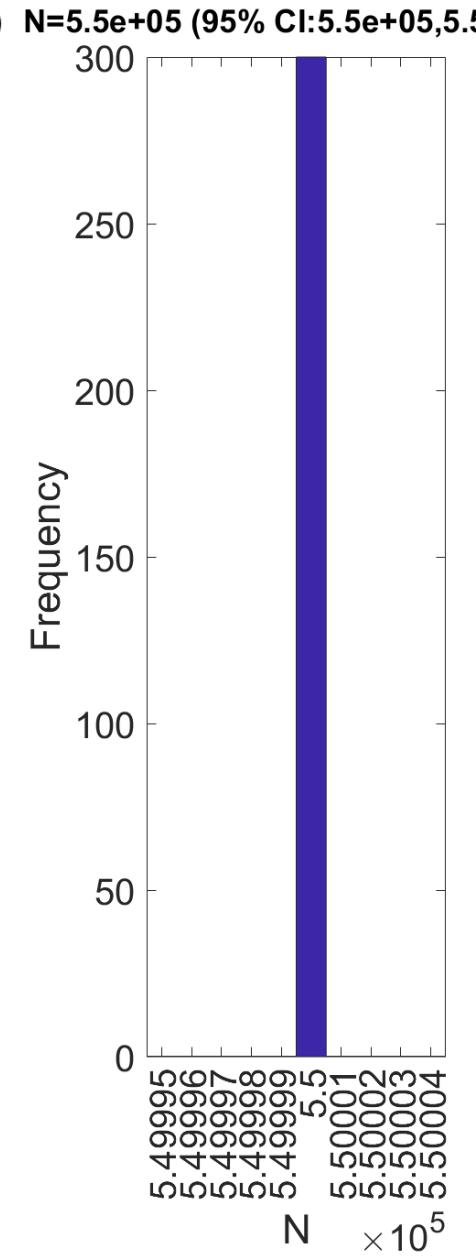
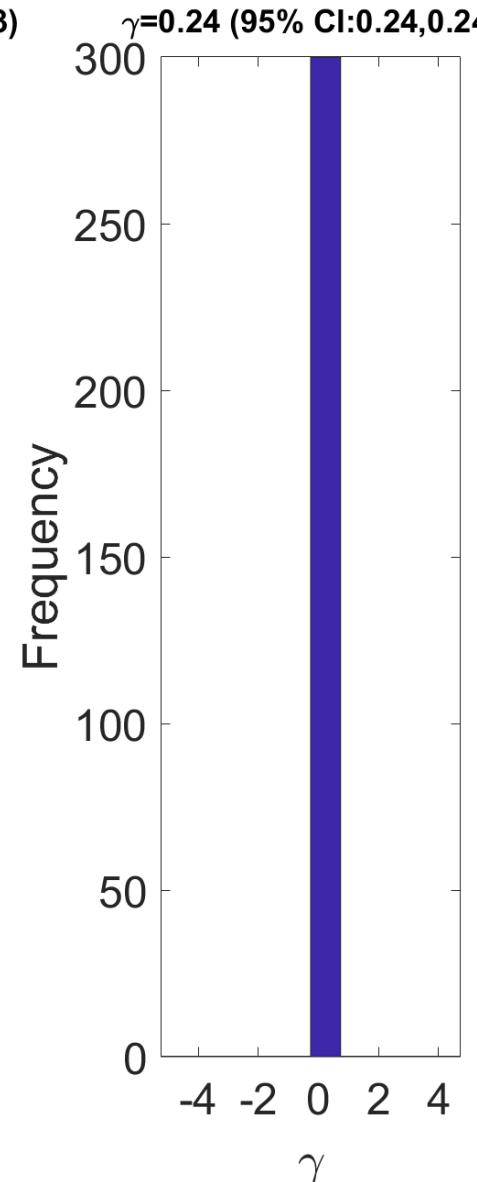
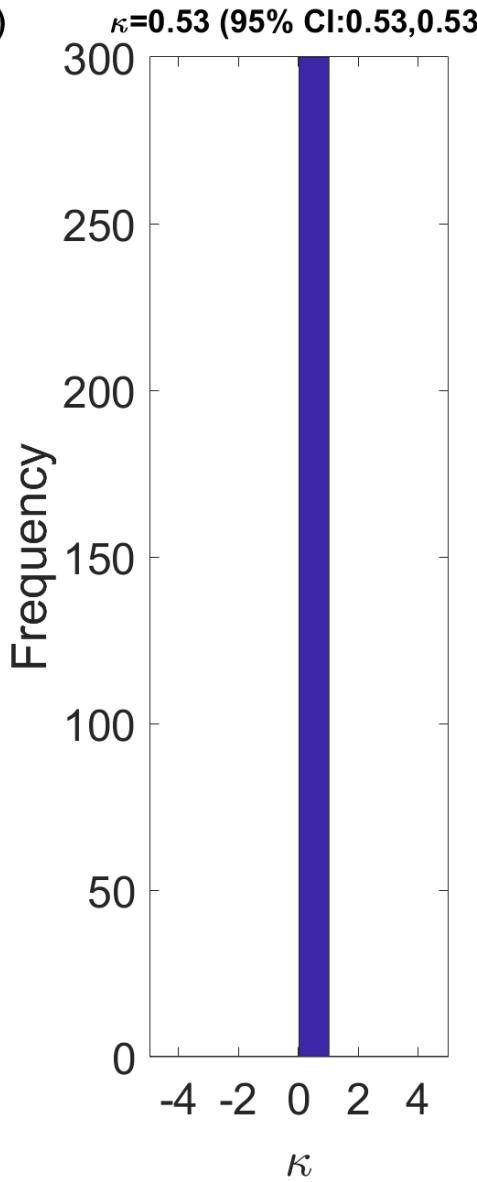
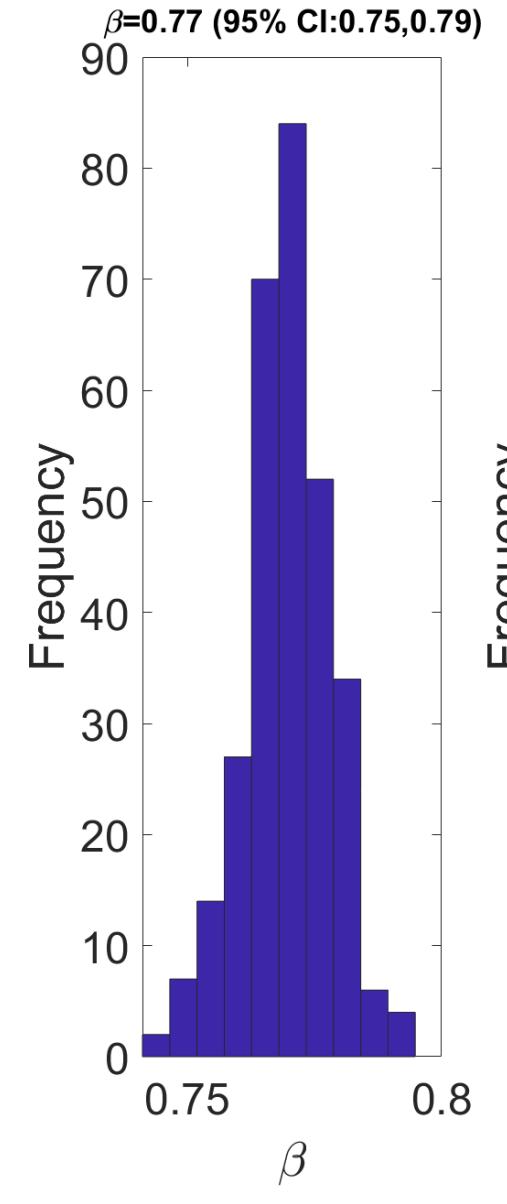
Tutorial Code Call: (1) Poisson & (2) Negative Binomial

```
(1) plotFit_ODEModel (@options_fit_SEIR_flu1918_dist1_1, 1, 1, 17)  
(2) plotFit_ODEModel (@options_fit_SEIR_flu1918_dist1_3, 1, 1, 17)
```

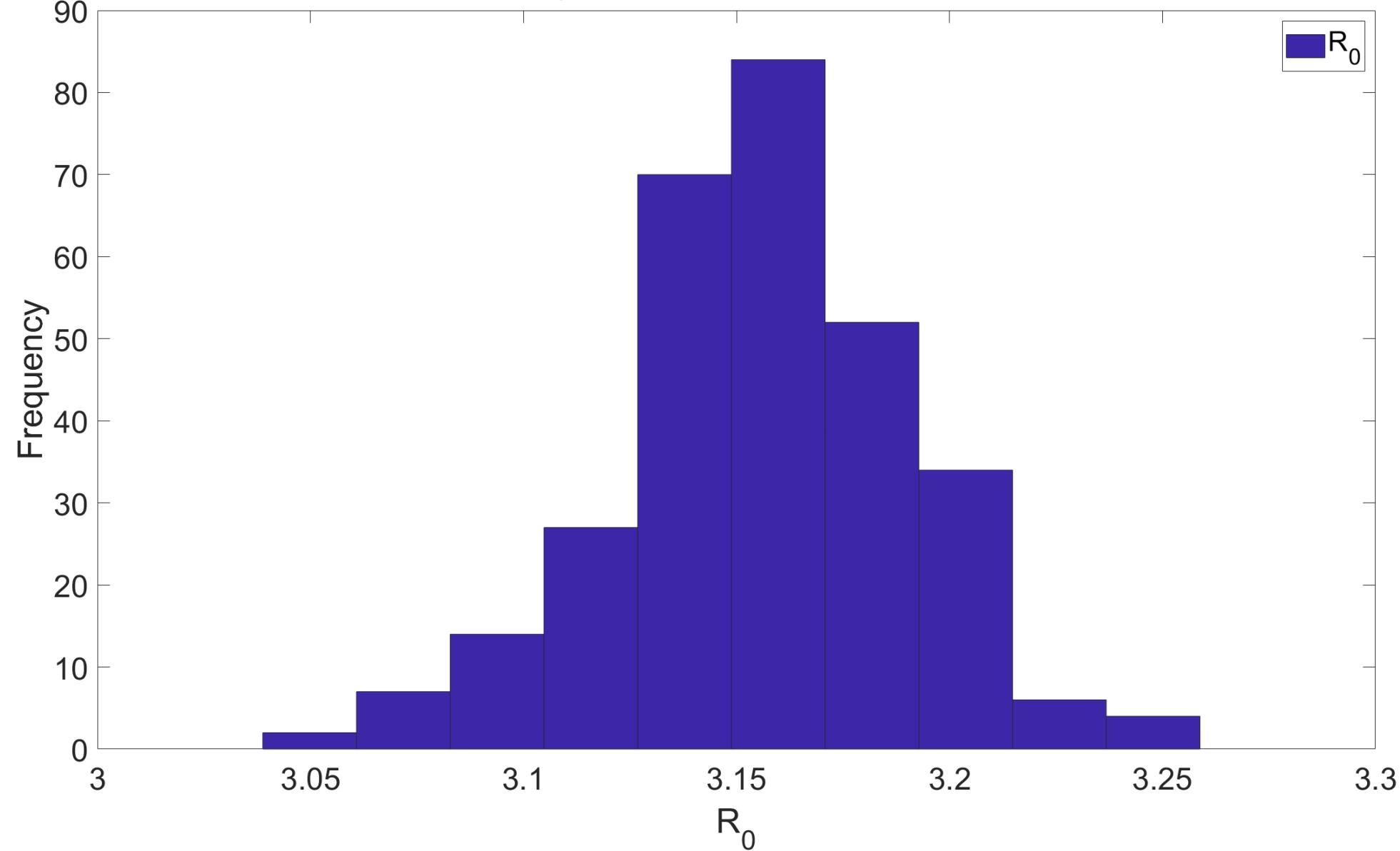
Poisson Output <dist1 = 1>

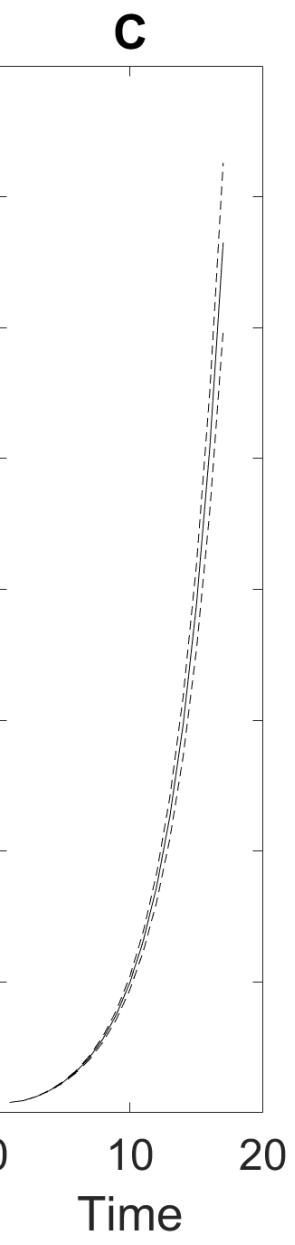
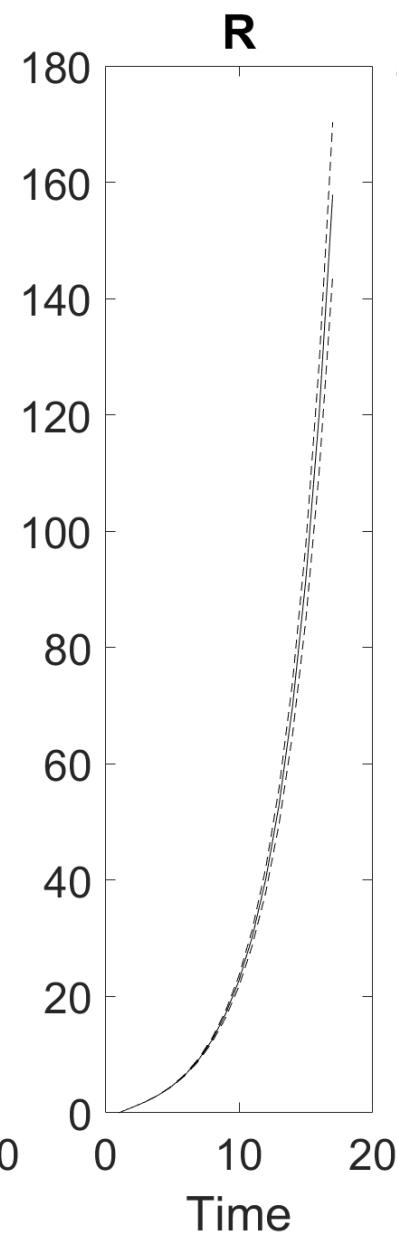
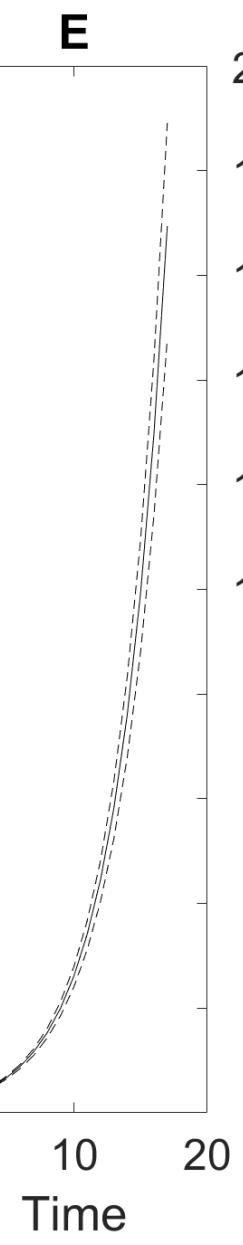
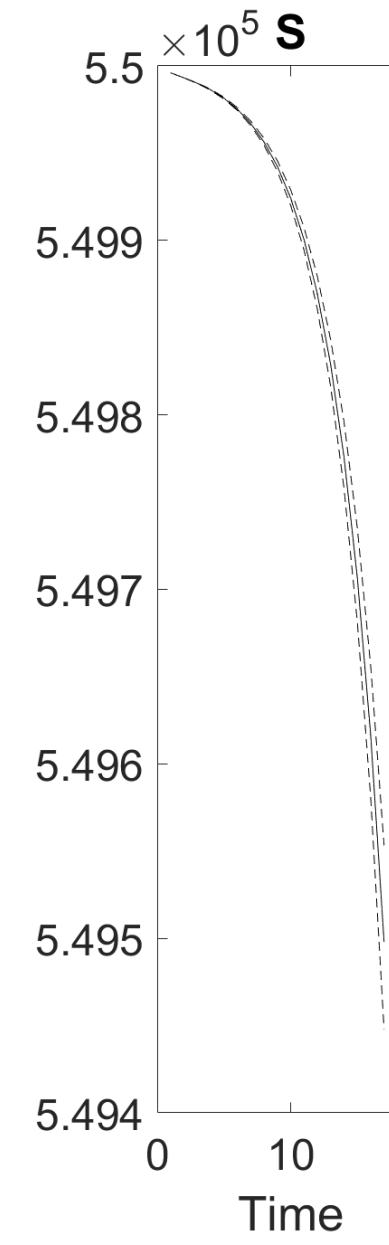
SEIR model





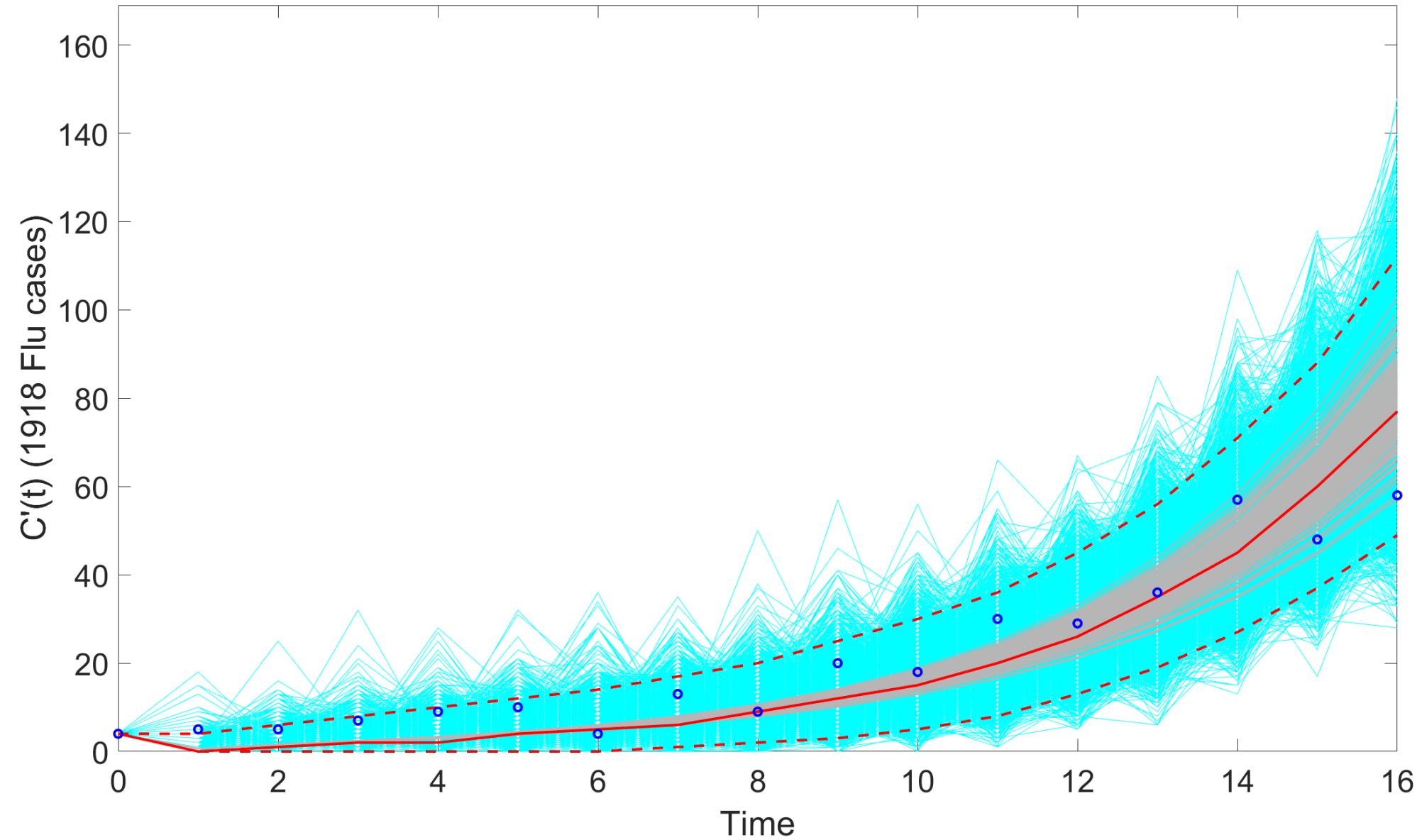
$$R_0=3.16 \text{ (95%CI:3.08,3.22)}$$

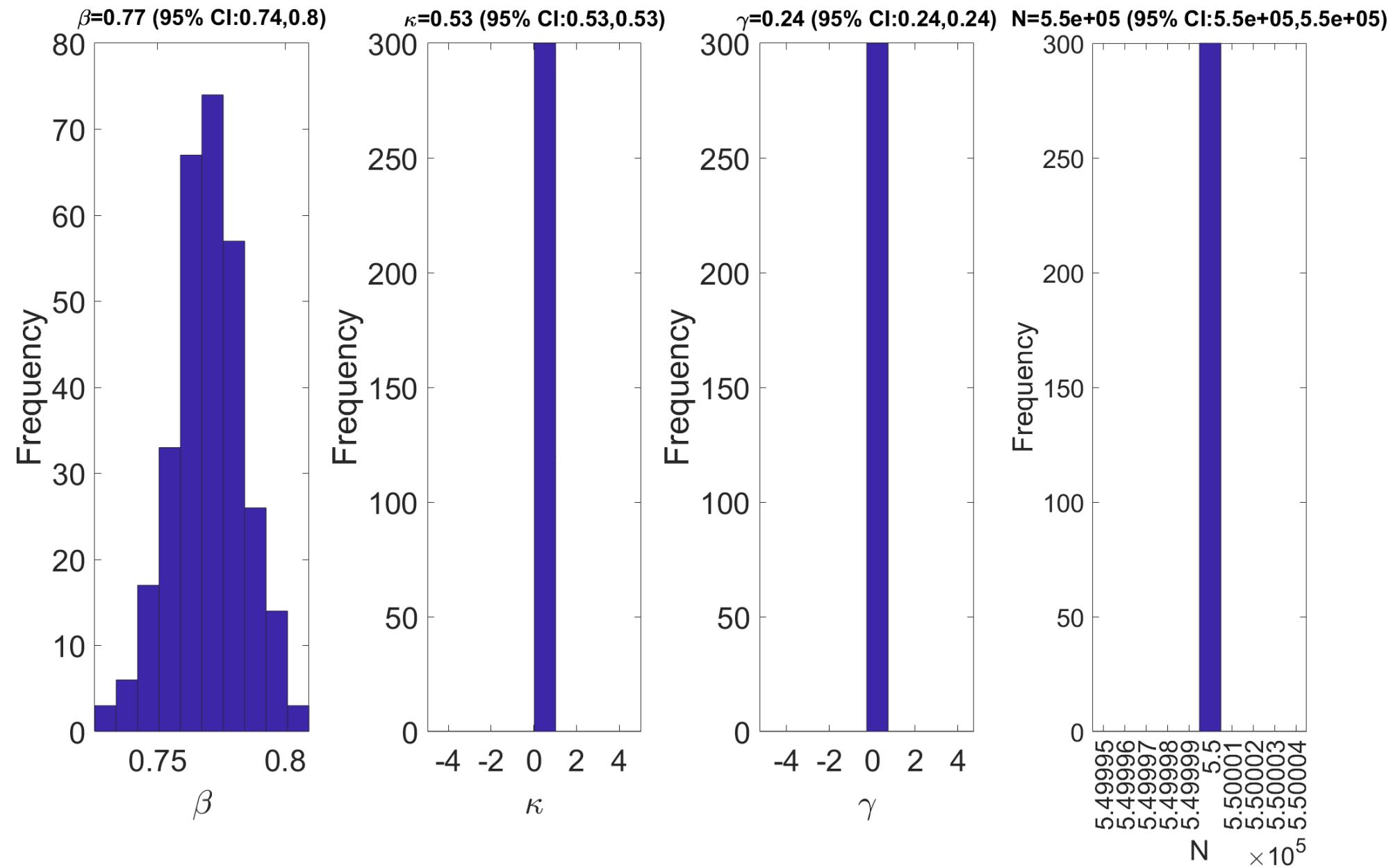


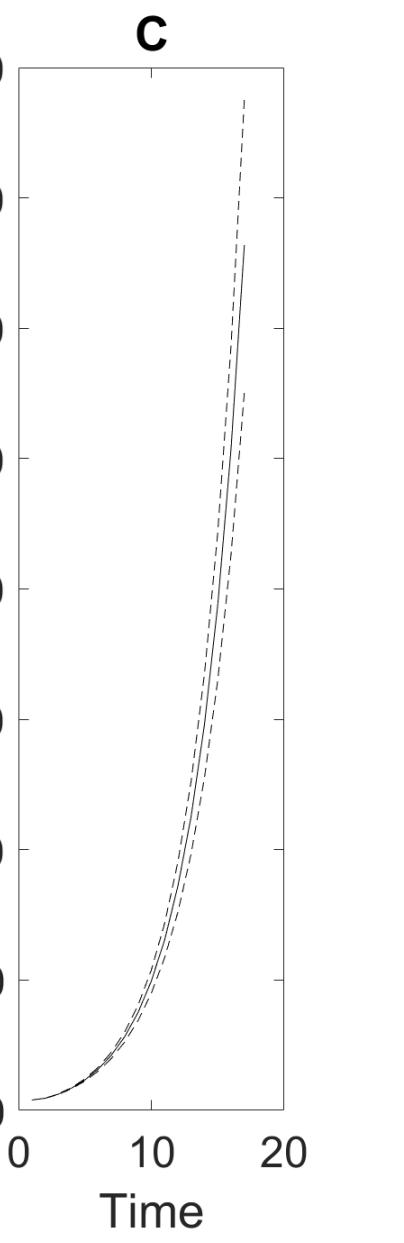
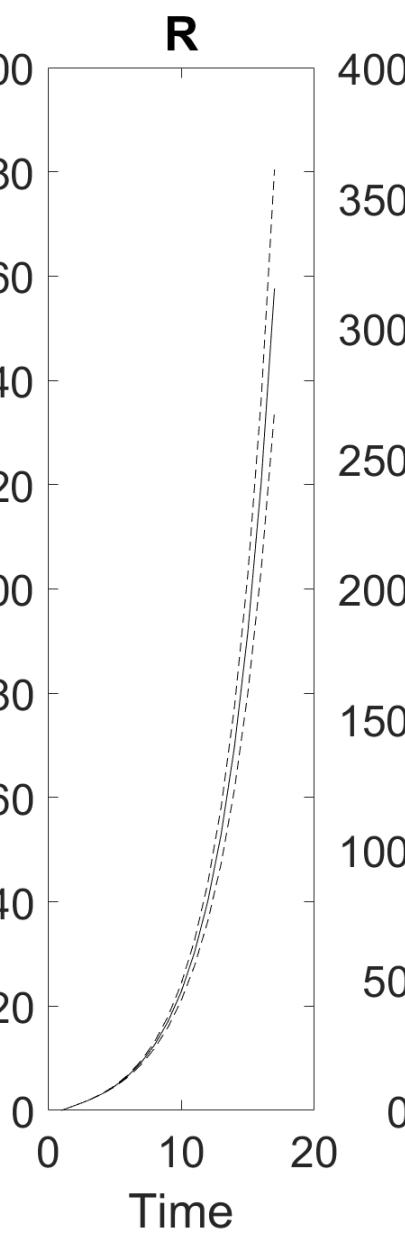
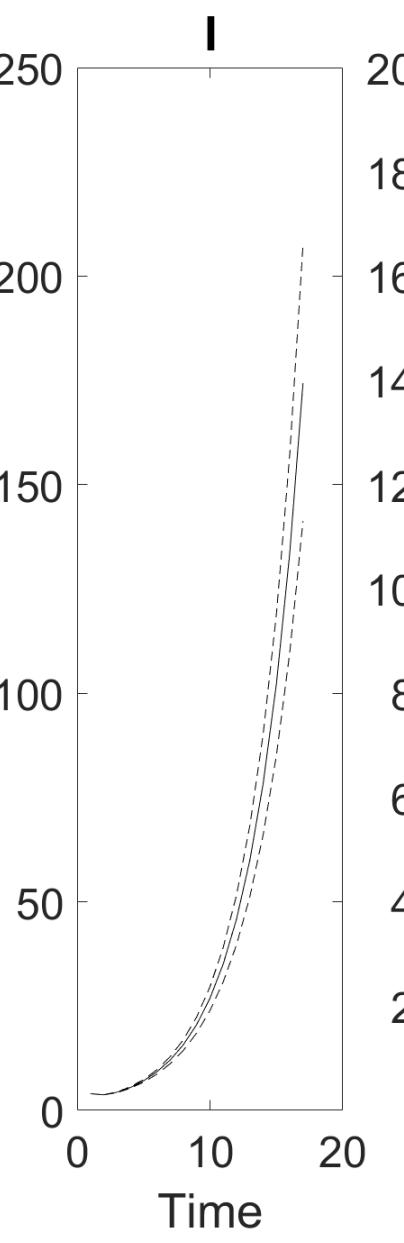
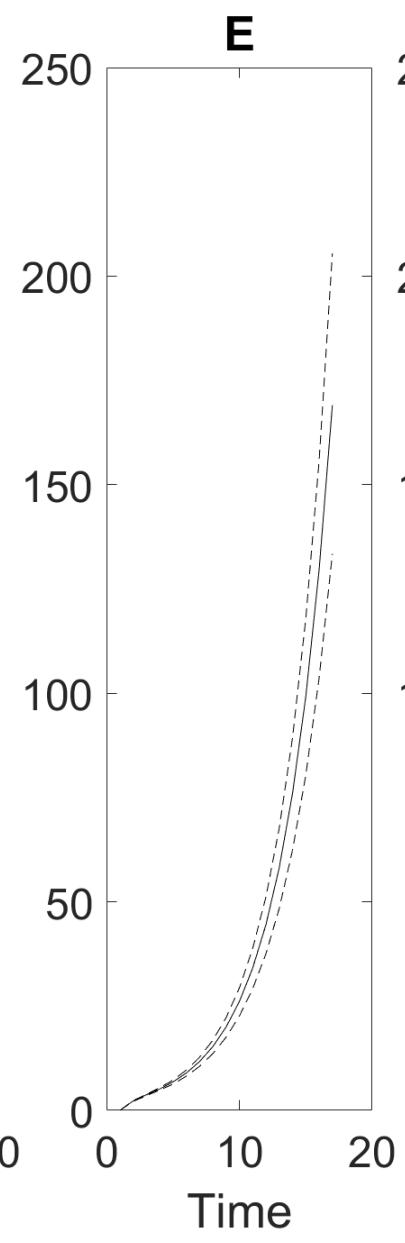
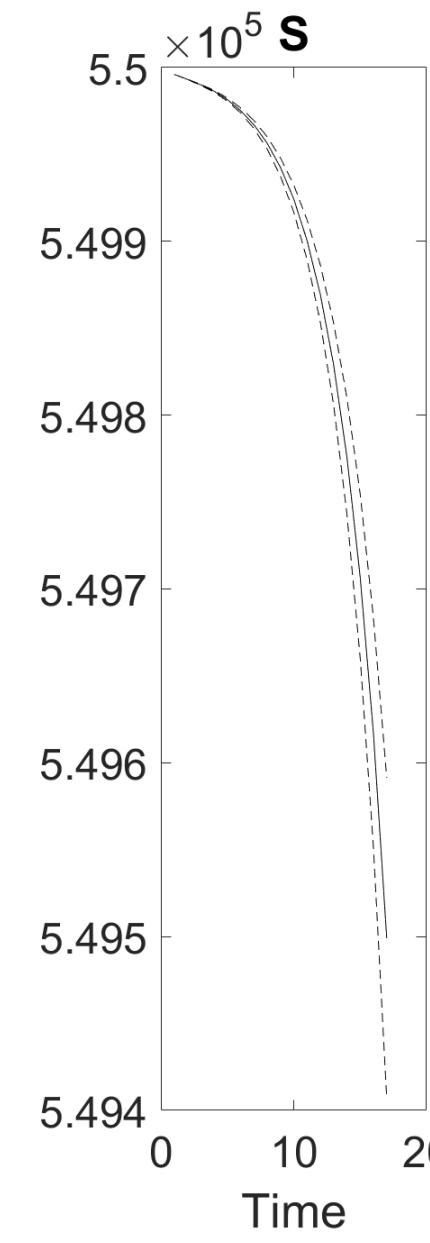


Neg. Binomial Output <dist1 = 3>

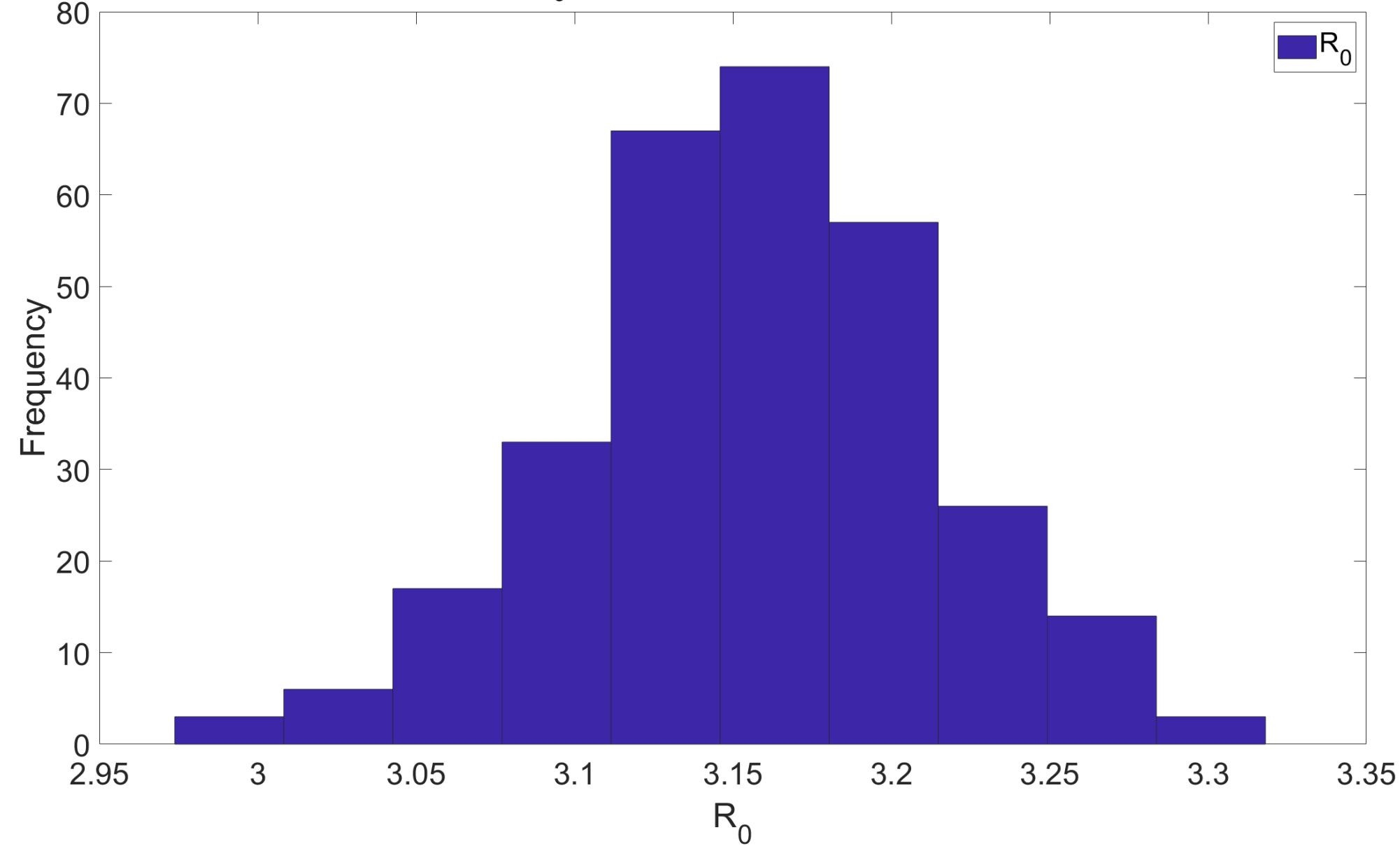
SEIR model





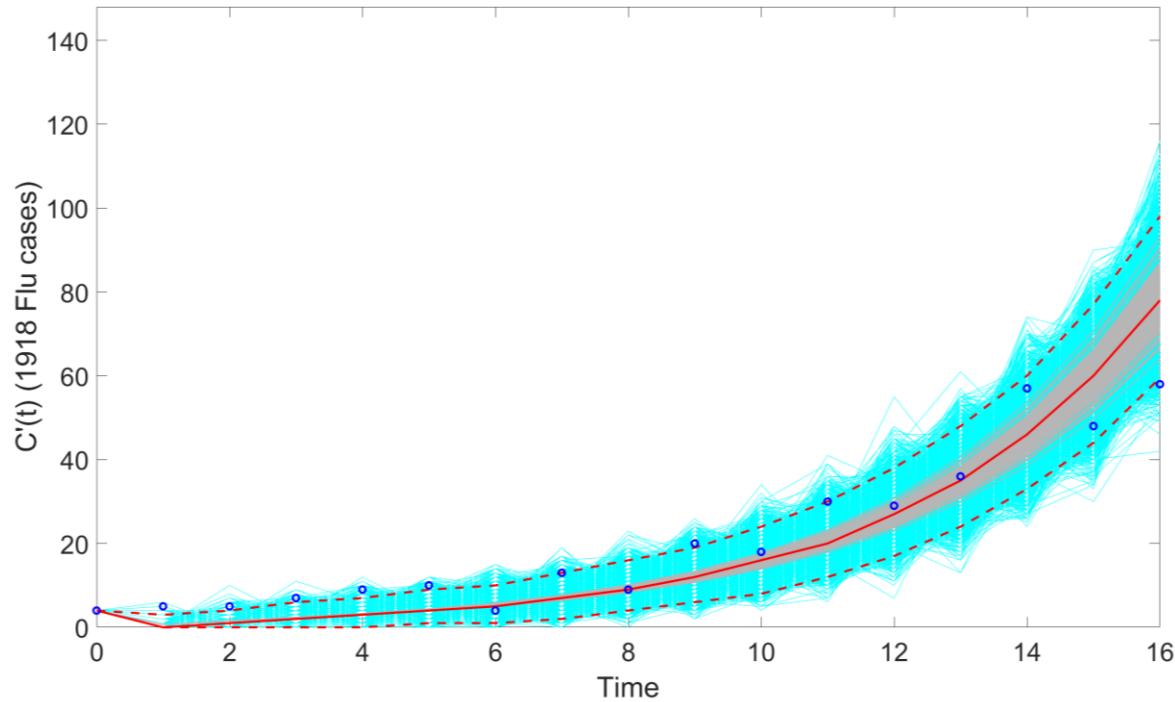


$$R_0=3.15 \text{ (95%CI:3.02,3.27)}$$



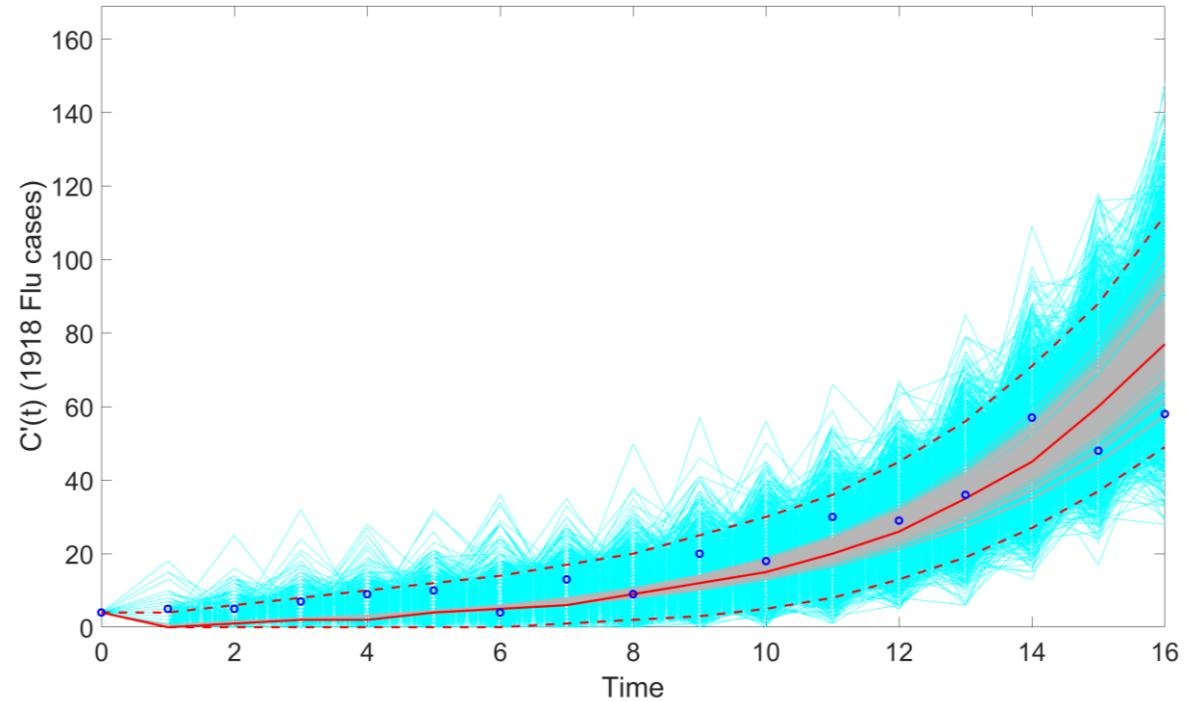
Model Fit Comparison

SEIR model

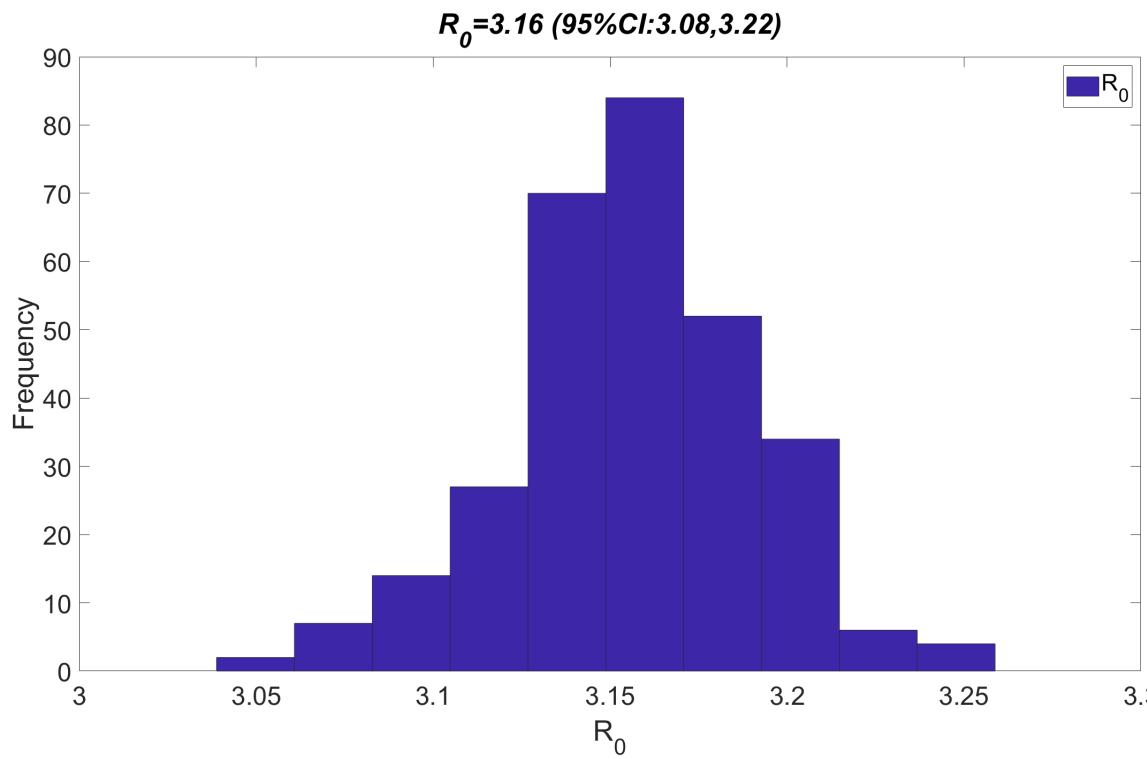


Poisson

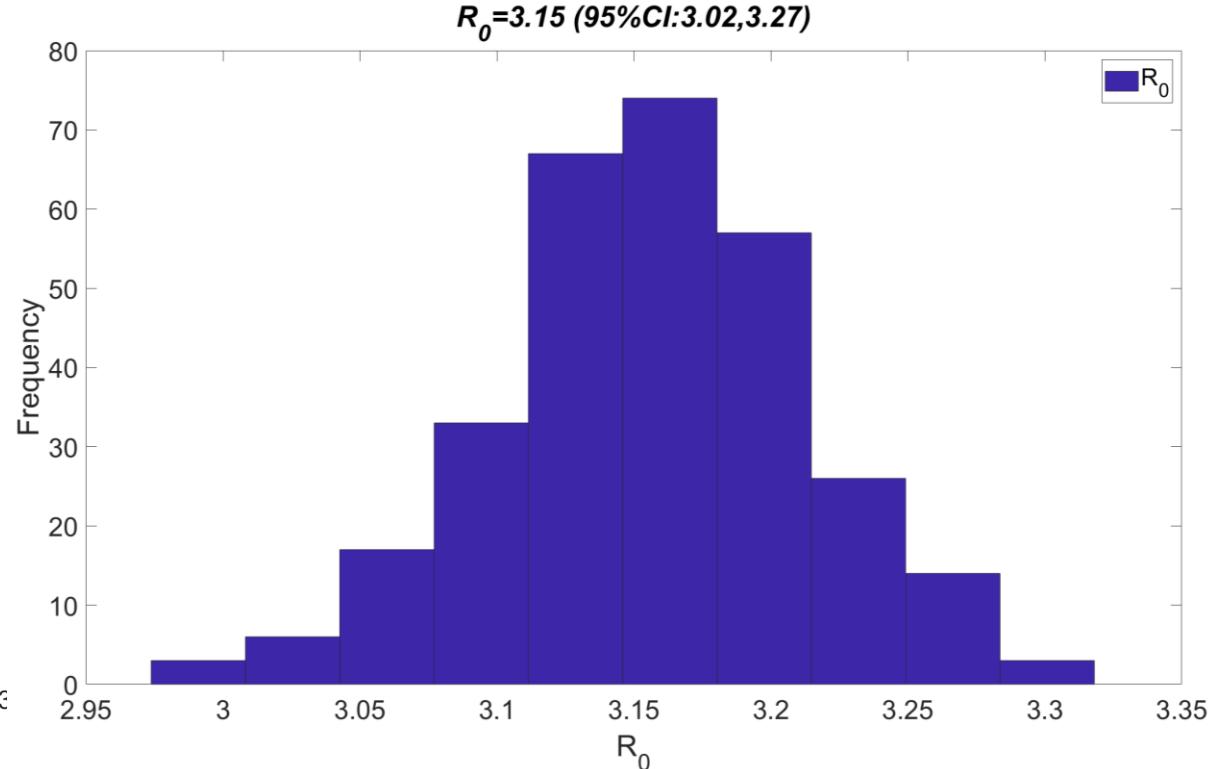
SEIR model



Neg. Binomial



Poisson



Neg. Binomial

Model	MAE	MSE	Coverage 95% PI	WIS
SEIR model with Poisson error structure ($\langle \text{dist1} \rangle = 1$)	5.72	58.59	58.82	3.87
SEIR model with negative binomial error structure ($\langle \text{dist1} \rangle = 3$)	5.73	58.21	94.12	3.67

* Obtained from .csv performance-calibration... files

Generating, plotting and assessing model-based forecasts

Neg. Binomial Error Structure & Exponential Model

Step One: Preparing the Code

```
% <===== Forecasting parameters =====>
% <===== Forecasting parameters =====>
% <===== Forecasting parameters =====>

getperformance=1; % flag or indicator variable (1/0) to calculate forecasting performance or not

forecastingperiod=10; % forecast horizon (number of time units ahead)
```

- (1) `getperformance`: Indicator for calculating forecast performance metrics.
 - If the forecasting period extends past the available data, `getperformance` must be marked zero, or an error will occur.
- (2) `forecastingperiod`: The forecasting horizon (i.e., how many time points ahead the toolbox should predict)

File Name: options_forecast_SEIR_flu1918_dist1_3

Step Two: Generating Forecasts

- Prior to obtaining model fits, forecasts, and associated evaluation criteria we must generate forecasts using the following code call:

```
Run_Forecasting_ODEModels(@OptionsForecastFileName,tstart1,tend1  
                           ,windowsize1, forecastingperiod)
```

- tstart1, tend1, windowsize1, and forecastingperiod correspond to the values entered in the rolling window analysis section of the options_forecast.m file.

Tutorial Code: (1) SEIR Negative Binomial

```
(1) Run_Forecasting_ODEModels(@options_forecast_SEIR_flu1918_dist1_3  
                               , 1, 1, 17,10)
```

Step Three: Plotting Forecasts

- After generating the model forecasts, we can obtain information related to: (1) forecasts, (2) model parameter estimates, (3) Monte Carlo standard errors, (4) AICc values, (5) calibration performance metrics, (6) forecast performance metrics, & (7) composite parameter information for the specified model and data using the following call (.csv files in output)

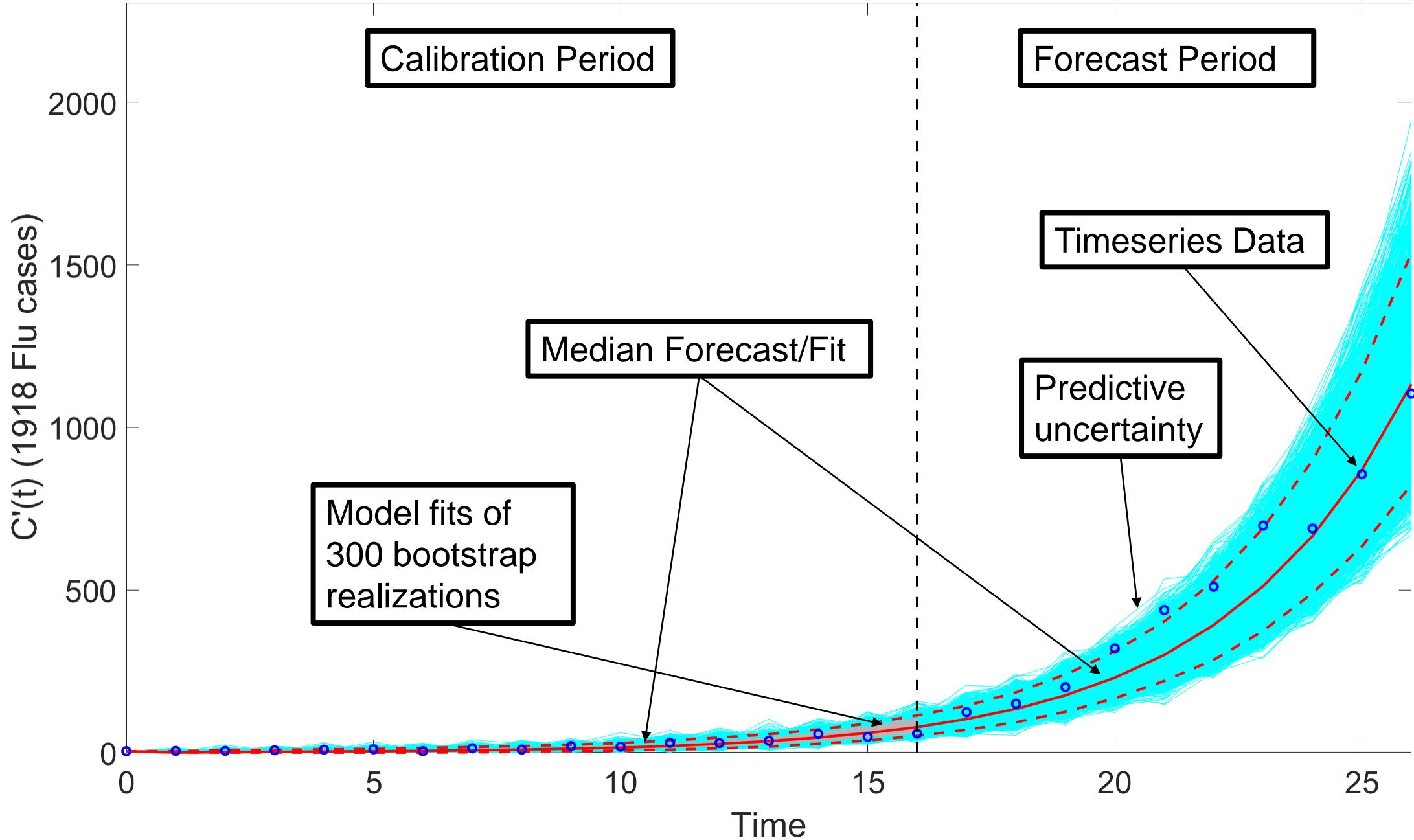
```
plotForecast_ODEModel (@OptionsForecastFileName, tstart1, tend1, windowSize1, forecastingPeriod)
```

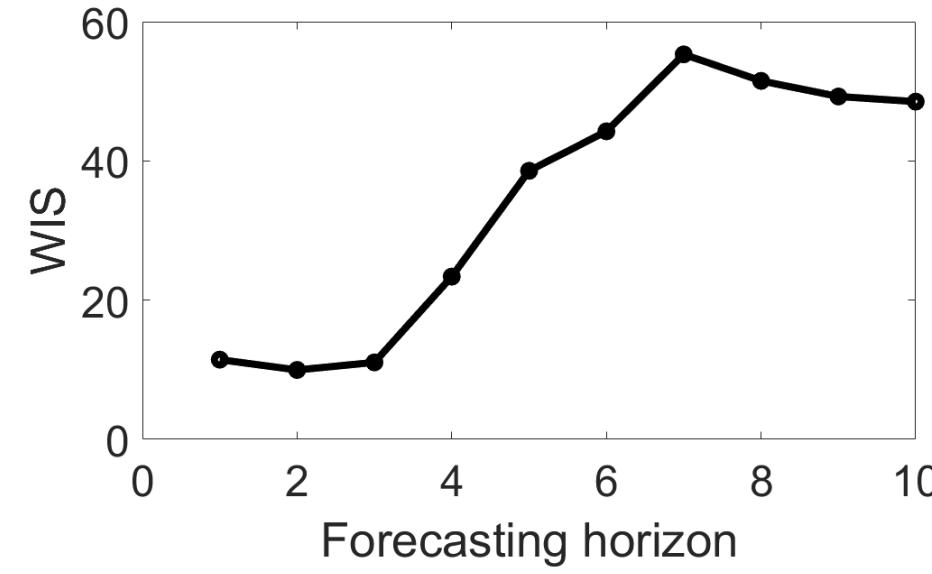
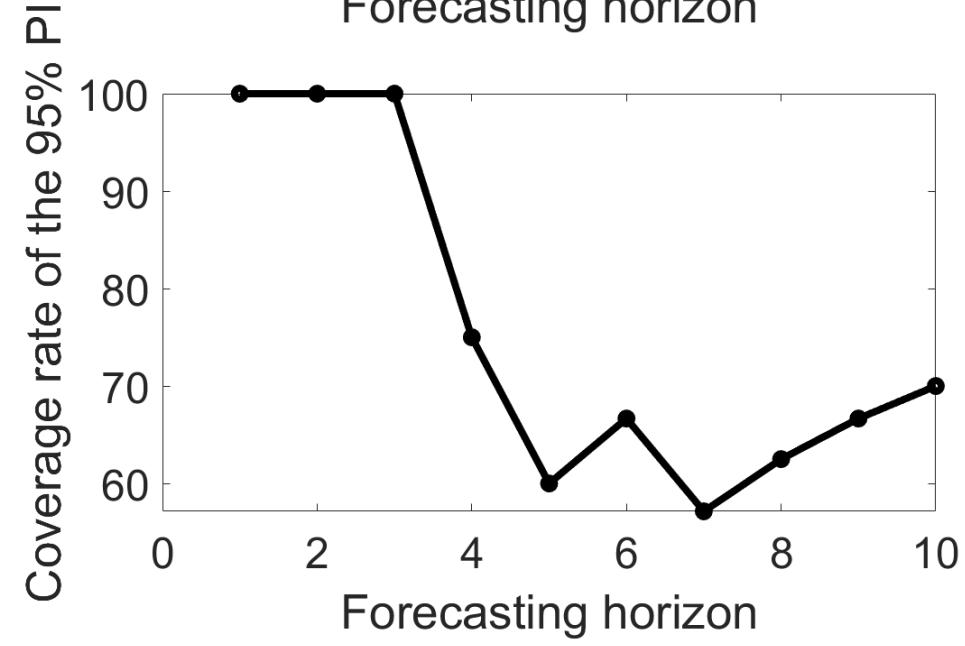
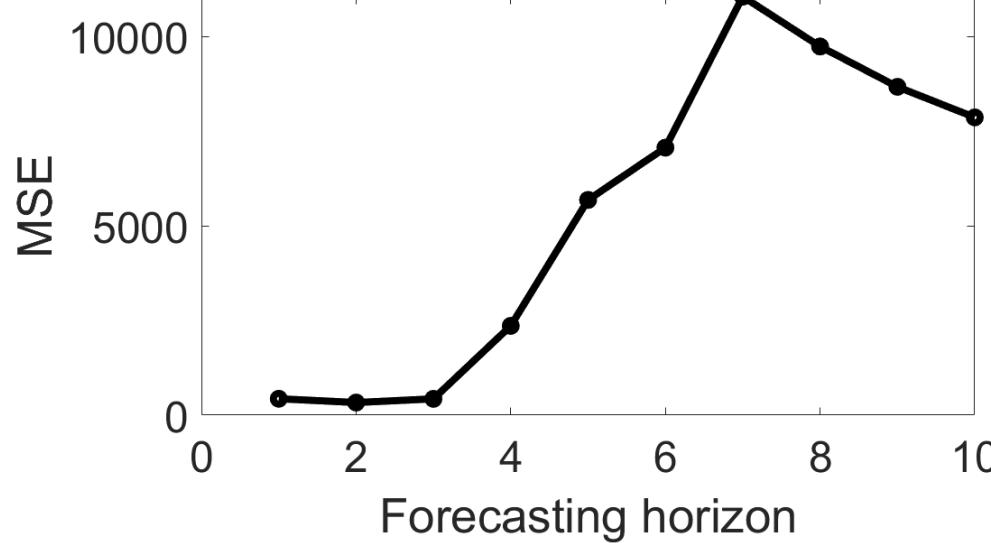
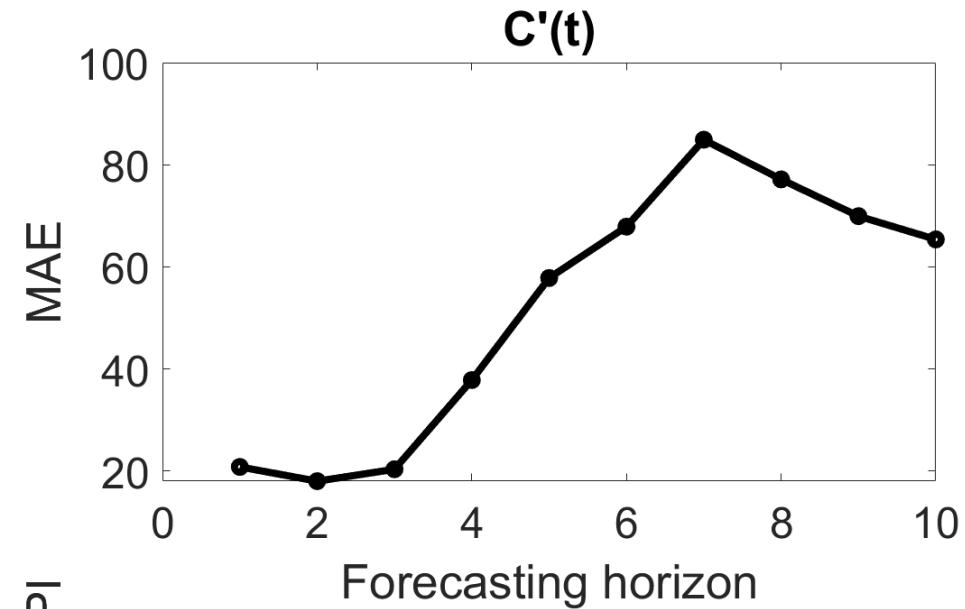
- `tstart1`, `tend1`, `windowSize1`, and `forecastingPeriod` correspond to the values entered in the rolling window analysis section of the `options_forecast.m` file.

Tutorial Code: (1) SEIR Negative Binomial

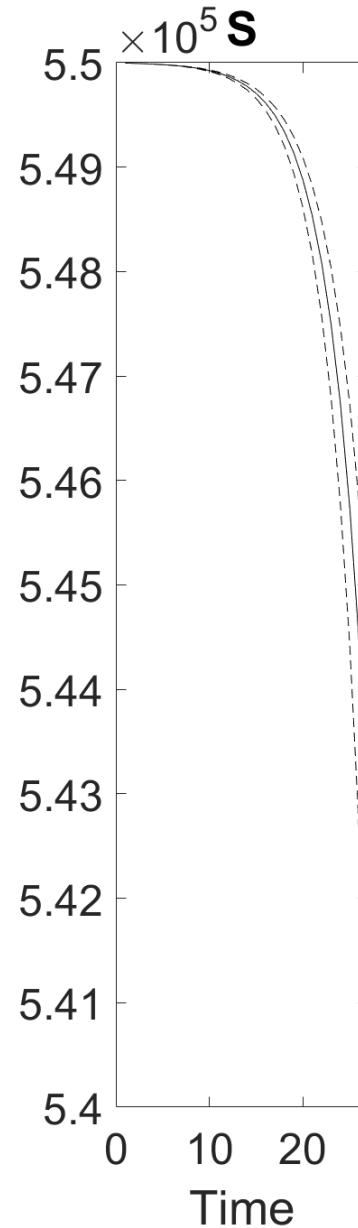
```
(1) plotForecast_ODEModel (@options_forecast_SEIR_f1918_dist1_3, 1, 1,  
17, 10)
```

SEIR model

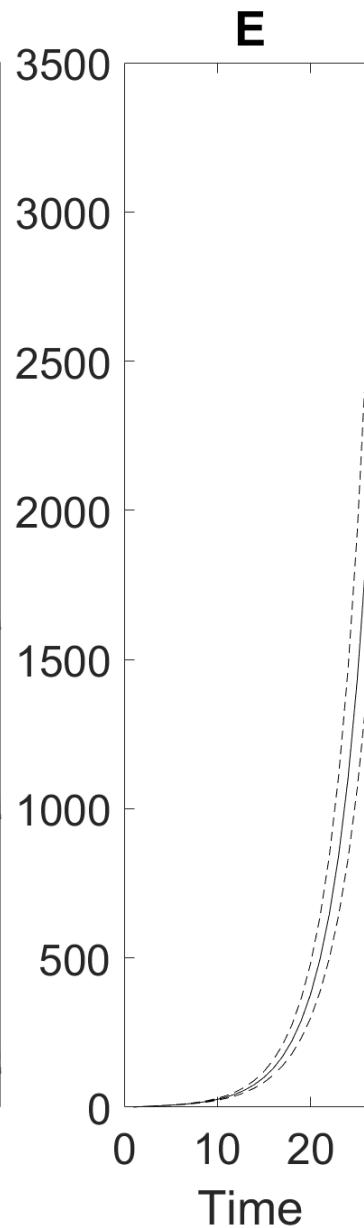




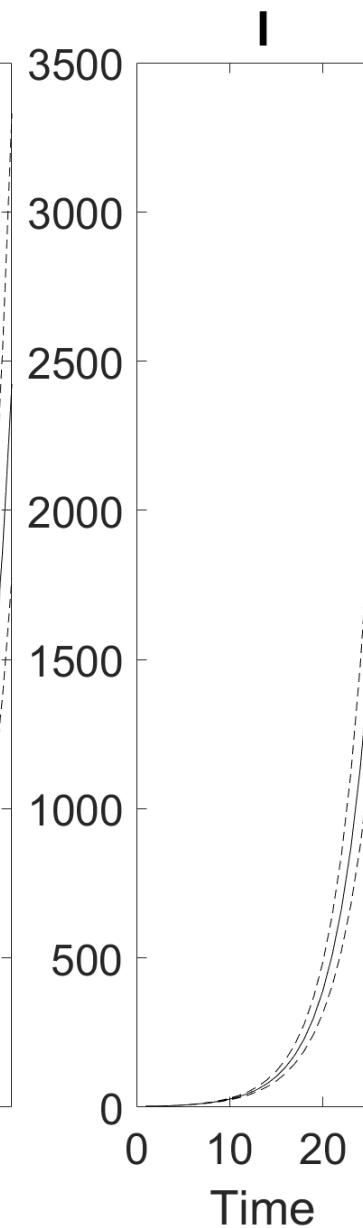
$\times 10^5$ S



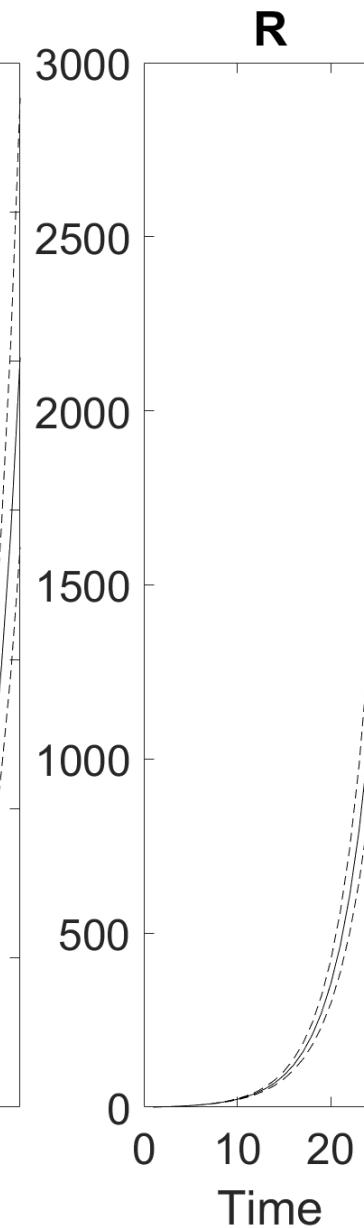
E



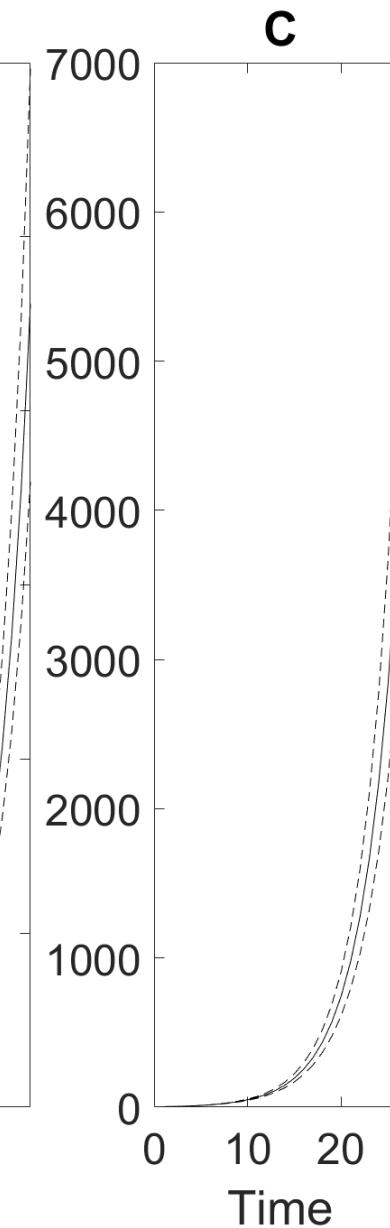
I



R



C



Model Comparison

SEIR Neg. Binomial & Exponential Neg. Binomial

Specifying the Exponential Model

```
% <=====
% < Author: Gerardo Chowell =====
% <=====

function dx=EXP(t,x,params0)

% parameters in order: r
dx=zeros(1,1);

dx(1,1)=params0(1)*x(1,1);
```

$$C'(t) = r * C(t)$$

r : Growth rate ($r > 0$)

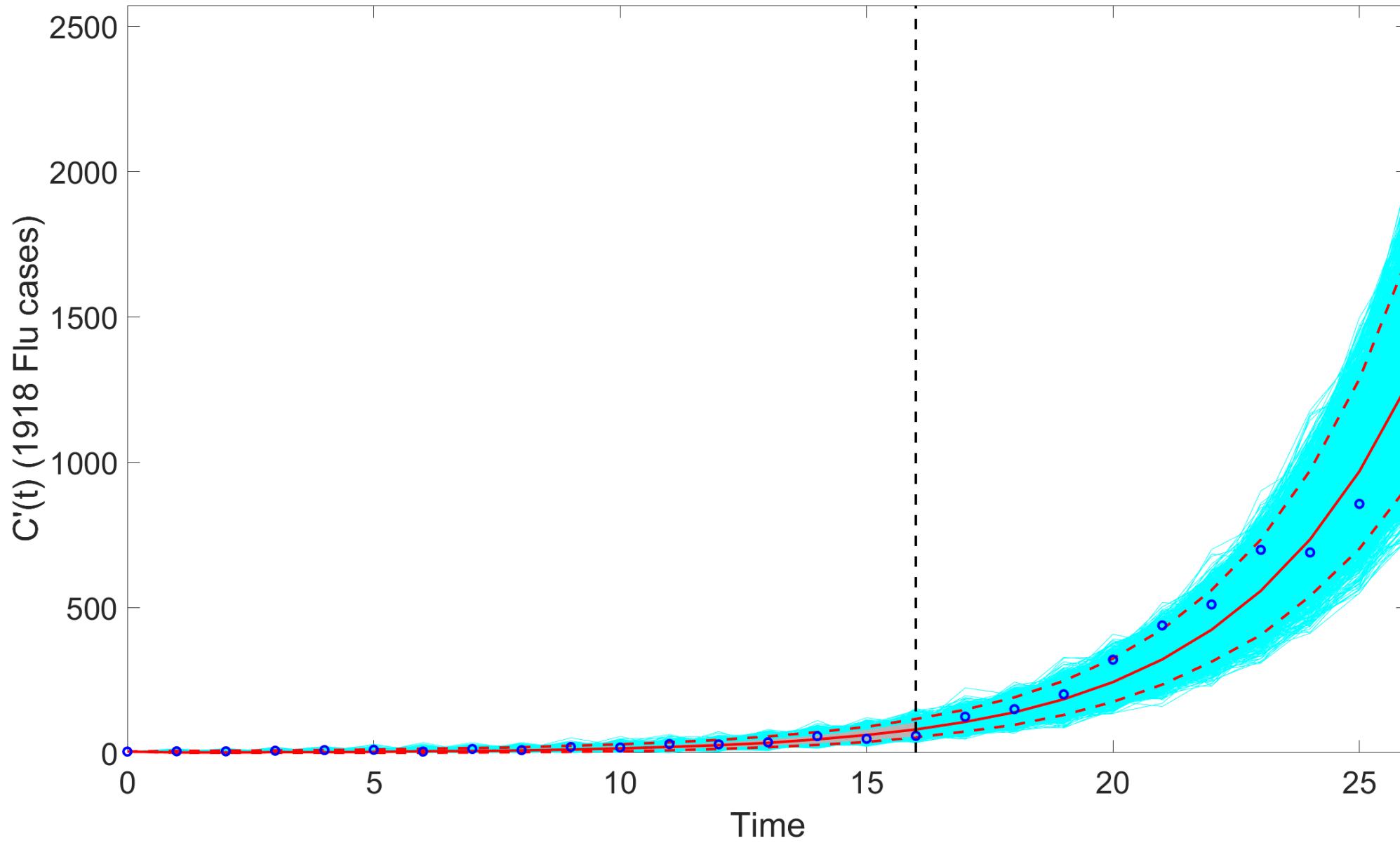
```
% <===== ODE model =====>
% <===== ODE model =====>
% <===== ODE model =====>

model.fc=@EXP; % name of the model function
model.name='EXP model'; % string indicating the name of the ODE model

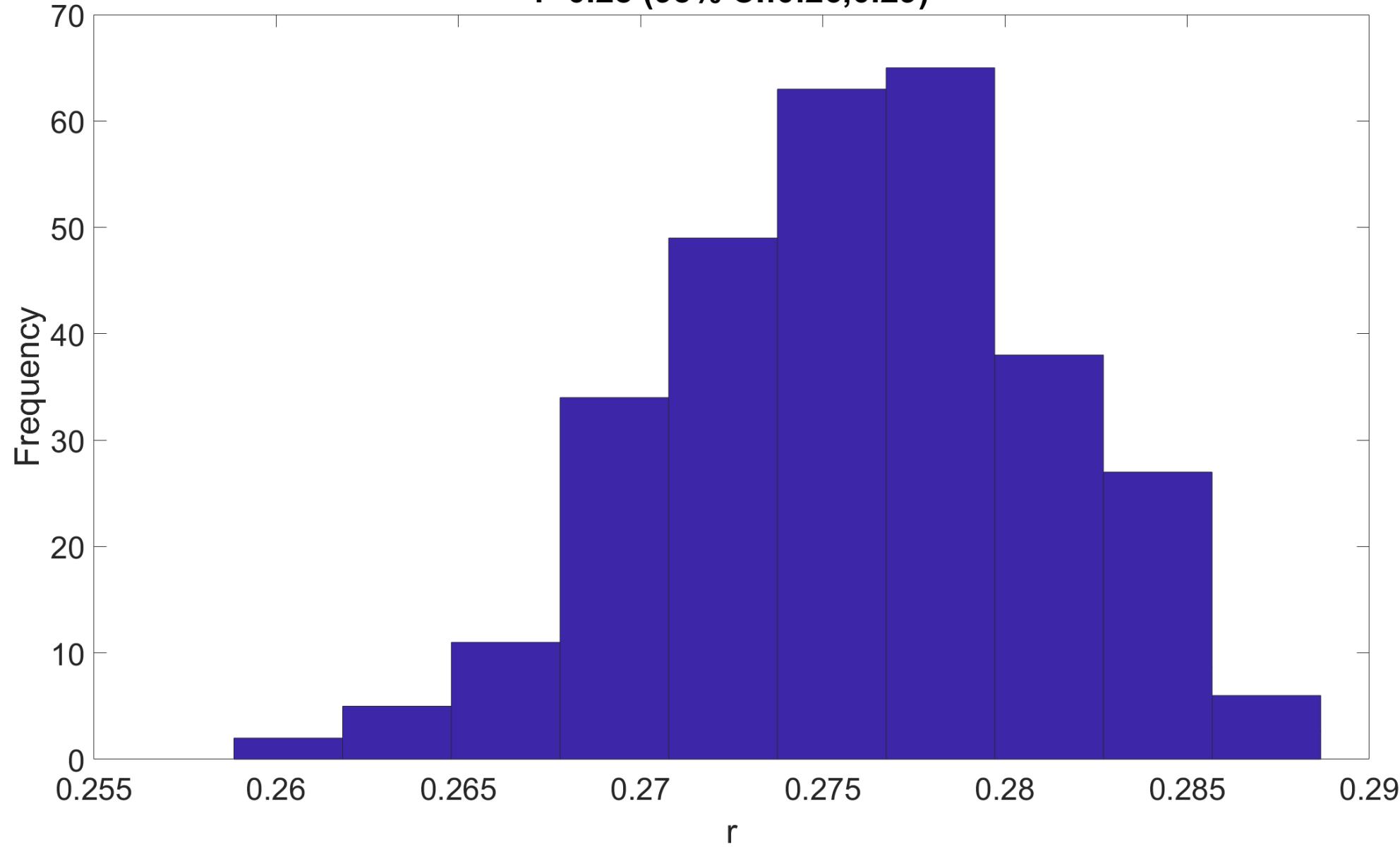
params.label={'r'}; % list of symbols to refer to the model parameters
params.LB=[0]; % lower bound values of the parameter estimates
params.UB=[10]; % upper bound values of the parameter estimates
params.initial=[0.18]; % initial parameter values/guesses
params.fixed=[0]; % Boolean vector to indicate any parameters that should remain fixed (1) to initial values indicated in par
params.fixI0=1; % Boolean variable indicating if the initial value of the fitting variable is fixed according to the first ob
params.composite=''; % Estimate a composite function of the individual model parameter estimates otherwise it is left empty.
params.composite_name=''; % Name of the composite parameter
params.extra0=[]; % used to pass any extra parameters (e.g., data, static variables) to the model function

vars.label={'C'}; % list of symbols to refer to the variables included in the model
vars.initial=5; % vector of initial conditions for the model variables
vars.fit_index=1; % index of the model's variable that will be fit to the observed time series data
vars.fit_diff=1; % boolean variable to indicate if the derivative of model's fitting variable should be fit to data.
```

EXP model

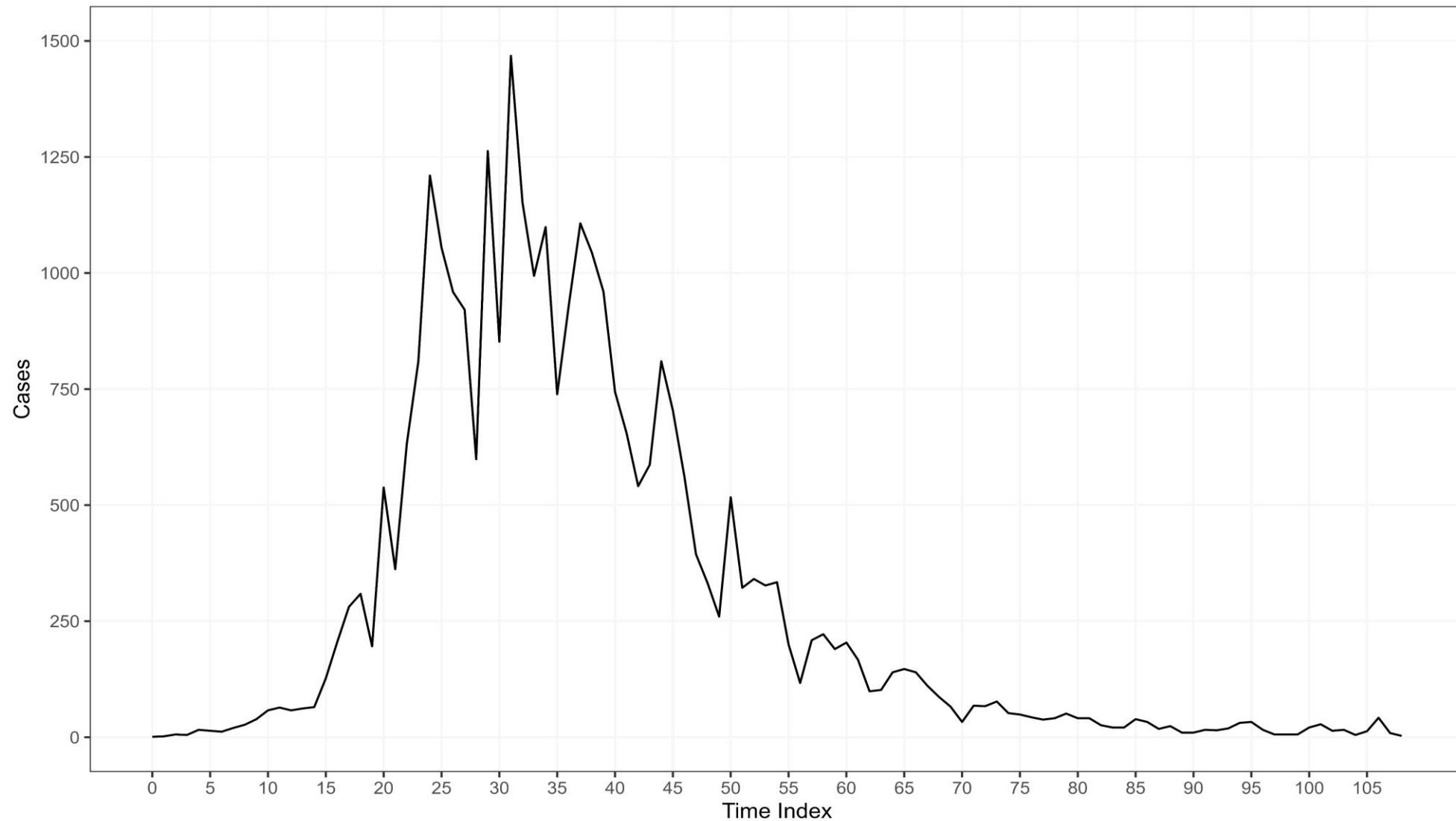


r=0.28 (95% CI:0.26,0.29)



Model	MAE	MSE	Coverage 95% PI	WIS
Calibration Performance				
SEIR model with NB error structure (<dist1>=3)	5.74	59.51	94.12	3.66
Exponential model with negative binomial error structure (<dist1>=3)	5.99	67.03	100.0	3.84
Forecast Performance				
SEIR model with NB error structure (<dist1>=3)	65.48	8058.65	70.0	49.80
Exponential model with negative binomial error structure (<dist1>=3)	79.79	9252.18	90.0	46.72

Tutorial #2: Swiss COVID-19



Daily number of reported cases of SARS-CoV-2 infection in Switzerland reported at the national level during the first wave (Feb. 2020 to June 2020).

Specifying SEIURC and R₀

Early Phase

```

% <=====
% < Author: Gerardo Chowell =====>
% <=====>

function dx=SEIRUnder(t,x,params0,extra0)

beta0=params0(1);
alpha=params0(2); % non-homogenous mixing parameter
rho=params0(3);
k=params0(4);
gamma1=params0(5);
N=params0(6);

dx=zeros(6,1); % define the vector of the state derivatives: S, E, I, U, R, C

dx(1,1)= -beta0*x(1,1).*((x(3,1)+x(4,1)).^alpha)./N; %S

dx(2,1)= beta0*x(1,1).*((x(3,1)+x(4,1)).^alpha)./N - k*x(2,1); %E

dx(3,1)= k*rho*x(2,1) - gamma1*x(3,1); %I

dx(4,1)= k*(1-rho)*x(2,1) - gamma1*x(4,1); %U

dx(5,1)= gamma1*(x(3,1)+x(4,1)); %R

dx(6,1)= k*rho*x(2,1); %C

```

$$\begin{cases} \dot{S} = -\beta S(t) \frac{I(t) + U(t)^\alpha}{N} \\ \dot{E} = \beta S(t) \frac{I(t) + U(t)^\alpha}{N} - \kappa E(t) \\ \dot{I} = \kappa \rho E(t) - \gamma I(t) \\ \dot{U} = \kappa (1 - \rho) E(t) - \gamma U(t) \\ \dot{R} = \gamma (I(t) + U(t)) \\ \dot{C} = \kappa \rho E(t) \end{cases}$$

```
% <===== ODE model =====>
% <===== ODE model =====>
% <===== ODE model =====>

model.fc=@SEIR_unreported; % name of the model function
model.name='SEIR swiss_covid underreporting'; % string indicating the name of the ODE model

params.label={'\beta_0','\alpha','\rho','\kappa','\gamma','N'}; % list of symbols to refer to the model parameters
params.LB=[0.001 0.5 0.01 0.01 0.01 47332614]; % lower bound values of the parameter estimates
params.UB=[5 1 1 1 1 47332614]; % upper bound values of the parameter estimates
params.initial=[0.6 1 1 1/5 1/4 47332614]; % initial parameter values/guesses
params.fixed=[0 1 0 1 1 1]; % Boolean vector to indicate any parameters that should remain fixed (1) to initial values in
params.fixI0=1; % Boolean variable indicating if the initial value of the fitting variable is fixed according to the first
params.composite=@R0s_unreported; % Estimate a composite function of the individual model parameter estimates otherwise
params.composite_name='R0s_unreported'; % Name of the composite parameter
params.extra0='';

vars.label={'S','E','I','U','R','C'}; % list of symbols to refer to the variables included in the model
vars.initial=[params.initial(6)-1 0 1 0 0 1]; % vector of initial conditions for the model variables
vars.fit_index=6; % index of the model's variable that will be fit to the observed time series data
vars.fit_diff=1; % boolean variable to indicate if the derivative of model's fitting variable should be fit to data.
```

Preparing options_fit.m & options_forecast.m

Fitting and forecasting the early growth phase (First 20-days)

```
% <===== Declare global variables =====>
% <===== Declare global variables =====>
% <===== Declare global variables =====>

global method1 % Parameter estimation method

% <===== Datasets properties =====>
% <===== Datasets properties =====>
% <===== Datasets properties =====>
% Located in the input folder, the time series data file is a text file with extension *.txt.
% The time series data file contains the incidence curve of the epidemic of interest.
% The first column corresponds to time index: 0,1,2, ... and the second
% column corresponds to the observed time series data.

cadfilename1='curve-covid-firstwave-swiss-reporteddt';
caddisease='COVID-19'; % string indicating the name of the disease related to the time series data
datatype='cases'; % string indicating the nature of the data (cases, deaths, hospitalizations, etc)
```

```

% <===== Parameter estimation =====>
% <===== Parameter estimation =====>
% <===== Parameter estimation =====>

method1=0; % Type of estimation method

% Nonlinear least squares (LSQ)=0,
% MLE Poisson=1,
% MLE (Neg Binomial)=3, with VAR=mean+alpha*mean;
% MLE (Neg Binomial)=4, with VAR=mean+alpha*mean^2;
% MLE (Neg Binomial)=5, with VAR=mean+alpha*mean^d;

dist1=0; % Define dist1 which is the type of error structure. See below:

%dist1=0; % Normal distribution to model error structure (method1=0)
%dist1=1; % Poisson error structure (method1=0 OR method1=1)
%dist1=2; % Neg. binomial error structure where var = factor1*mean where
    % factor1 is empirically estimated from the time series
    % data (method1=0)
%dist1=3; % MLE (Neg Binomial) with VAR=mean+alpha*mean (method1=3)
%dist1=4; % MLE (Neg Binomial) with VAR=mean+alpha*mean^2 (method1=4)
%dist1=5; % MLE (Neg Binomial)with VAR=mean+alpha*mean^d (method1=5)

switch method1
    case 1
        dist1=1;
    case 3
        dist1=3;
    case 4
        dist1=4;
    case 5
        dist1=5;
end

numstartpoints=2*4^2; % Number of initial guesses for optimization procedure using MultiStart
B=300; % number of bootstrap realizations to characterize parameter uncertainty

```

Tutorial

- As the case counts are much higher compared to Tutorial #1, we will be using nonlinear least squares (NLSQ) and assuming a normal error distribution.
- Using 32 start points ($\langle \text{numstartpoints} \rangle = 32$)
- Using 300 bootstrap realizations to characterize parameter uncertainty ($\langle B \rangle = 300$)

Fitting to the first 20 days

```
% <===== Forecasting parameters =====>
% <===== Parameters of the rolling window analysis =====>
% <=====>

getperformance=1; % flag or indicator variable (1/0) to calculate forecasting performance or not

forecastingperiod=10; % forecast horizon (number of time units ahead)

windowsize1=20; % moving window size

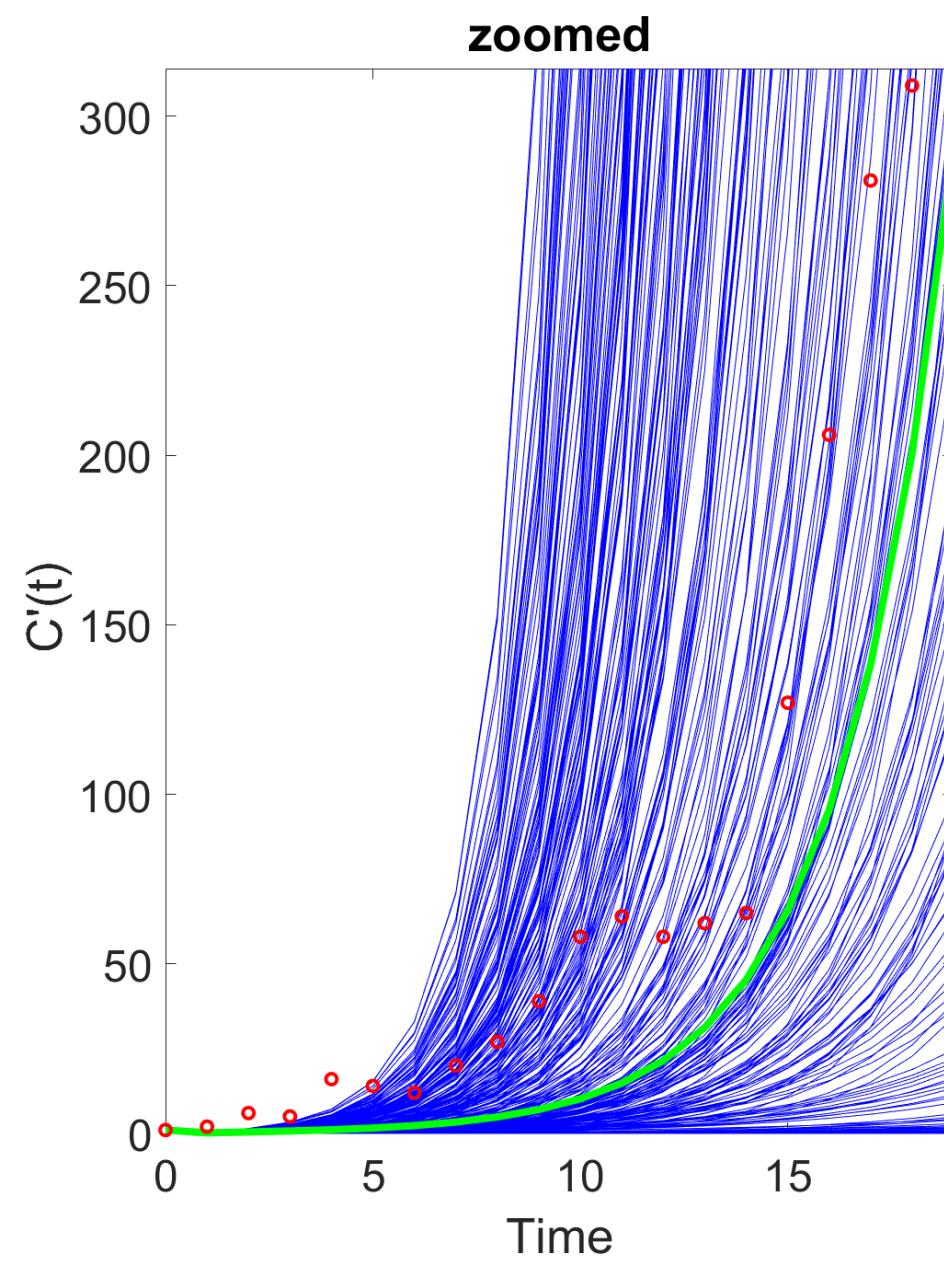
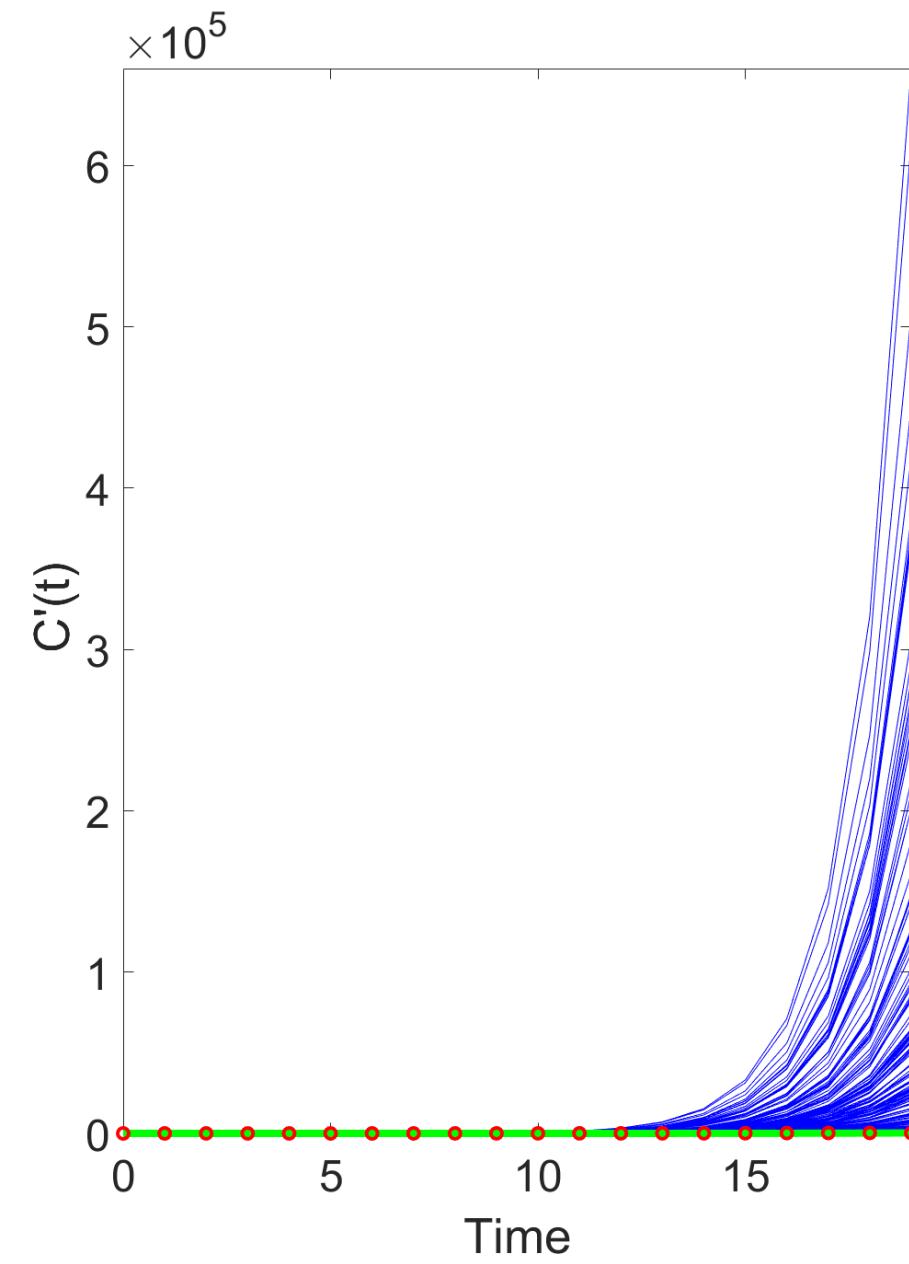
tstart1=1; % time point for the start of rolling window analysis

tend1=1; %time point for the end of the rolling window analysis

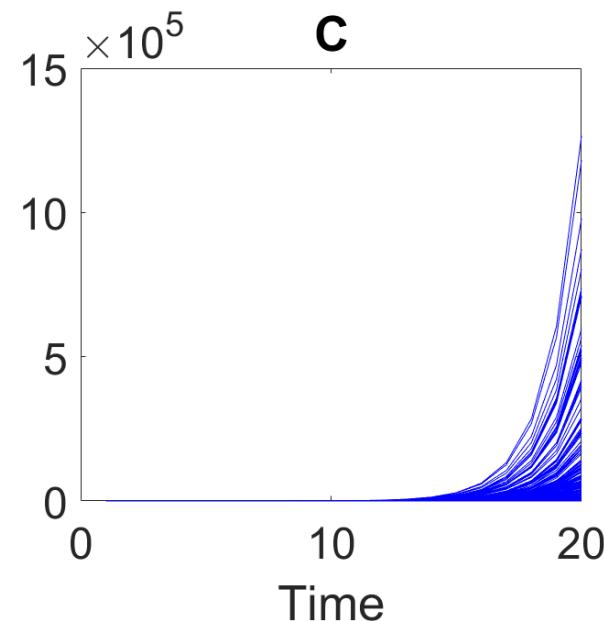
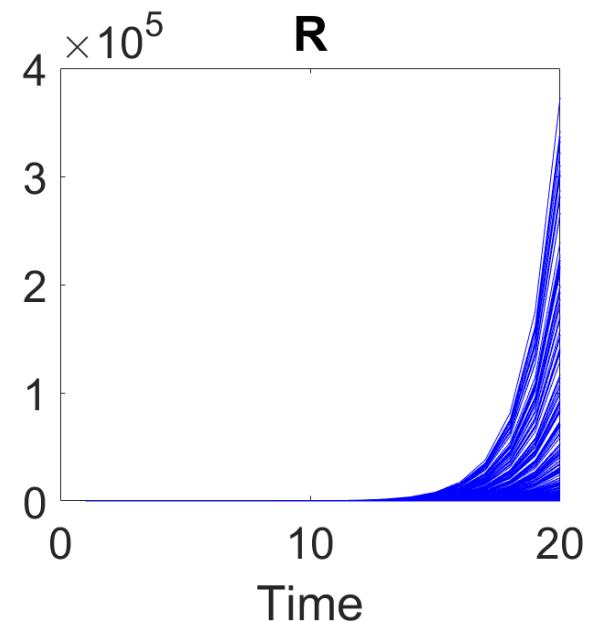
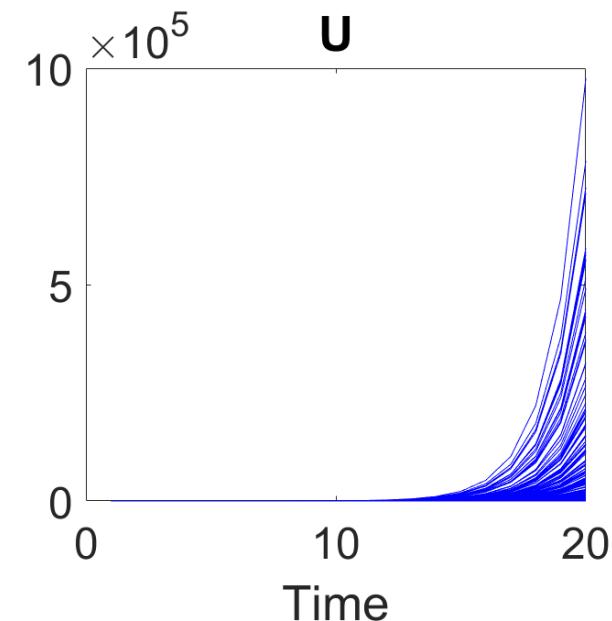
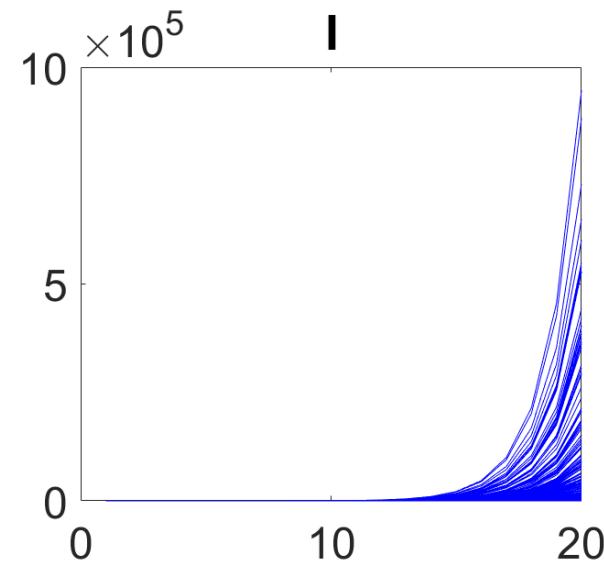
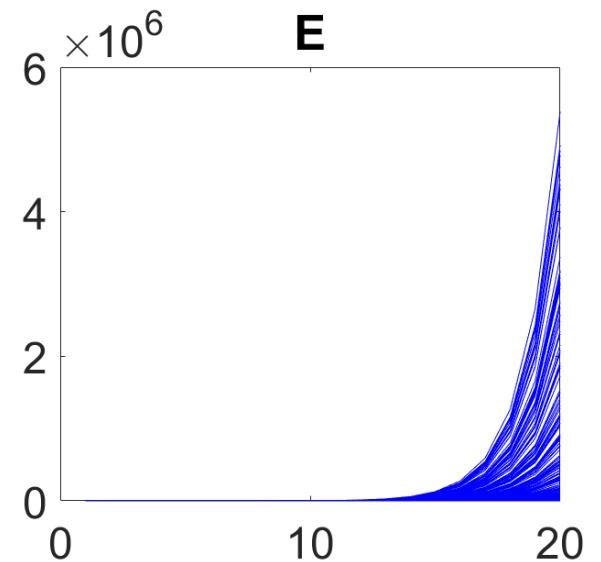
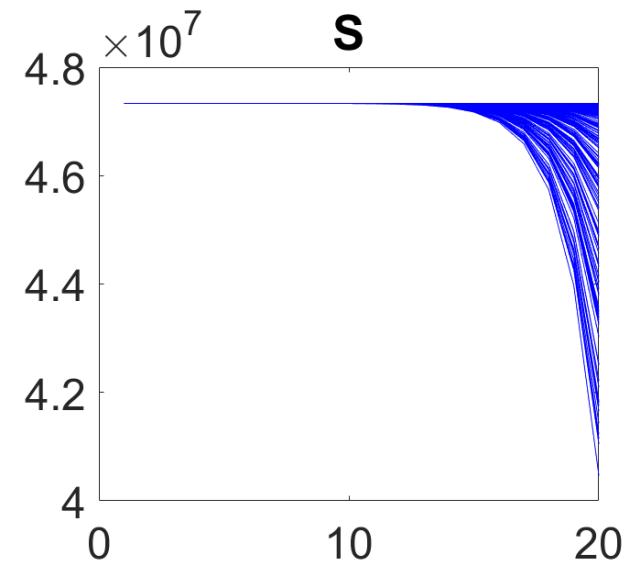
printscreen1=1;
```

Generating preliminary model solutions

```
plotODEModel(@options_fit_SEIR_unreportedNIn_covid_s  
wiss_dist1_0)
```



Fixed α

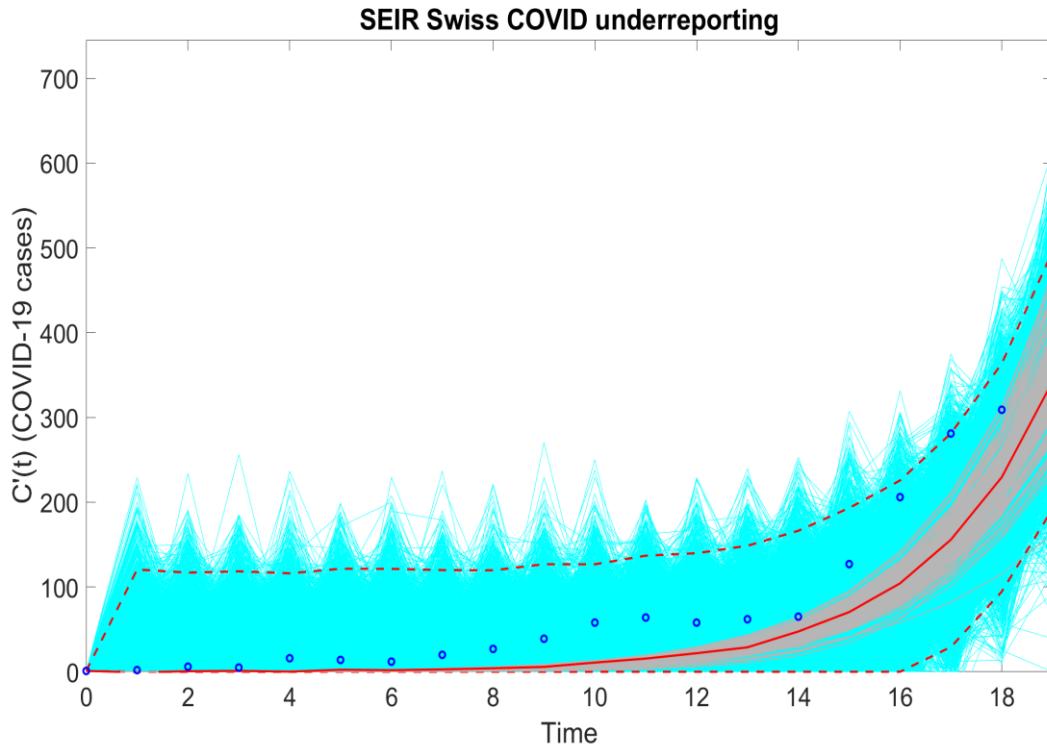


Fixed α

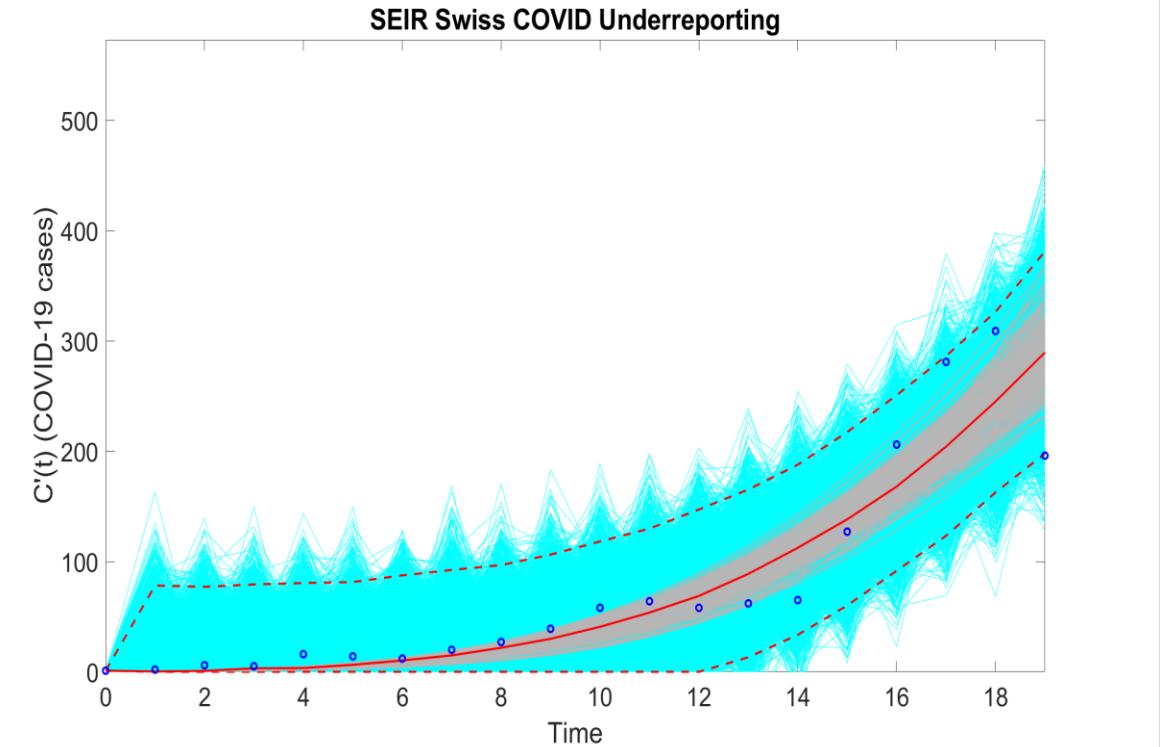
Fitting, plotting, and evaluating the model with quantified uncertainty

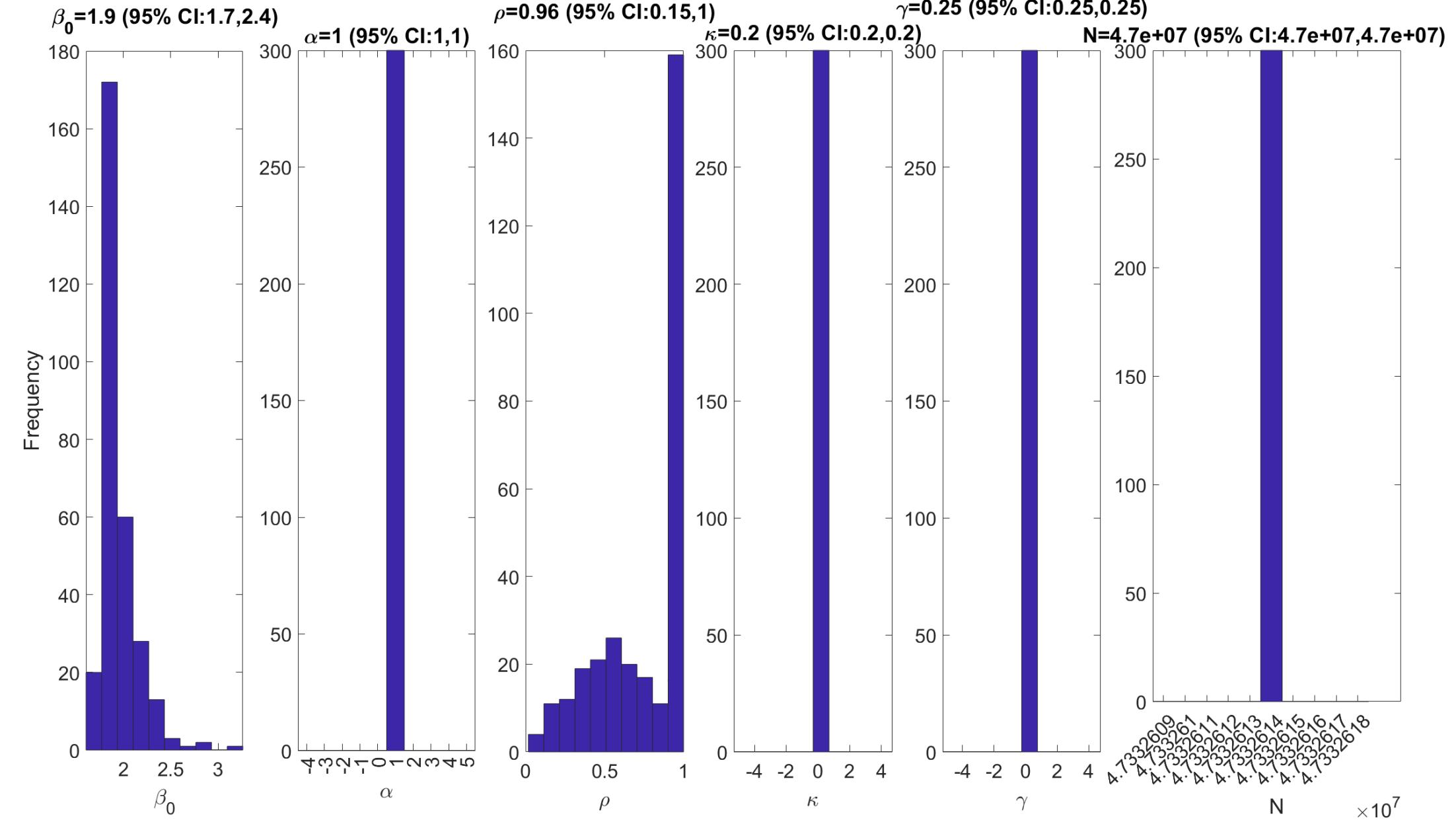
```
(1) Run_Fit_ODEModel(@options_fit_SEIR_unreportedNIn  
                      _covid_swiss_dist1_0, 1, 1, 20)  
(2) plotFit_ODEModel(@options_fit_SEIR_unreportedNIn  
                      _covid_swiss_dist1_0, 1, 1, 20)
```

Fixed α

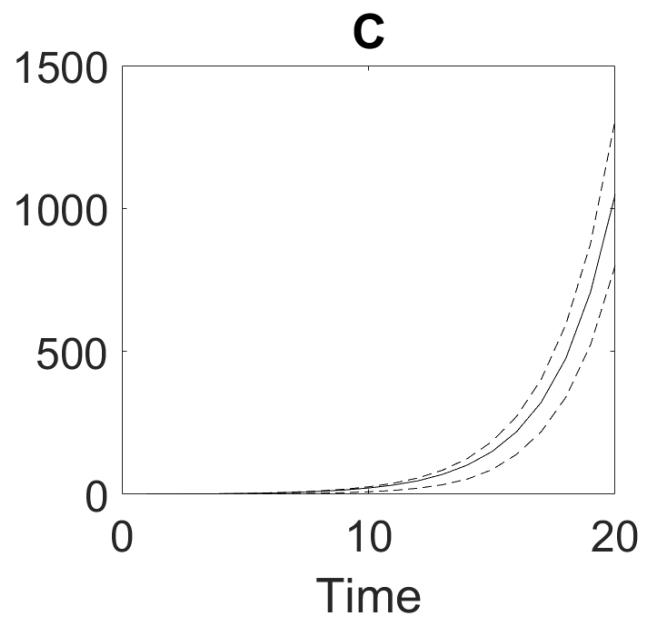
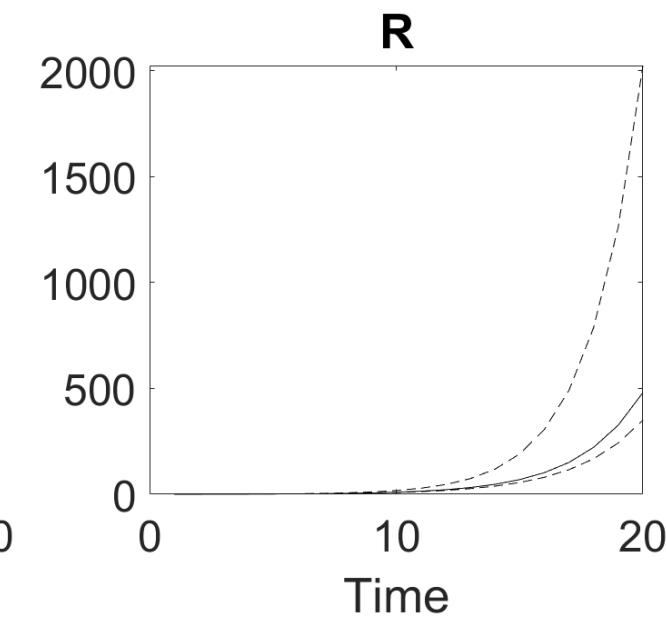
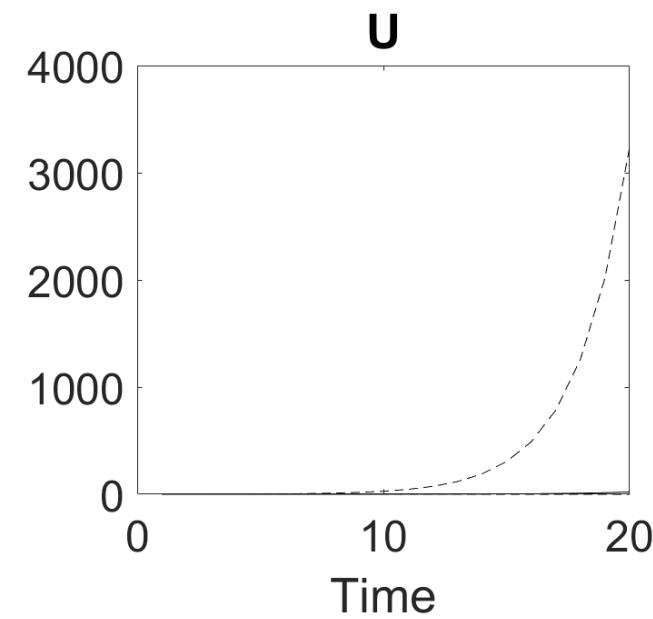
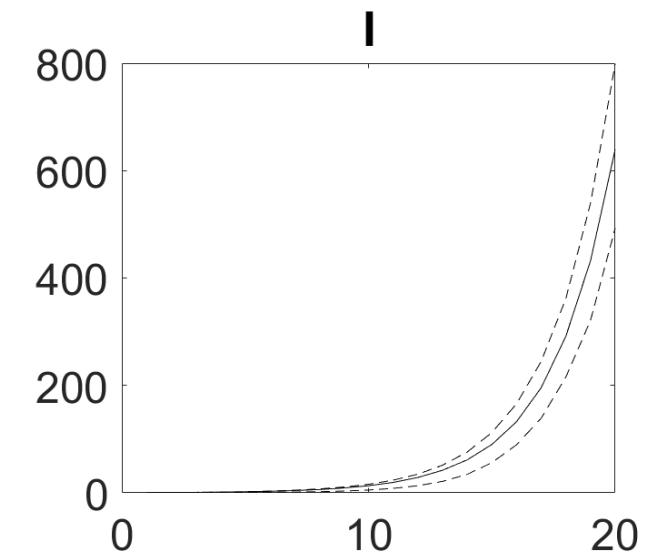
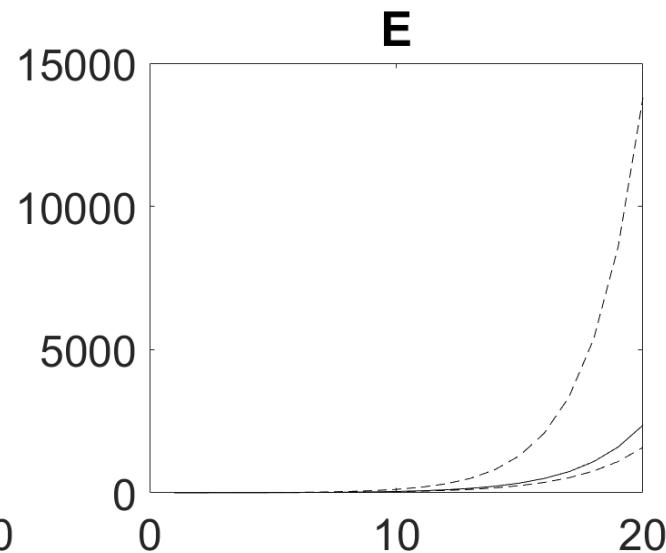
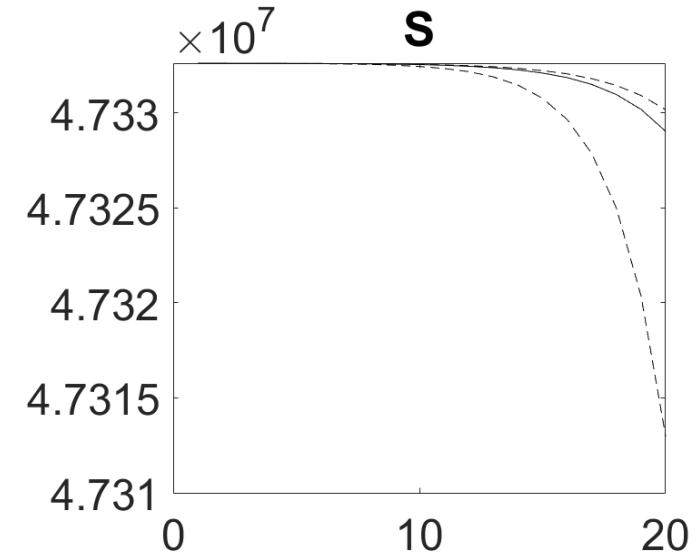


Estimated α



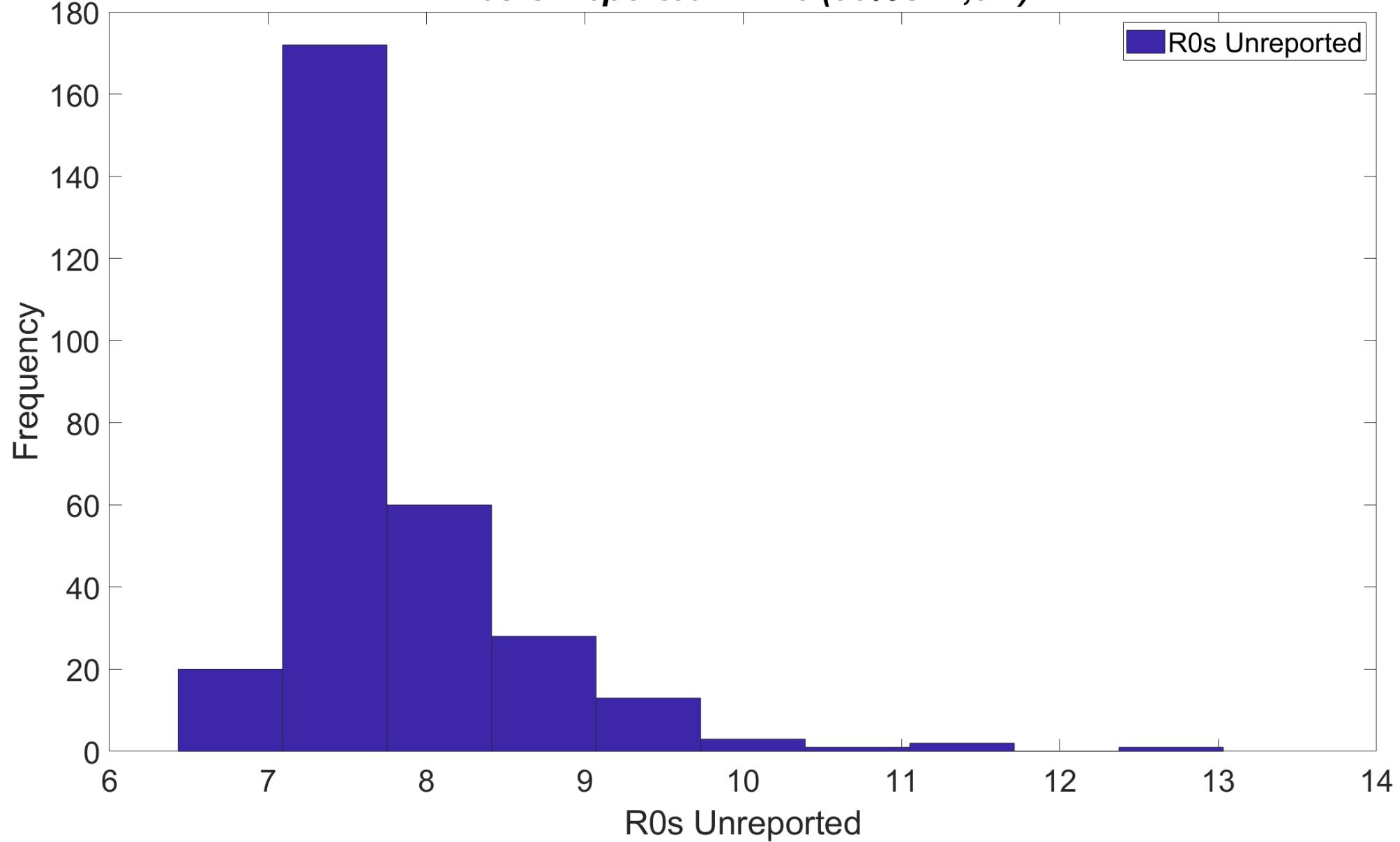


Fixed α



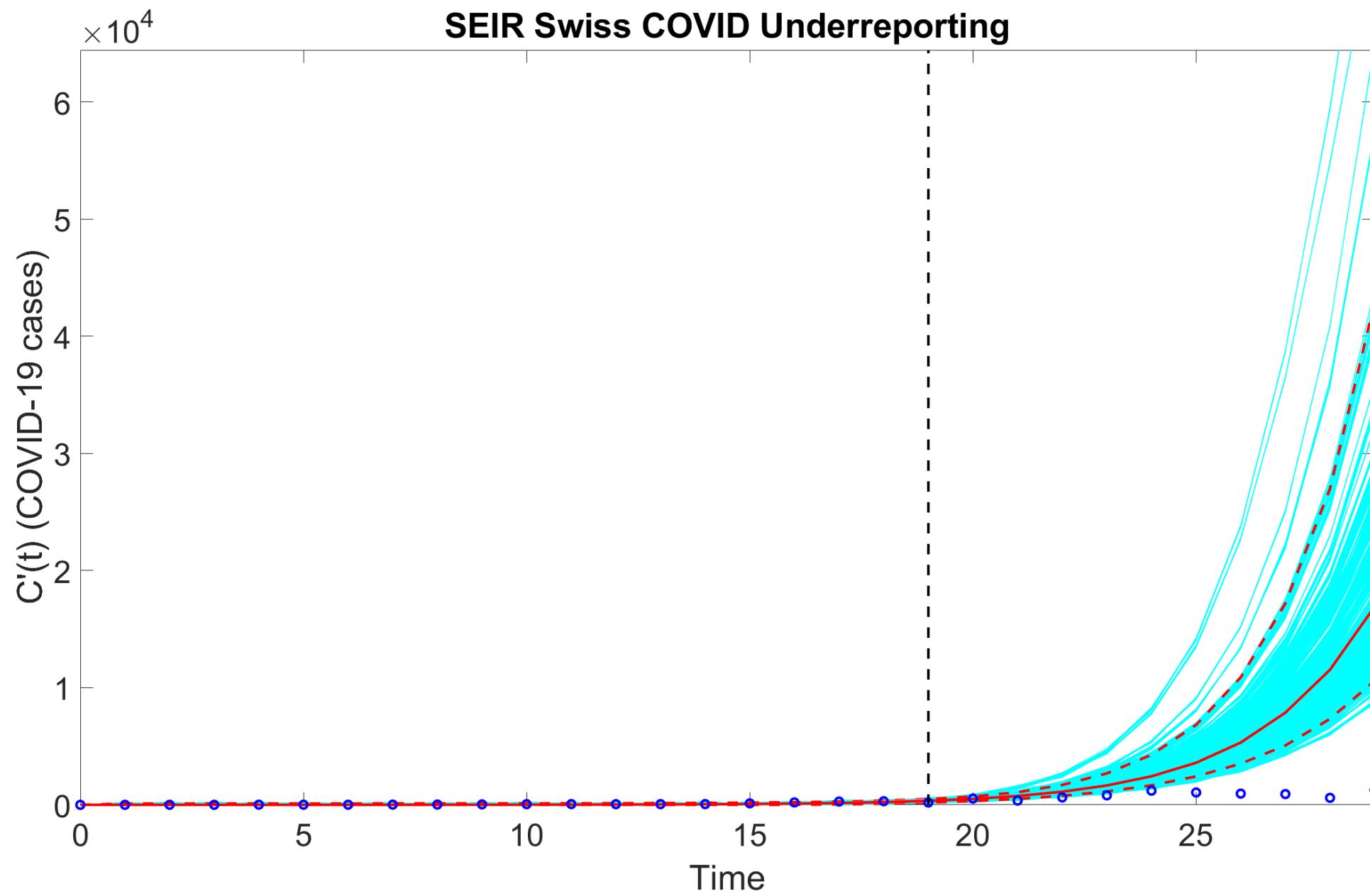
Fixed α

R0s Unreported = 7.46 (95%CI:7,9.7)

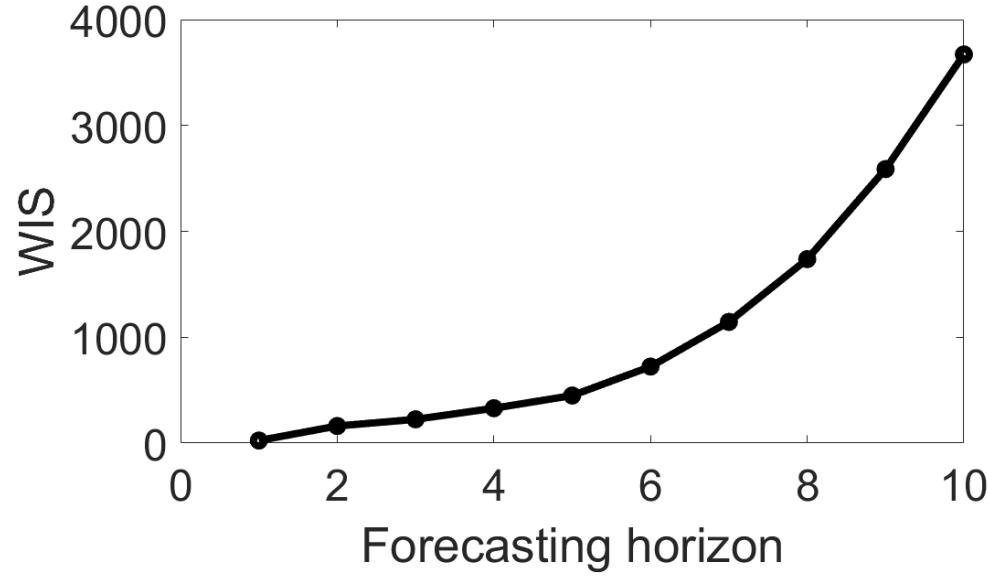
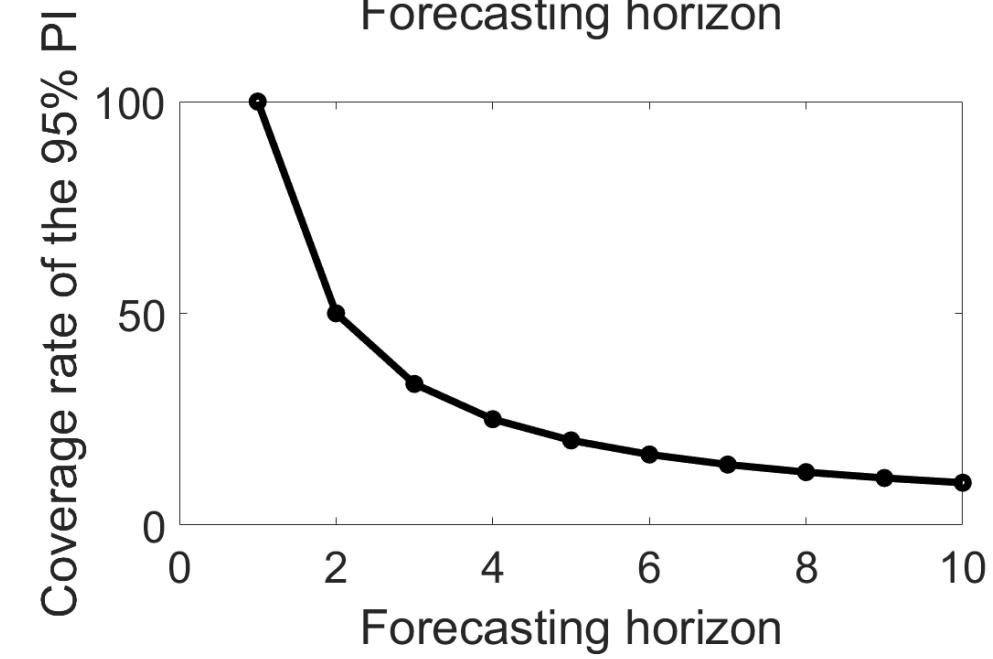
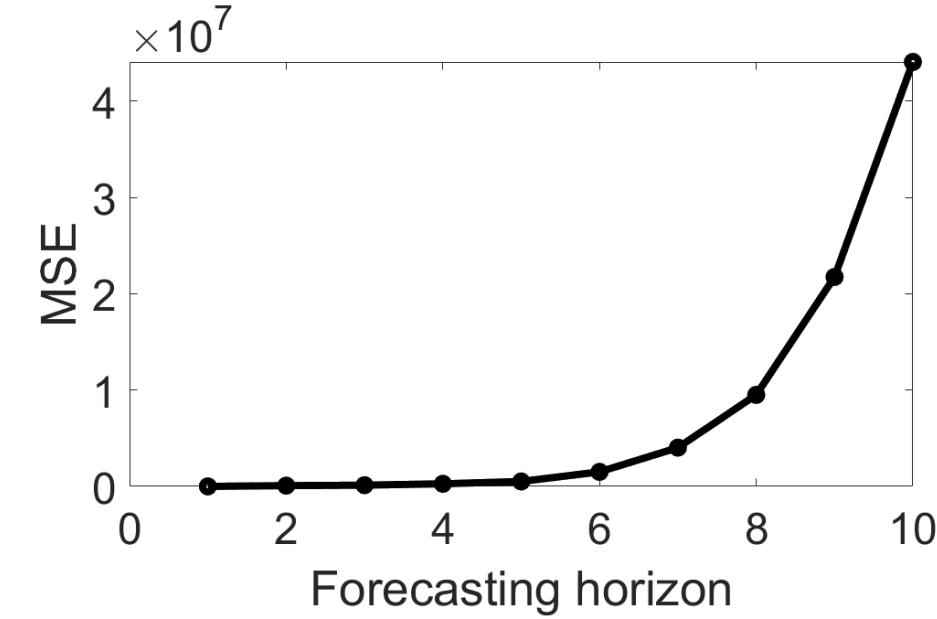
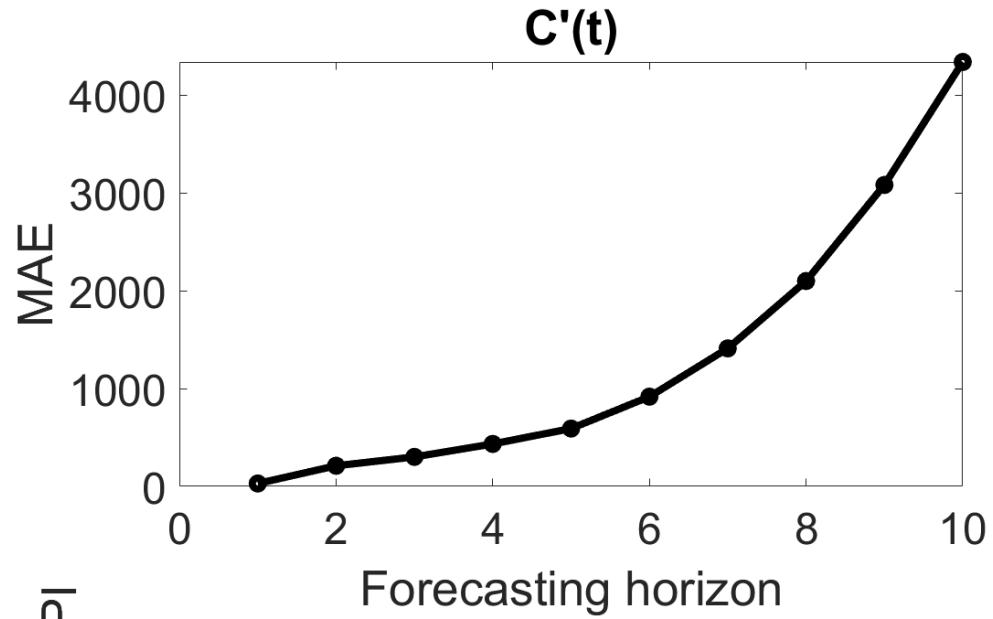


Generating, plotting and assessing model-based forecasts

```
(1) Run_Forecasting_ODEModel(@options_forecast_SEIR_
    unreportedNIn_covid_swiss_dist1_0, 1, 1, 20,10)
(2) plotForecast_ODEModel(@options_forecast_SEIR_unr
    eportedNIn_covid_swiss_dist1_0, 1, 1, 20,10)
```



Fixed α



Fixed α

Model Comparison (Early Phase)

SEIUR Normal, GGM Normal & RICH Normal

Specifying the GGM Model

```
% <=====
% < Author: Gerardo Chowell =====
% <=====

function dx=GGM(t,x,params0,extra0)

% parameters in order: r, p

dx=zeros(1,1);

dx(1,1)=params0(1)*x(1,1).^params0(2);
```

$$C'(t) = rC(t)^p$$

r : Growth rate ($r > 0$)
 p : Scaling of growth parameter

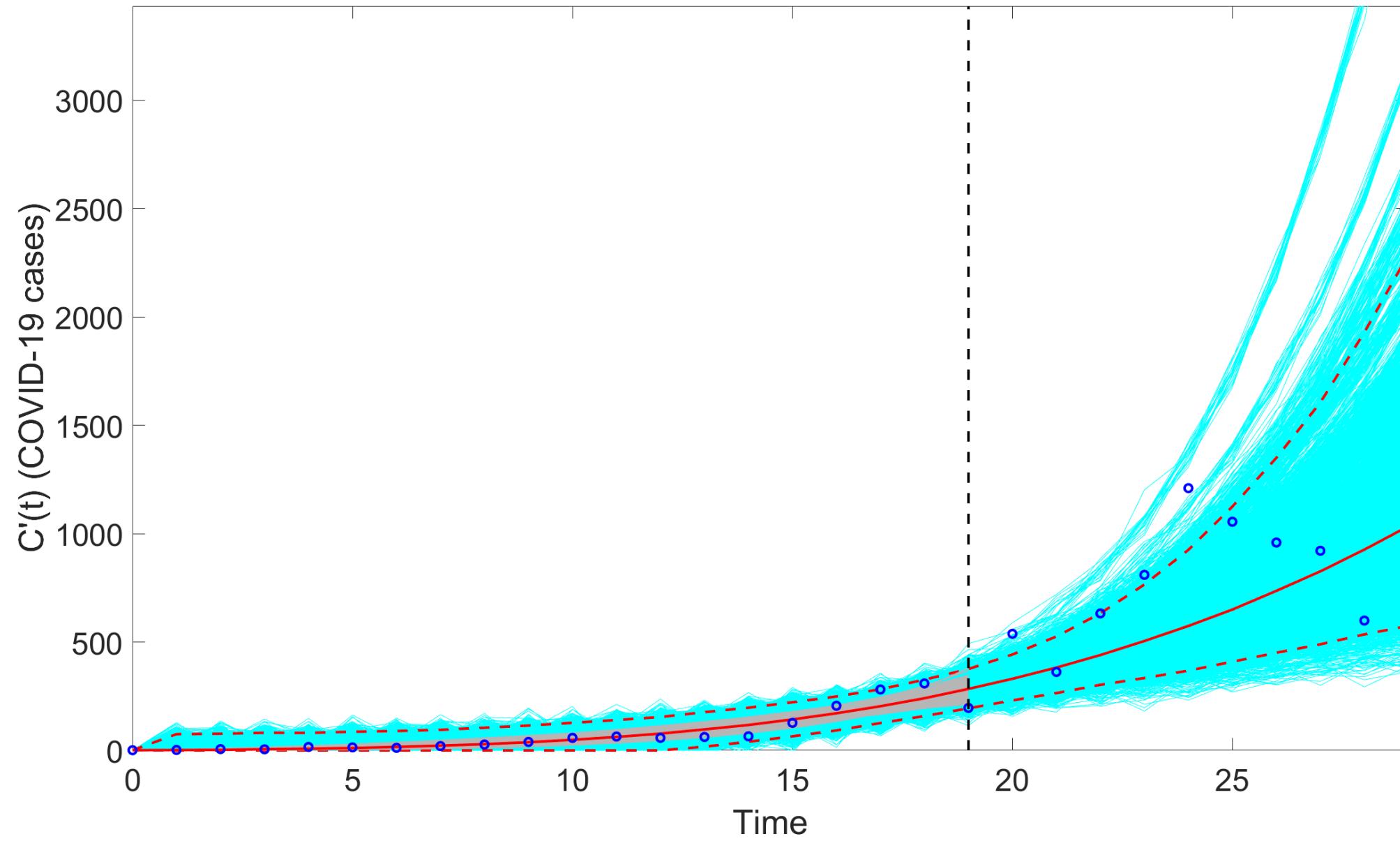
```
% <===== ODE model =====>
% <===== ODE model =====>
% <===== ODE model =====>

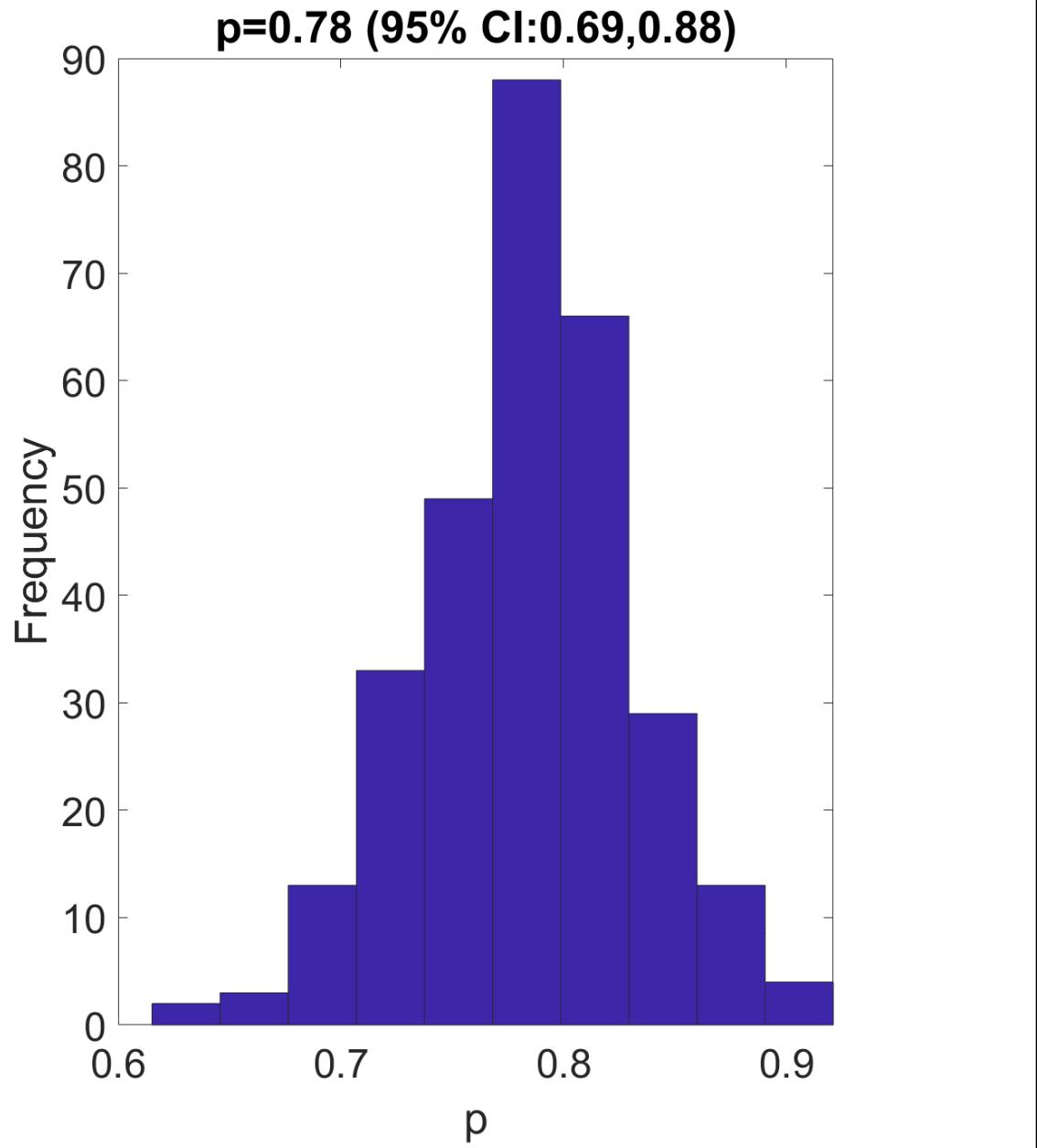
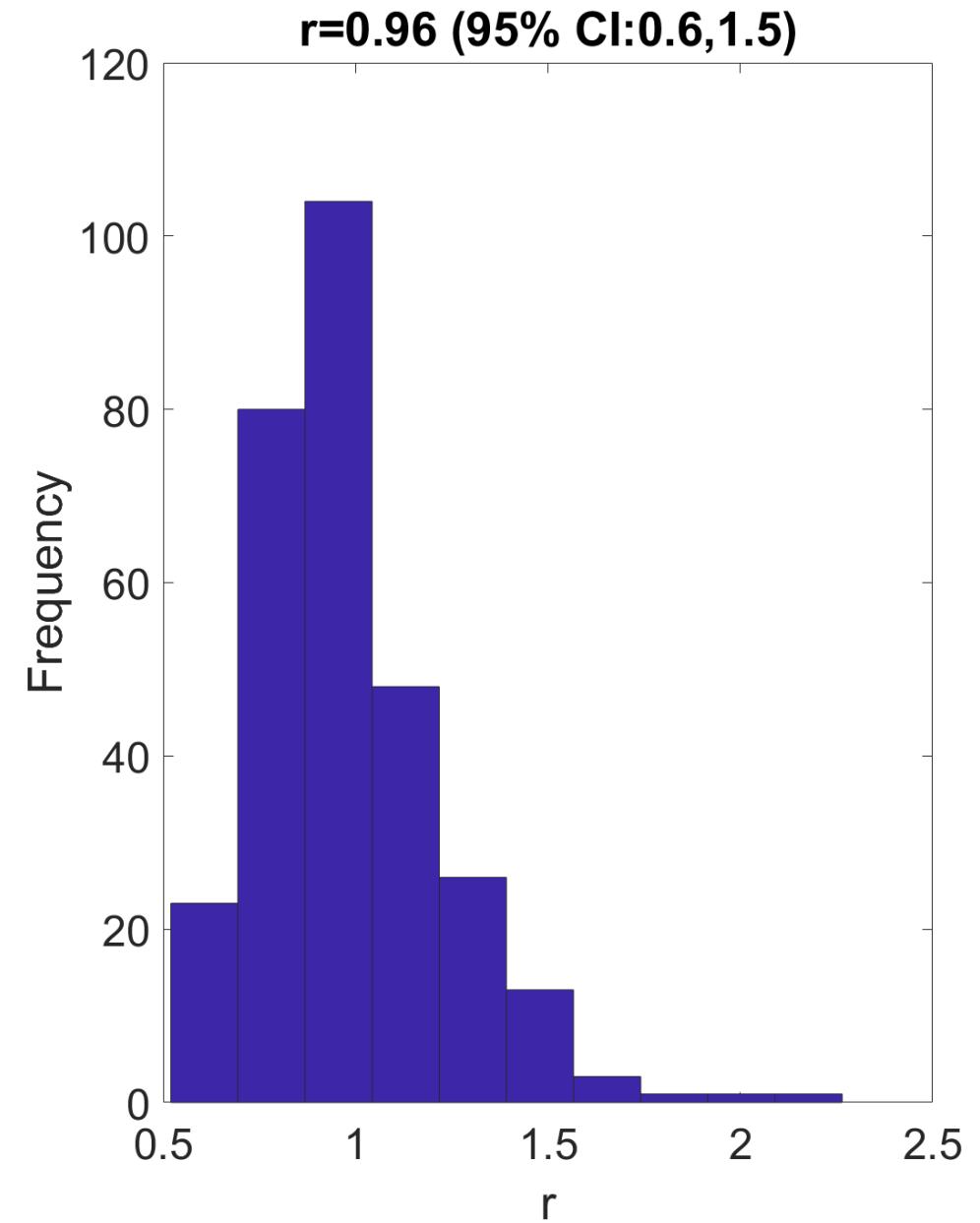
model.fc=@GGM; % name of the model function
model.name='GGM model'; % string indicating the name of the ODE model

params.label={'r', 'p'}; % list of symbols to refer to the model parameters
params.LB=[0 0]; % lower bound values of the parameter estimates
params.UB=[10 1]; % upper bound values of the parameter estimates
params.initial=[0.91 0.79]; % initial parameter values/guesses
params.fixed=[0 0]; % Boolean vector to indicate any parameters that should remain fixed (1) to initial values indicated above
params.fixI0=1; % Boolean variable indicating if the initial value of the fitting variable is fixed according to the fixed parameter above
params.composite=''; % Estimate a composite function of the individual model parameter estimates otherwise it is left empty
params.composite_name=''; % Name of the composite parameter
params.extra0=[]; % used to pass any extra parameters (e.g., data, static variables) to the model function

vars.label={'C'}; % list of symbols to refer to the variables included in the model
vars.initial=5; % vector of initial conditions for the model variables
vars.fit_index=1; % index of the model's variable that will be fit to the observed time series data
vars.fit_diff=1; % boolean variable to indicate if the derivative of model's fitting variable should be fit to data.
```

GGM model





Specifying the RICH Model

```
% <=====
% < Author: Gerardo Chowell =====
% <=====

function dx=RICH(t,x,params0,extra0)

% parameters in order: r, a, K0

dx=zeros(1,1);

dx(1,1)=params0(1)*x(1,1)*(1-(x(1,1)/params0(3)).^params0(2));
```

$$C'(t) = rC(t)\left[1 - \left(\frac{C(t)}{K_0}\right)^p\right]$$

r : Growth rate ($r > 0$)

p : Scaling of growth parameter

K_0 : Final cumulative epidemic size

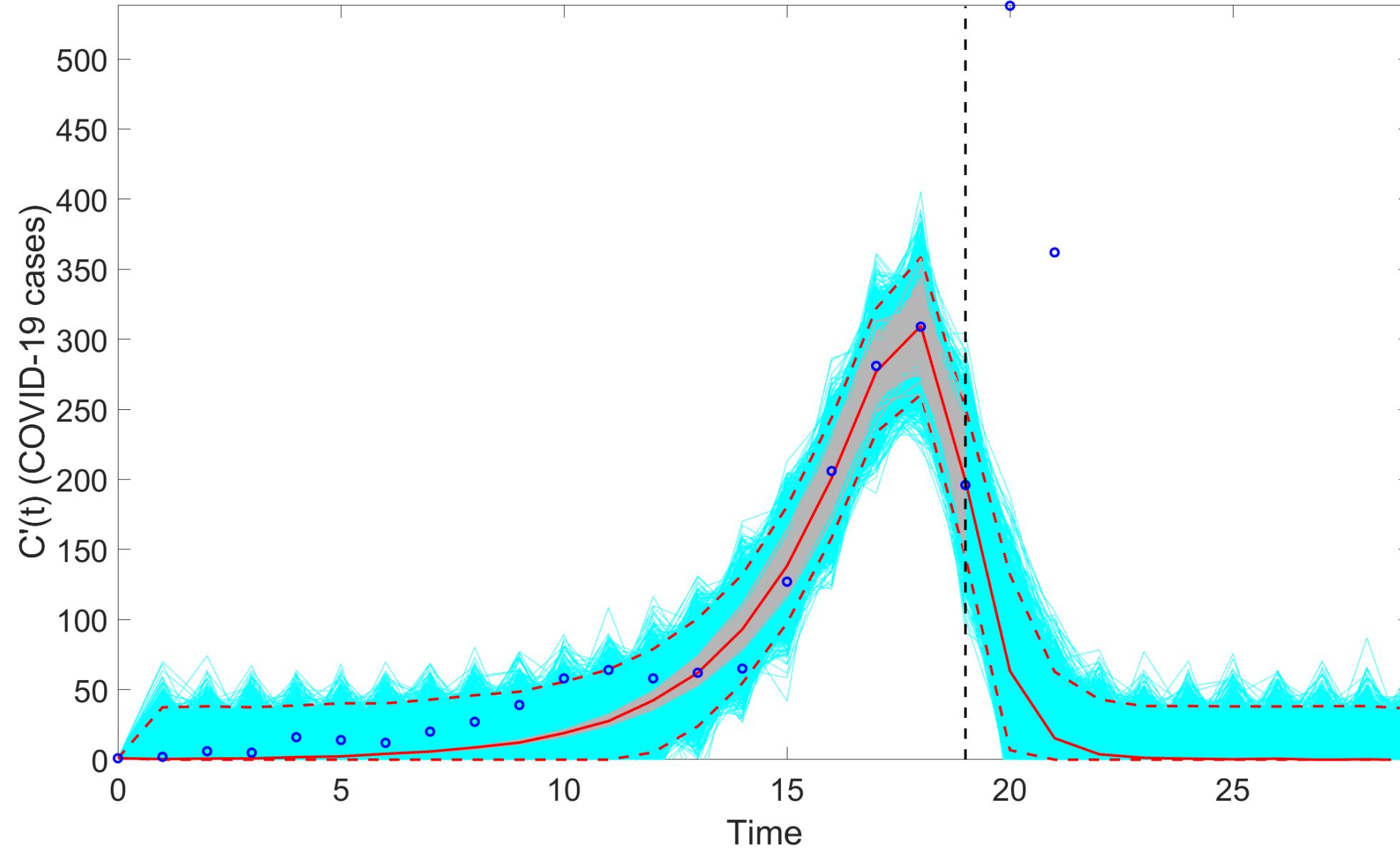
```
% <===== ODE model =====>
% <===== ODE model =====>
% <===== ODE model =====>

model.fc=@RICH; % name of the model function
model.name='RICH model'; % string indicating the name of the ODE model

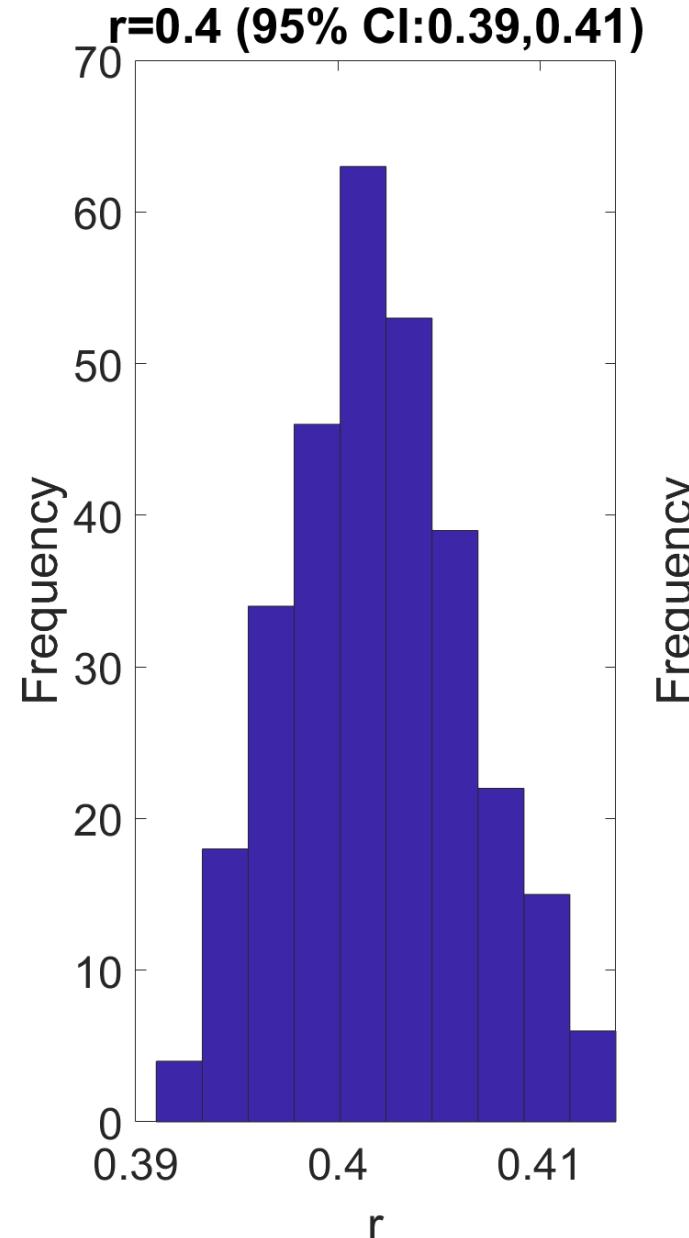
params.label={'r', 'a', 'K0'}; % list of symbols to refer to the model parameters
params.LB=[0 0 10]; % lower bound values of the parameter estimates
params.UB=[10 10 40000]; % upper bound values of the parameter estimates
params.initial=[0.91 0.79 30994]; % initial parameter values/guesses
params.fixed=[0 0 0]; % Boolean vector to indicate any parameters that should remain fixed (1) to initial values indicated above
params.fixI0=1; % Boolean variable indicating if the initial value of the fitting variable is fixed according to the fixed parameter
params.composite=''; % Estimate a composite function of the individual model parameter estimates otherwise it is left empty
params.composite_name=''; % Name of the composite parameter
params.extra0=[]; % used to pass any extra parameters (e.g., data, static variables) to the model function

vars.label={'C'}; % list of symbols to refer to the variables included in the model
vars.initial=5; % vector of initial conditions for the model variables
vars.fit_index=1; % index of the model's variable that will be fit to the observed time series data
vars.fit_diff=1; % boolean variable to indicate if the derivative of model's fitting variable should be fit to data.
```

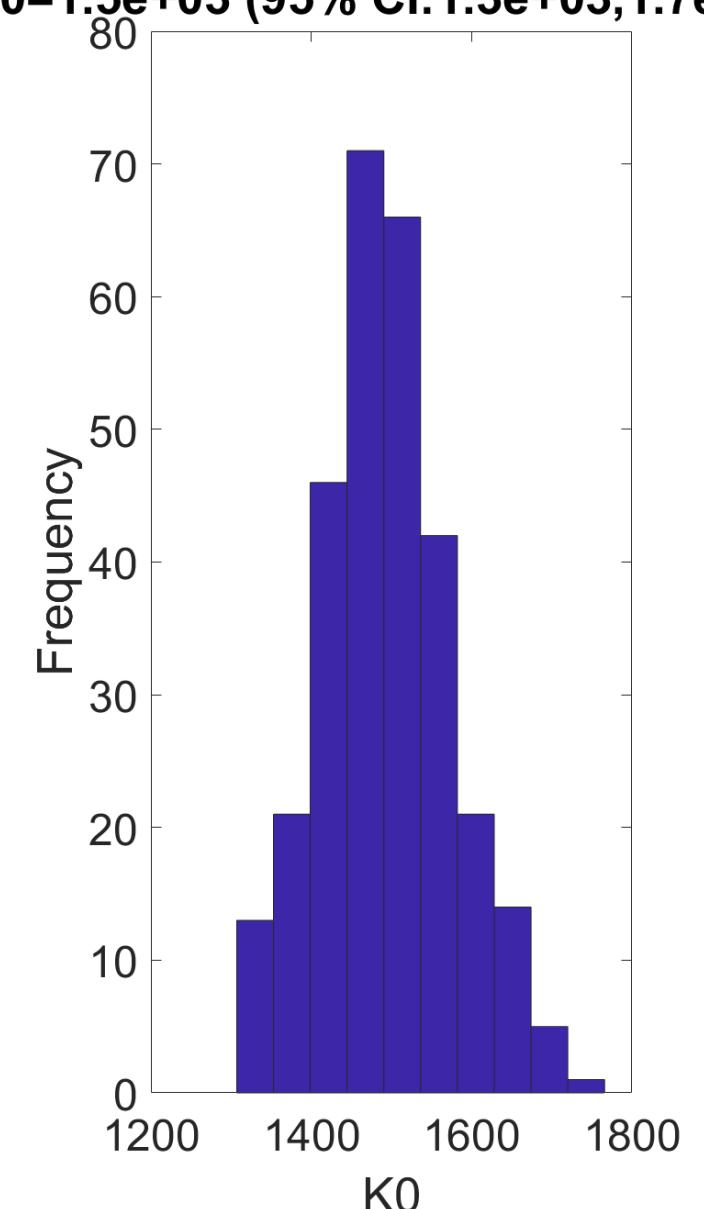
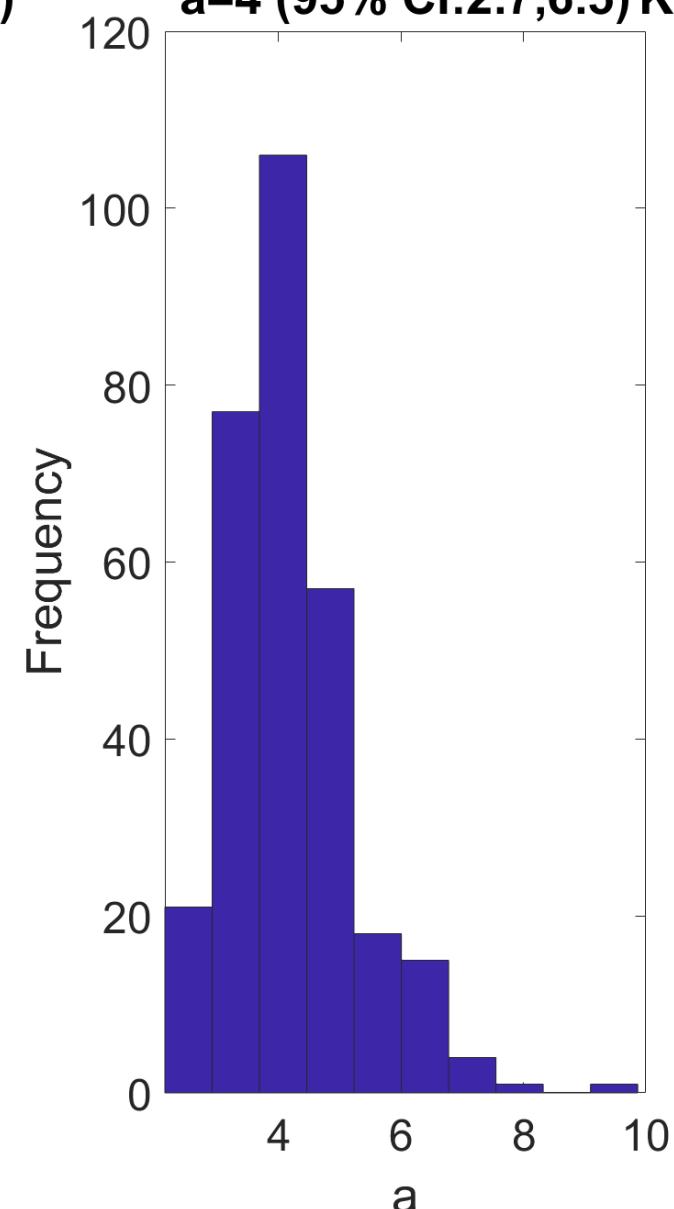
RICH model



$r=0.4$ (95% CI:0.39,0.41)



$a=4$ (95% CI:2.7,6.5) $K0=1.5e+03$ (95% CI:1.3e+03,1.7e+03)



Model	MAE	MSE	Coverage 95% PI	WIS
Calibration Performance				
SEIUR model with NLSQ/ Normal error structure ($\langle \text{dist1} \rangle = 0$)	40.1	3299.6	90.0	23.8
GGM with Normal error structure ($\langle \text{dist1} \rangle = 0$)	21.1	1191.0	100.0	15.2
RICH with negative binomial error structure ($\langle \text{dist1} \rangle = 0$)	22.6	1037.1	100.0	14.0
Forecast Performance				
SEIUR model with NB error structure ($\langle \text{dist1} \rangle = 0$)	4475.0	46849766.5	10.0	3893.0
GGM with negative binomial error structure ($\langle \text{dist1} \rangle = 0$)	254.6	91255.1	70.0	171.5
RICH with negative binomial error structure ($\langle \text{dist1} \rangle = 0$)	533.4	368516.5	20.0	428.6

Specifying SEIURC

All Data

```

% <=====
% < Author: Gerardo Chowell  =====>
% <=====

function dx=SEIR1(t,x,params0,extra0)

beta0=params0(1);
beta1=params0(2);
alpha=params0(3); % non-homogenous mixing parameter
q1=params0(4);
rho=params0(5);
k=params0(6);
gamma1=params0(7);
N=params0(8);

if t<20

    betaf=beta0;

else

    %betaf=beta0*(1./(t+1)).^q1; % power-law decline
    %betaf=beta0*exp(-q1*t); % exponential decline
    betaf=beta1+(beta0-beta1)*exp(-q1*(t-20)); % exponential decline

end

dx=zeros(6,1); % define the vector of the state derivatives: S, E, I, U, R, C

dx(1,1)= -betaf*x(1,1).*((x(3,1)+x(4,1)).^alpha)./N; %S

dx(2,1)= betaf*x(1,1).*((x(3,1)+x(4,1)).^alpha)./N - k*x(2,1); %E

dx(3,1)= k*rho*x(2,1) - gamma1*x(3,1); %I

dx(4,1)= k*(1-rho)*x(2,1) - gamma1*x(4,1); %U

dx(5,1)= gamma1*(x(3,1)+x(4,1)); %R

dx(6,1)= k*rho*x(2,1); %C

```

$$f(\beta) = \begin{cases} \beta_0, & t < 20 \\ \beta_1 + (\beta_0 - \beta_1)e^{-q_1(t-20)}, & \text{Otherwise} \end{cases}$$

$$\left\{ \begin{array}{l} \dot{S} = -f(\beta)S(t) \frac{I(t) + U(t)^\alpha}{N} \\ \dot{E} = f(\beta)S(t) \frac{I(t) + U(t)^\alpha}{N} - \kappa E(t) \\ \dot{I} = \kappa \rho E(t) - \gamma I(t) \\ \dot{U} = \kappa(1 - \rho)E(t) - \gamma U(t) \\ \dot{R} = \gamma(I(t) + U(t)) \\ \dot{C} = \kappa \rho E(t) \end{array} \right.$$

```

% <===== ODE model =====>
% <===== ODE model =====>
% <===== ODE model =====>

model.fc=@SEIR_unreported_expo_decline; % name of the model function
model.name='SEIR_swiss_covid_underreporting'; % string indicating the name of the ODE model

params.label={'\beta_0','\beta_1','\alpha','q','\rho','\kappa','\gamma','N'}; % list of symbols to refer to the model parameters
params.LB=[0.001 0.5 0 0.01 0.01 0.01 47332614]; % lower bound values of the parameter estimates
params.UB=[5 2 1 10 1 1 1 47332614]; % upper bound values of the parameter estimates
params.initial=[0.6 0.01 1 0.01 1 1/5 1/4 47332614]; % initial parameter values/guesses
params.fixed=[0 0 1 0 0 1 1 1]; % Boolean vector to indicate any parameters that should remain fixed (1) to initial values indicated in params.
params.fixI0=1; % Boolean variable indicating if the initial value of the fitting variable is fixed according to the first observation in the t
params.composite=@R0s_unreported; % Estimate a composite function of the individual model parameter estimates otherwise it is left empty.
params.composite_name='R0s_unreported'; % Name of the composite parameter
params.extra0='';

vars.label={'S','E','I','U','R','C'}; % list of symbols to refer to the variables included in the model
vars.initial=[params.initial(8)-1 0 1 0 0 1]; % vector of initial conditions for the model variables
vars.fit_index=6; % index of the model's variable that will be fit to the observed time series data
vars.fit_diff=1; % boolean variable to indicate if the derivative of model's fitting variable should be fit to data.

```

Preparing options_fit.m

Fitting using all data

```
% <===== Parameters of the rolling window analysis =====>
% <===== Parameters of the rolling window analysis =====>
% <===== Parameters of the rolling window analysis =====>

windowsize1=109; % moving window size

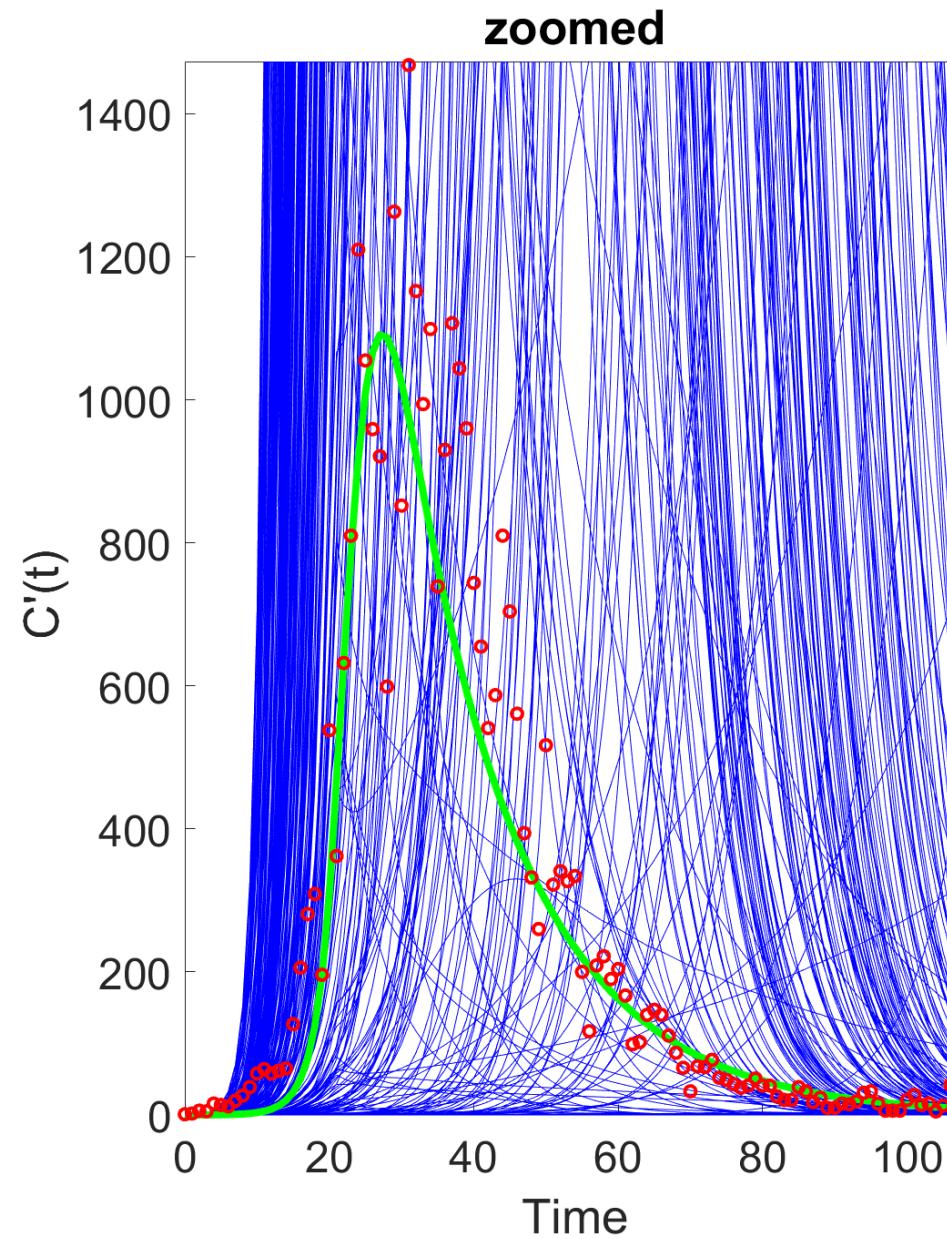
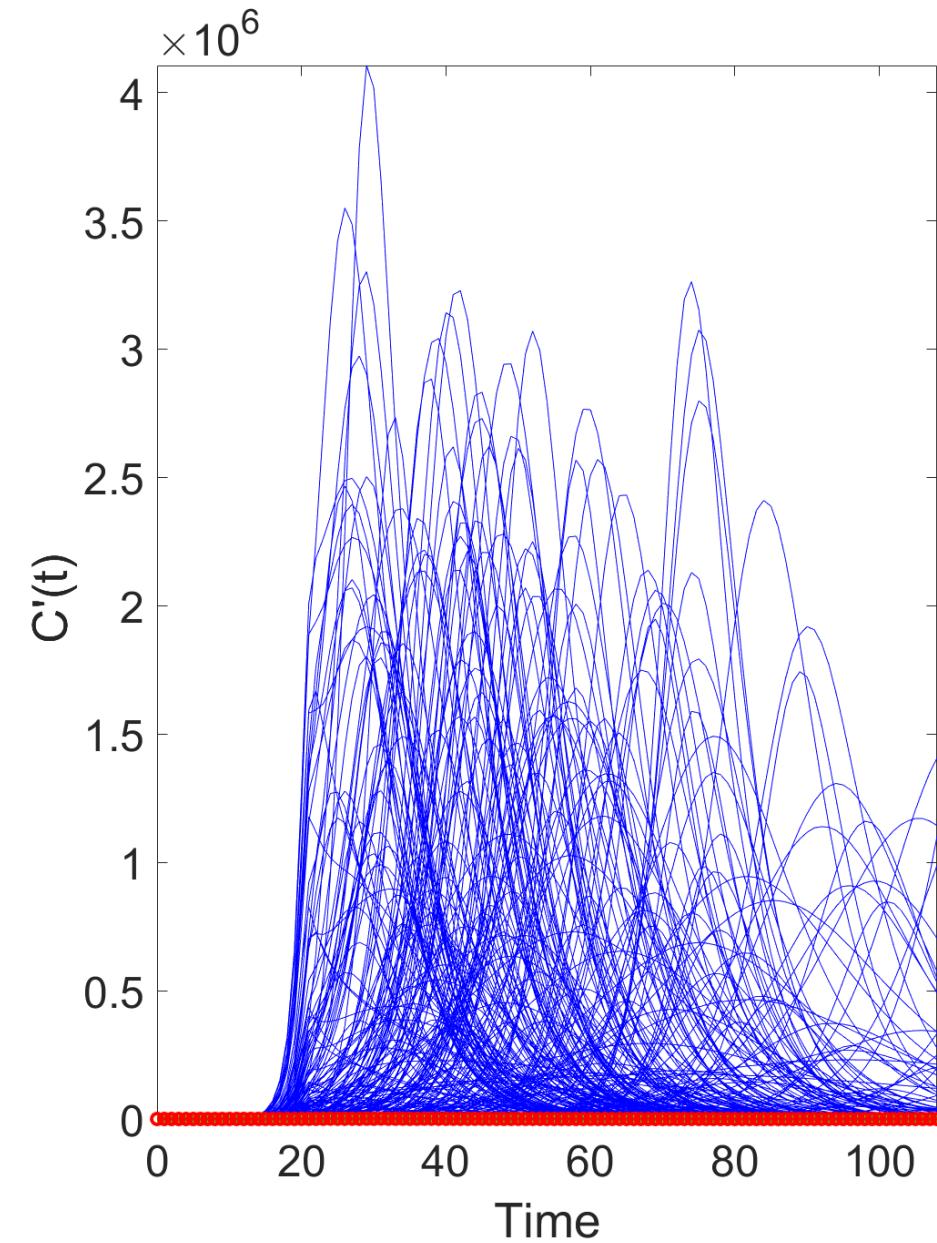
tstart1=1; % time point for the start of rolling window analysis

tend1=1; %time point for the end of the rolling window analysis

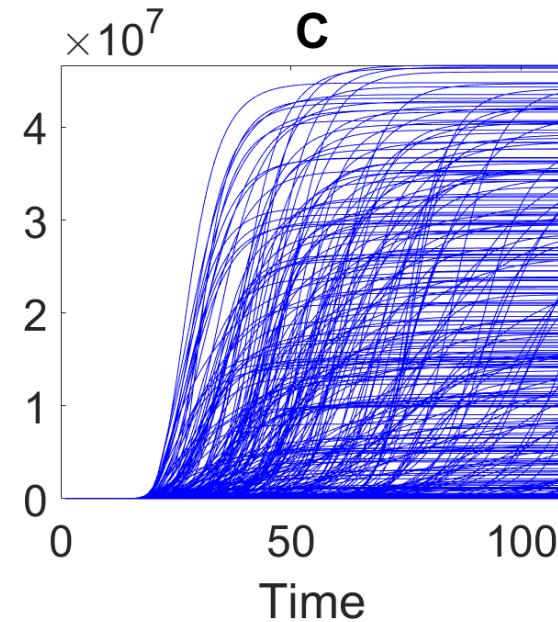
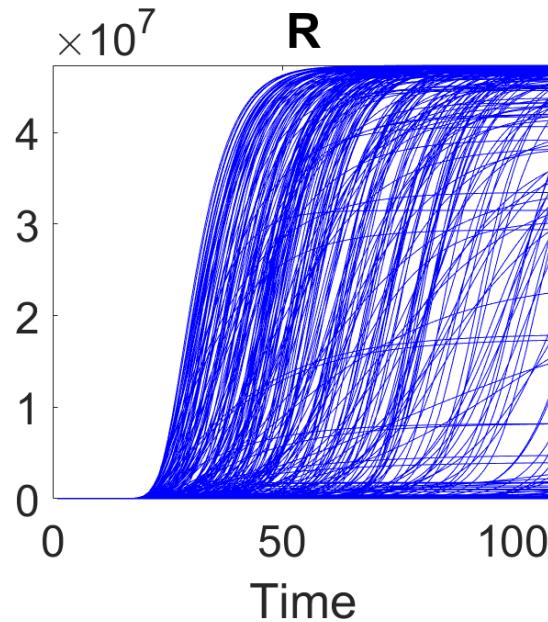
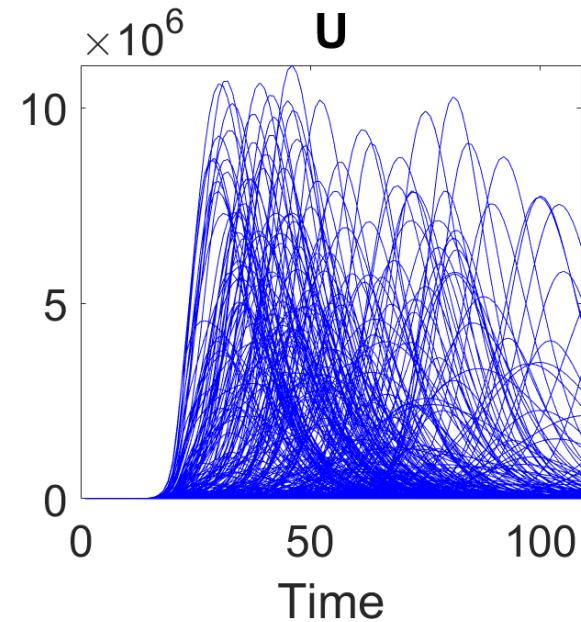
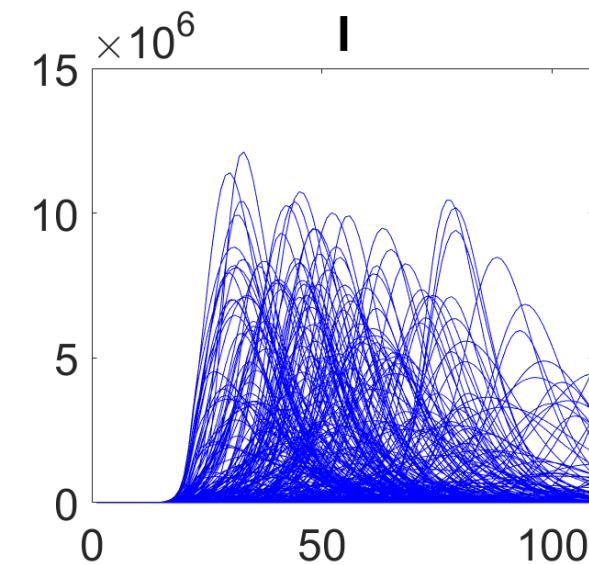
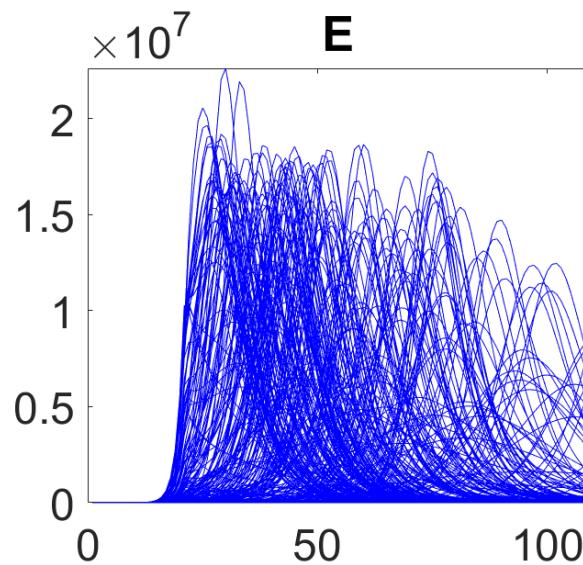
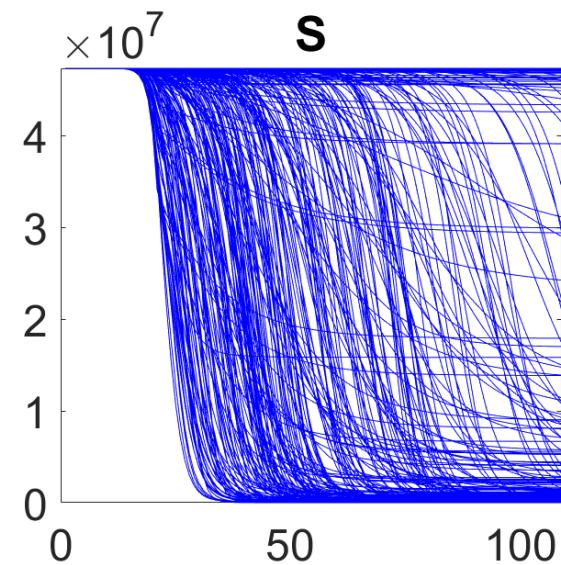
printscreen1=1;
```

Generating preliminary model solutions

```
plotODEModel(@options_fit_SEIR_unreported_covid_swiss_dist1_0)
```



Estimated α

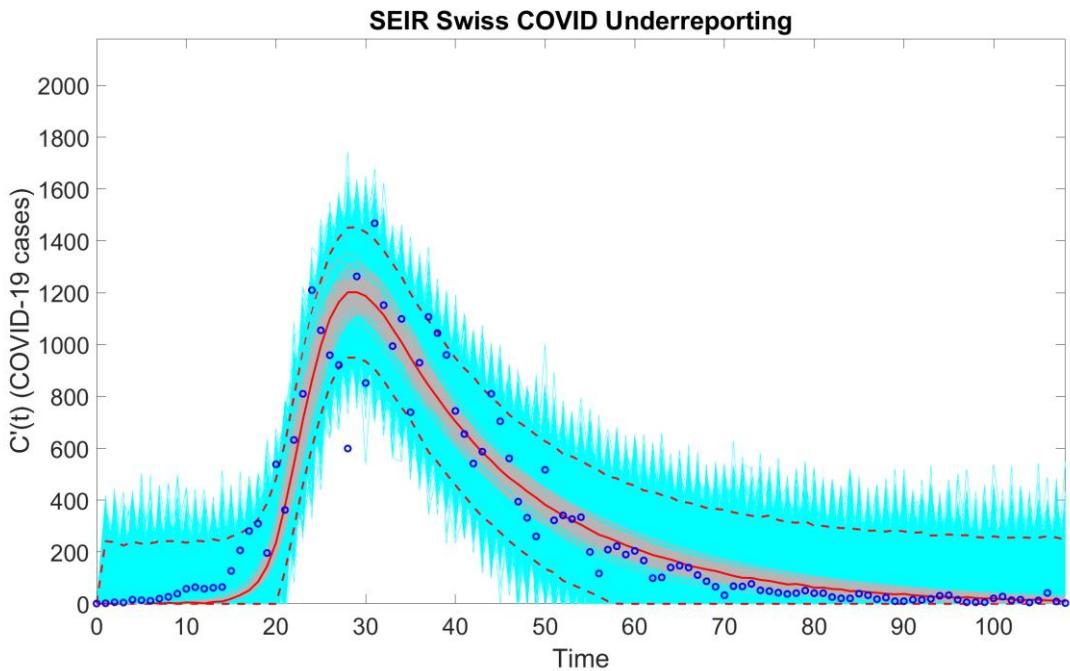


Estimated α

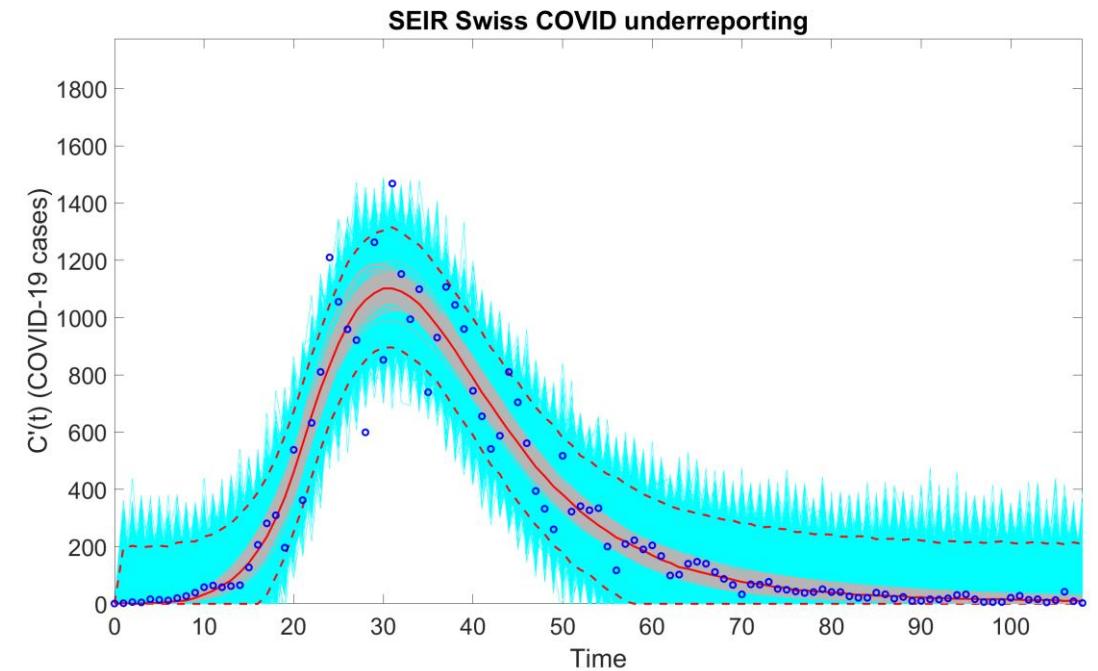
Fitting, plotting, and evaluating the model with quantified uncertainty

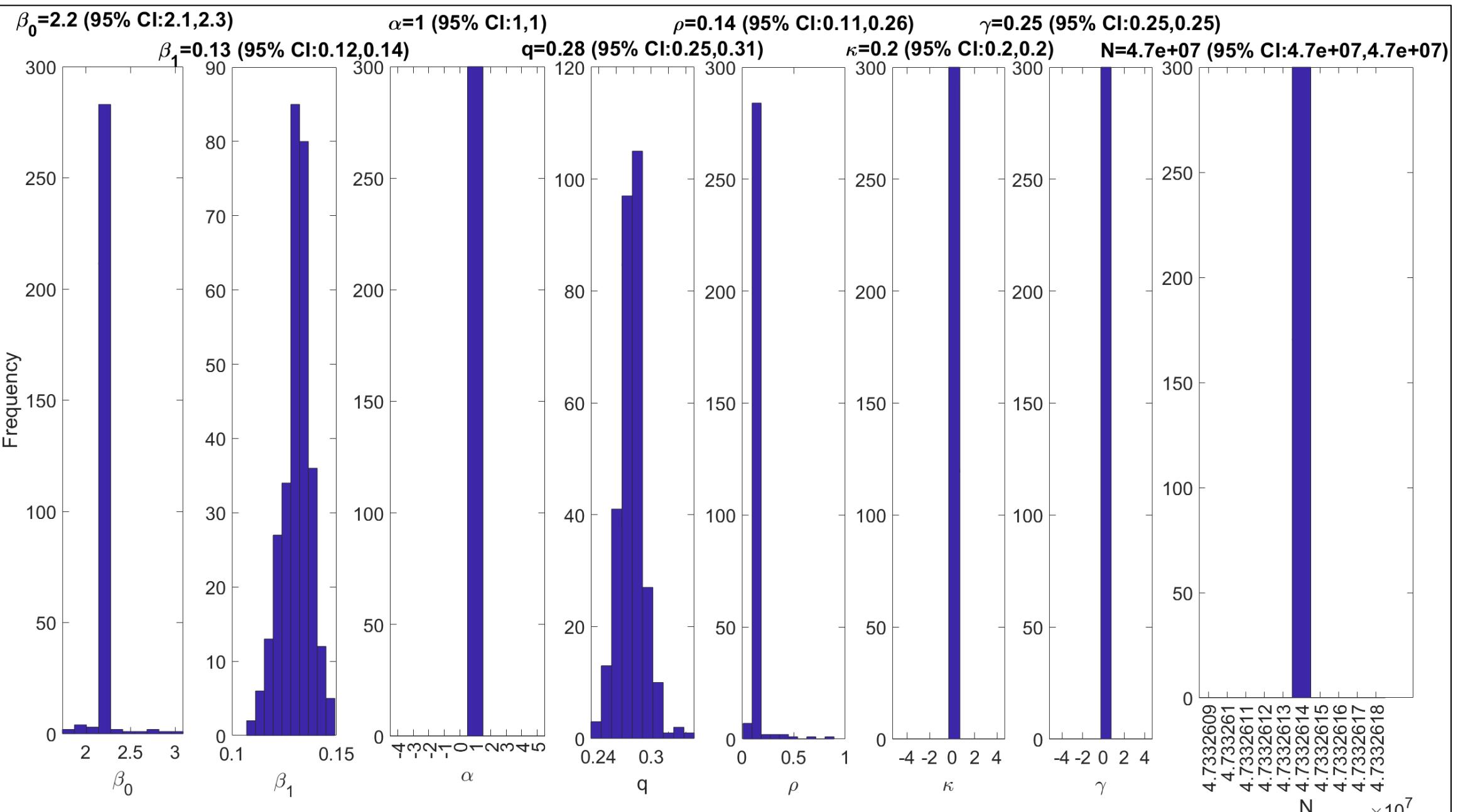
```
(1) Run_Fit_ODEModel(@options_fit_SEIR_unreported_covid_swiss_dist1_0  
, 1, 1, 109)  
(2) plotForecast_ODEModel(@options_fit_SEIR_unreported_covid_swiss_di  
st1_0, 1, 1, 109)
```

Fixed α

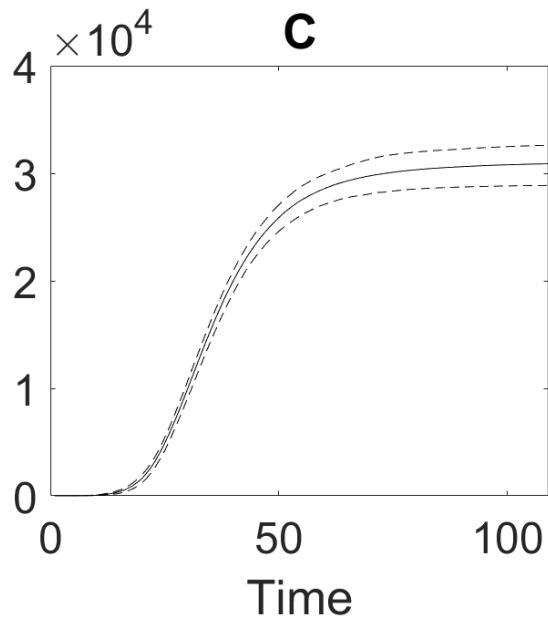
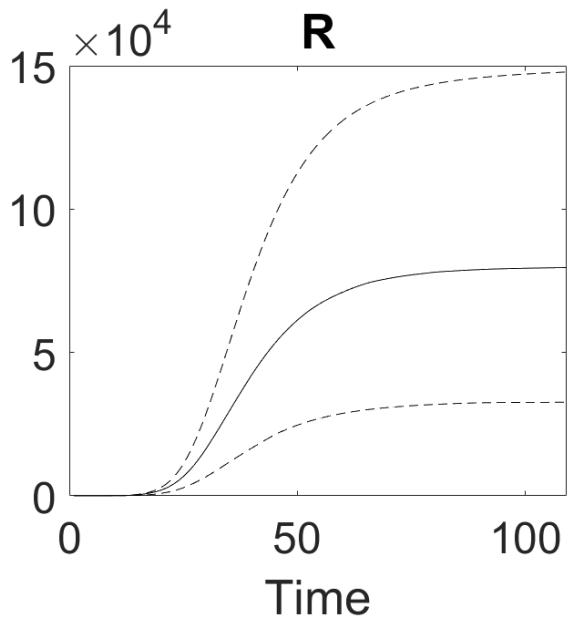
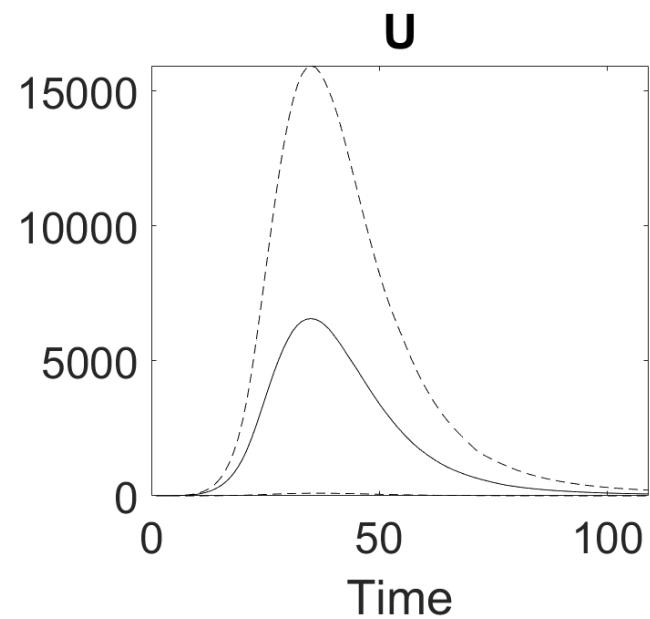
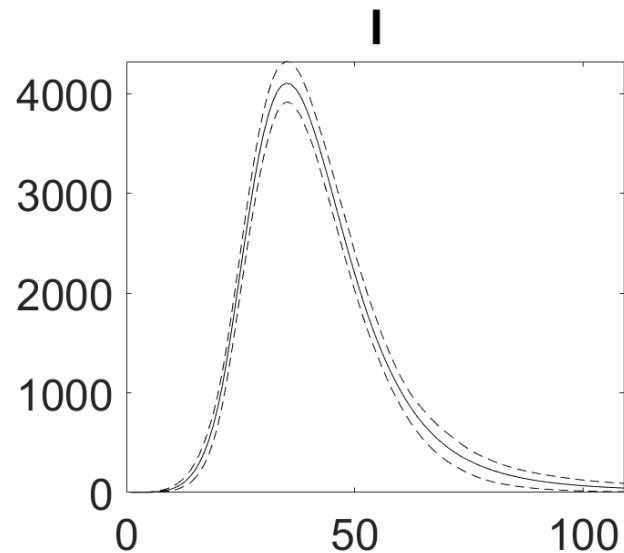
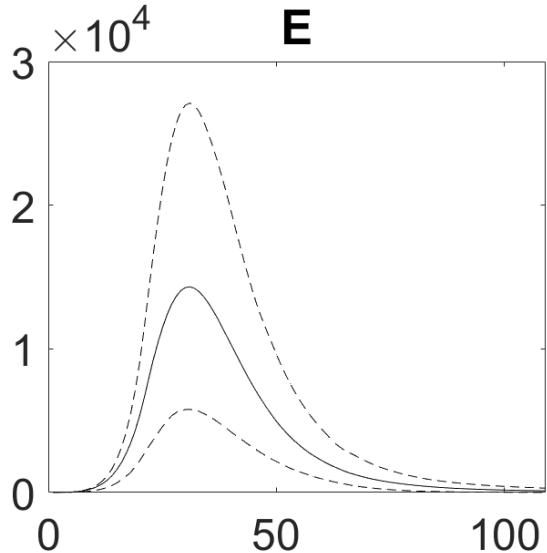
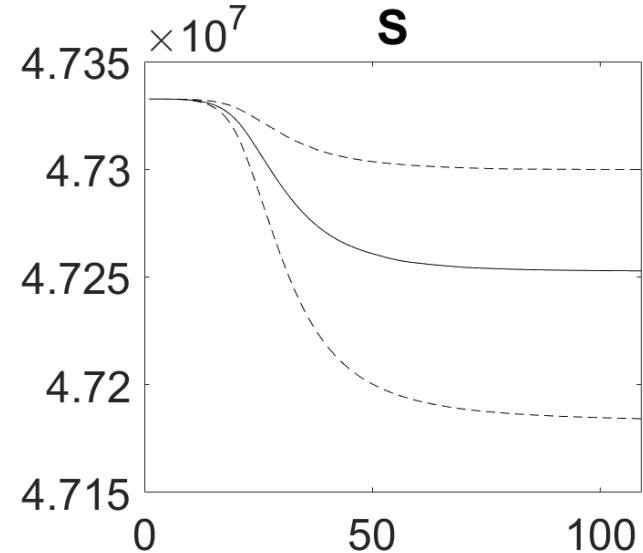


Estimated α





Estimated α

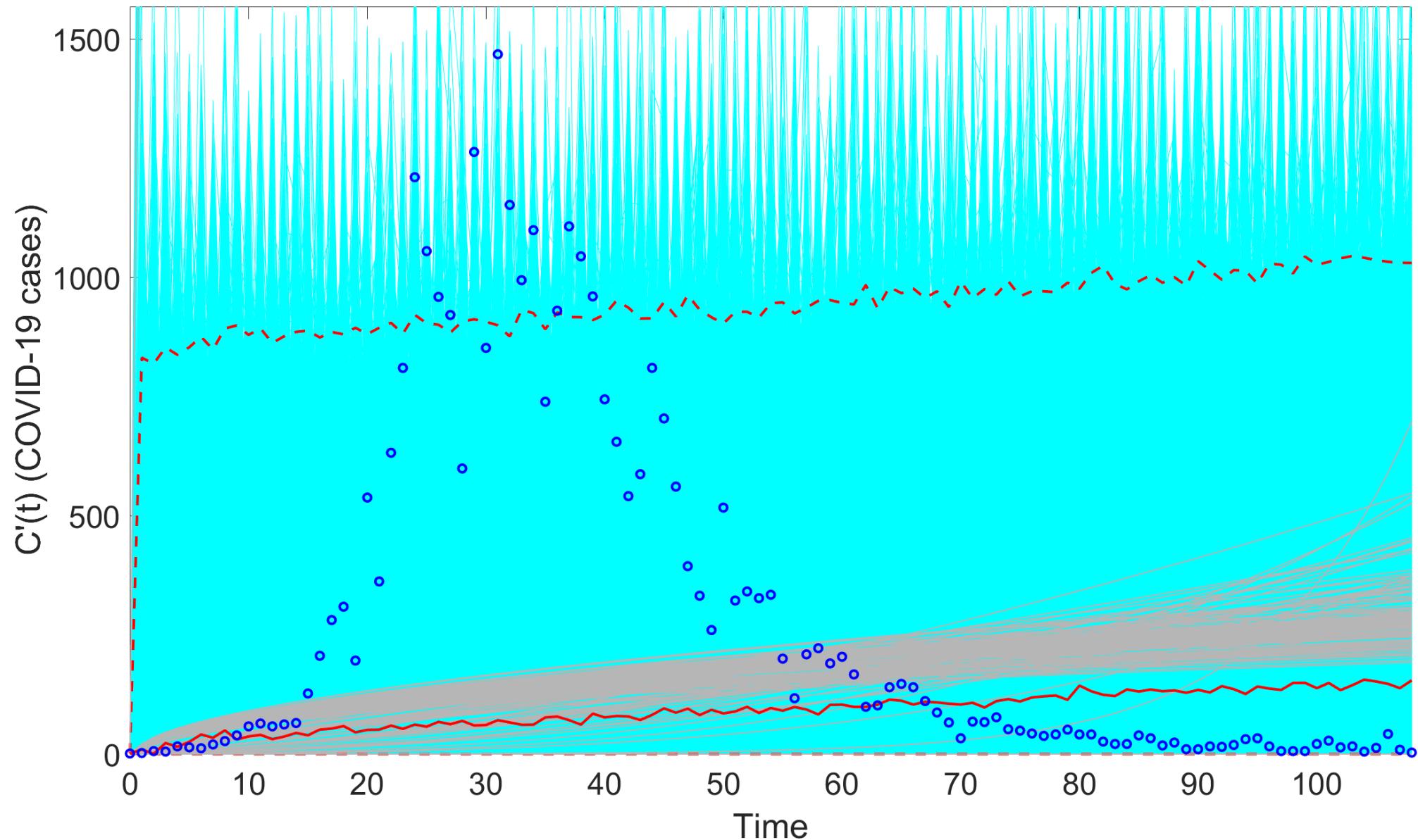


Estimated α

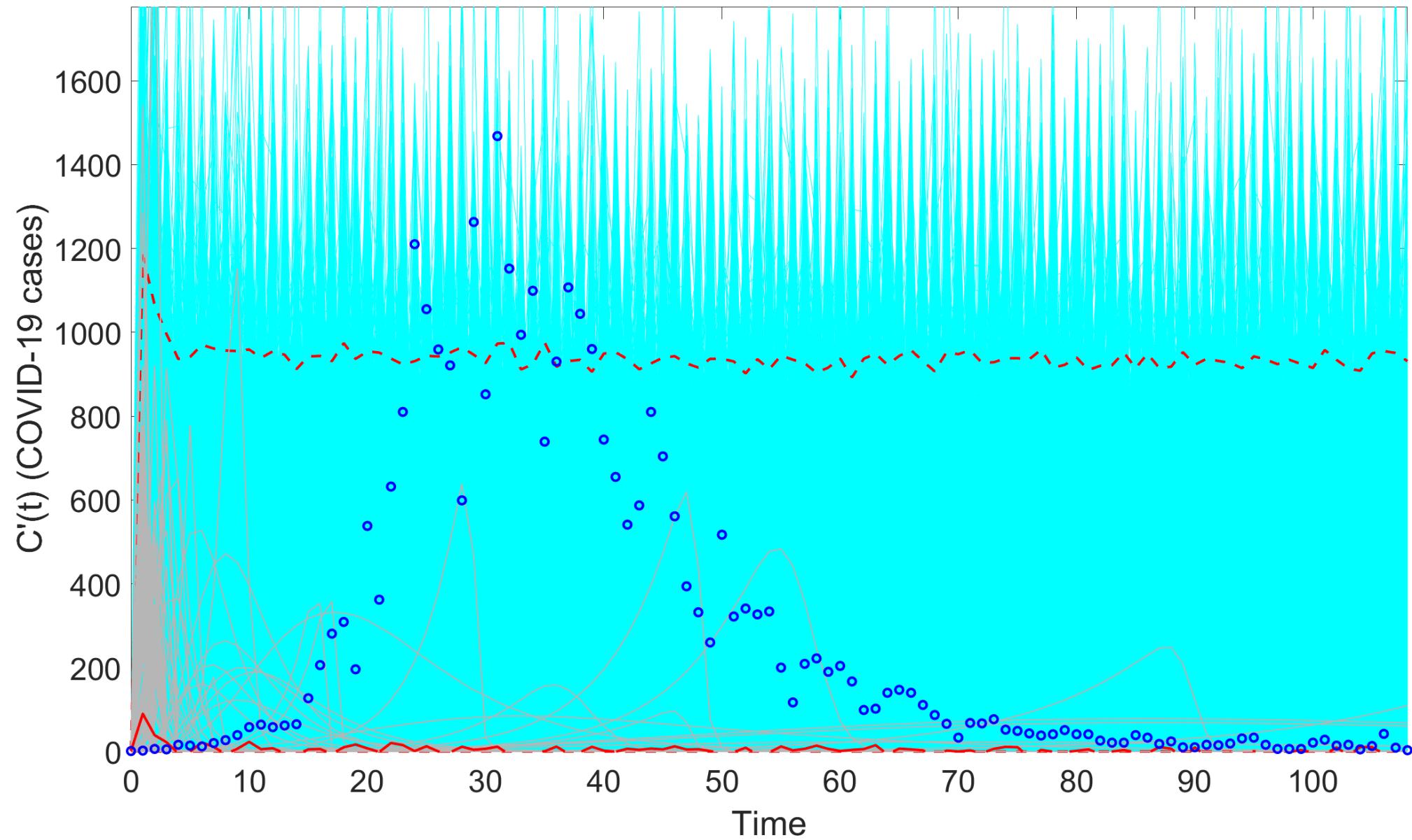
Model Comparison (All Data)

SEIUR Normal, GGM Normal & RICH Normal

GGM model



RICH model



Model	MAE	MSE	Coverage 95% PI	WIS
Calibration Performance				
SEIUR model with NLSQ/ Normal error structure ($\langle \text{dist1} \rangle = 0$)	67.2	13189.4	92.7	48.3
GGM with Normal error structure ($\langle \text{dist1} \rangle = 0$)	297.6	197605.8	88.1	176.6
RICH with negative binomial error structure ($\langle \text{dist1} \rangle = 0$)	284.3	216949.7	89.0	173.1