Estimating and Simulating a SIRD Model of COVID-19 for Many Countries, States, and Cities

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Extended results for Miami

Based on data through October 9, 2020

Outline of Slides

- Basic data from Johns Hopkins CSSE (raw and smoothed)
- Brief summary of the model
- Baseline results ($\delta = 1.0\%$, $\gamma = 0.2$, $\theta = 0.1$)
- Simulation of re-opening – possibilities for raising R_0
- Results with alternative parameter values:
    - Lower mortality rate, $\delta = 0.8\%$
    - Higher mortality rate, $\delta = 1.2\%$
    - Infections last longer, $\gamma = 0.15$
    - Cases resolve more quickly, $\theta = 0.2$
    - Cases resolve more slowly, $\theta = 0.07$
- Data underlying estimates of $R_0(t)$

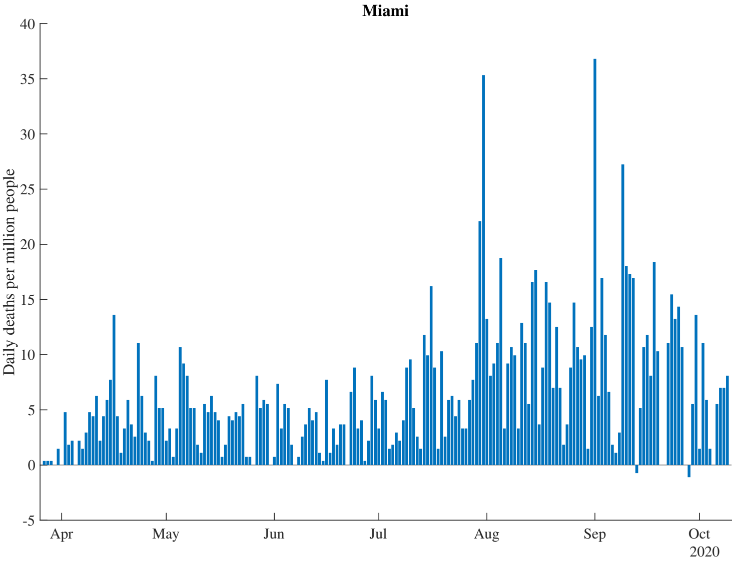
Underlying data from Johns Hopkins CSSE

– Raw data

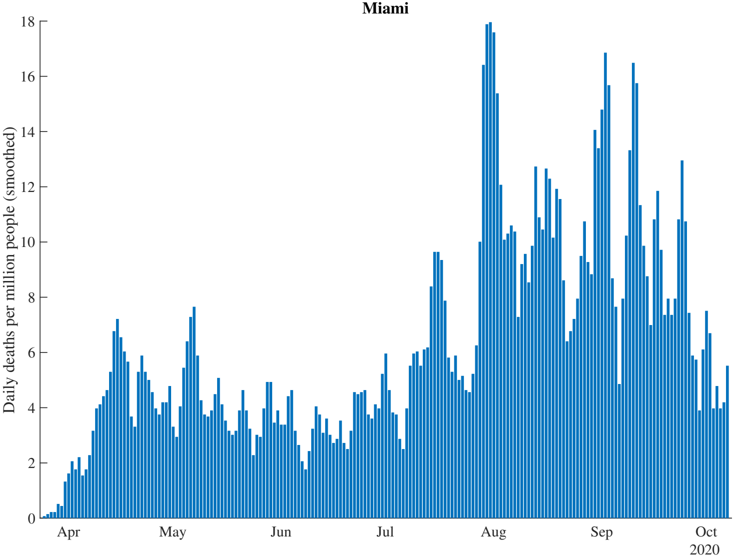
– Smoothed = 7 day centered moving average

– No “excess deaths” correction (change as of Aug 6 run)

Miami: Daily Deaths per Million People



Miami: Daily Deaths per Million People (Smoothed)



Brief Summary of Model

- See the paper for a full exposition
- A 5-state SIRDC model with a time-varying R_0

Parameter	Baseline	Description
δ	1.0%	Mortality rate from infections (IFR)
γ	0.2	Rate at which people stop being infectious
θ	0.1	Rate at which cases (post-infection) resolve
α	0.05	Rate at which $R_0(t)$ decays with daily deaths
R_0	...	Initial base reproduction rate
$R_0(t)$	...	Base reproduction rate at date t (β_t/γ)

Estimates of Time-Varying R_0

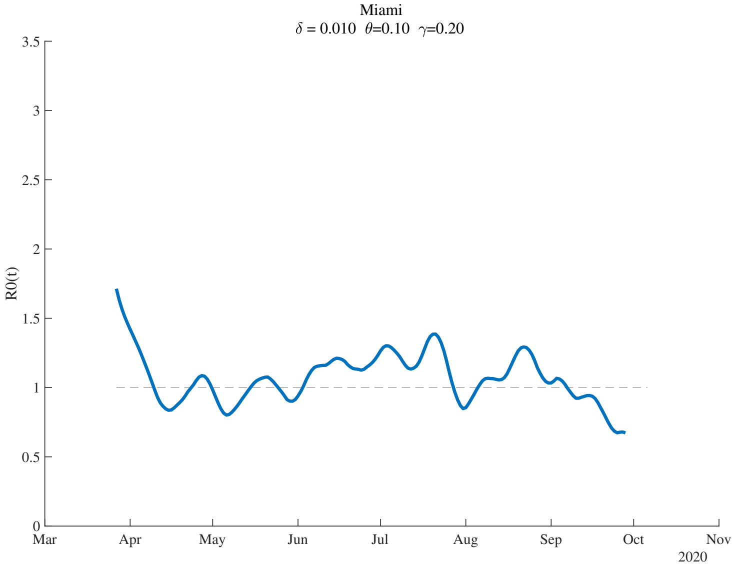
– Inferred from daily deaths, and

– the change in daily deaths, and

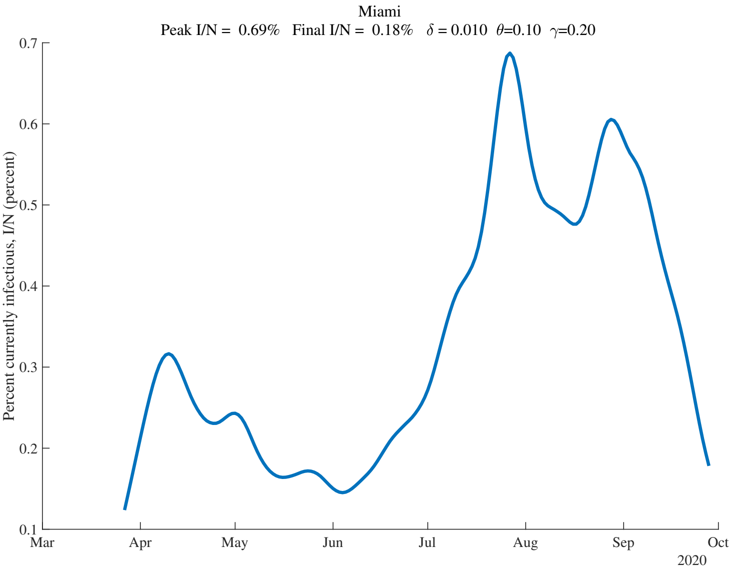
– the change in (the change in daily deaths)

(see end of slide deck for this data)

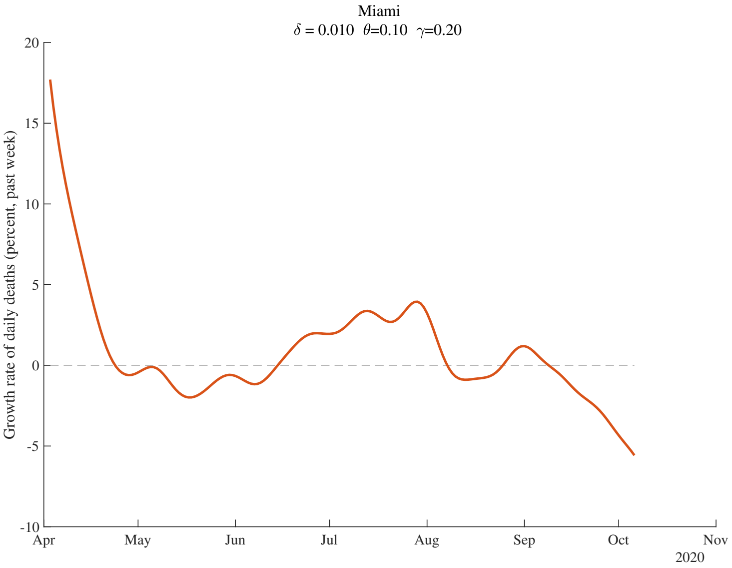
Miami: Estimates of $R_0(t)$



Miami: Percent Currently Infectious



Miami: Growth Rate of Daily Deaths over Past Week (percent)



Notes on Intepreting Results

Guide to Graphs

- Warning: Results are often very uncertain; this can be seen by comparing across multiple graphs. See the original paper.
- 7 days of forecasts: Rainbow color order!
  ROY-G-BIV (old to new, low to high)
  - Black=current
  - Red = oldest, Orange = second oldest, Yellow =third oldest...
  - Violet (purple) = one day earlier
- For robustness graphs, same idea
  - Black = baseline (e.g. $\delta = 1.0\%$)
  - Red = lowest parameter value (e.g. $\delta = 0.8\%$)
  - Green = highest parameter value (e.g. $\delta = 1.2\%$)

How does R_0 change over time?

- Inferred from death data when we have it
- For future, two approaches:

  1. Alternatively, we fit this equation:

$$\log R_0(t) = a_0 - \alpha(\text{Daily Deaths})$$

$$\Rightarrow \alpha \approx .05$$

R_0 declines by 5 percent for each new daily death, or rises by 5 percent when daily deaths decline

- Robustness: Assume $R_0(t) =$ final empirical value. Constant in future, so no α adjustment $\rightarrow \alpha = 0$

Repeated “Forecasts” from the past 7 days of data

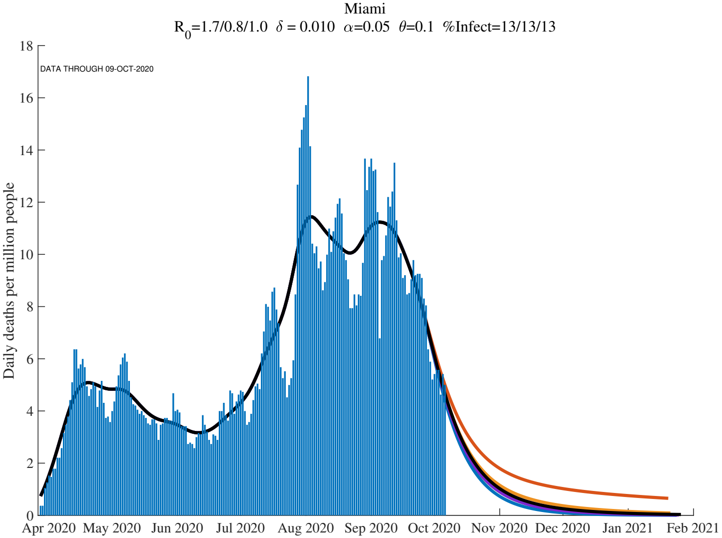
– After peak, forecasts settle down.

– Before that, very noisy!

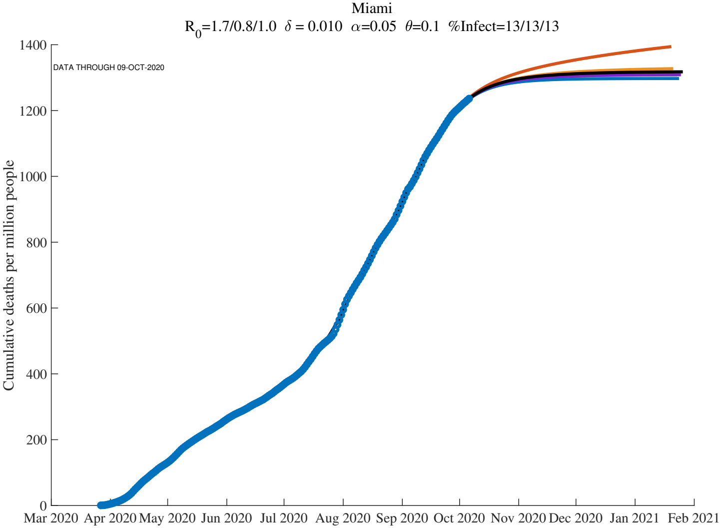
– If the region has not peaked, do not trust

– With $\alpha = .05$ (see robustness section for $\alpha = 0$)

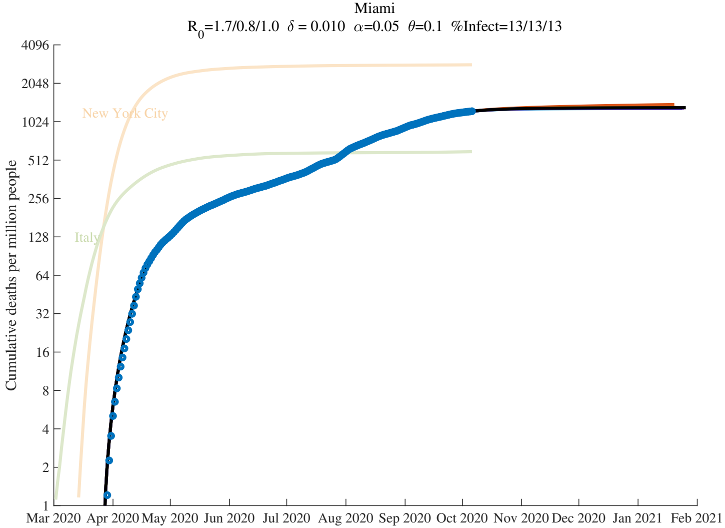
Miami (7 days): Daily Deaths per Million People ($\alpha = .05$)



Miami (7 days): Cumulative Deaths per Million (Future, $\alpha = .05$)

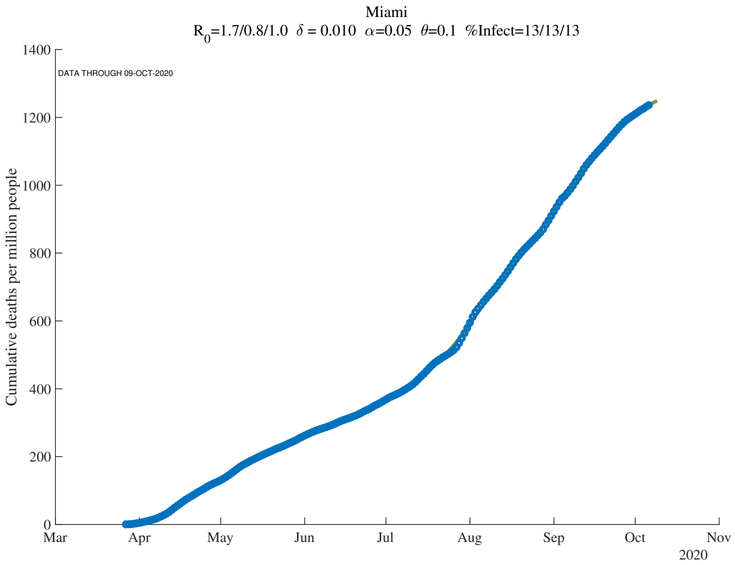


Miami (7 days): Cumulative Deaths per Million, Log Scale ($\alpha = .05$)

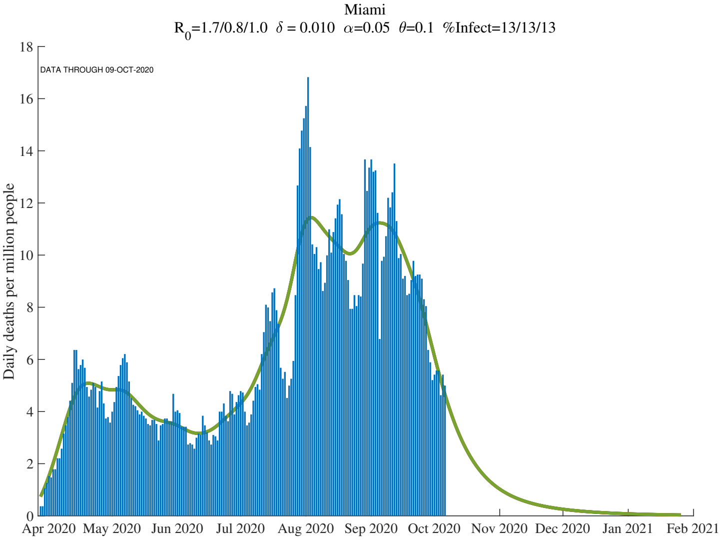


Robustness to Mortality Rate, δ

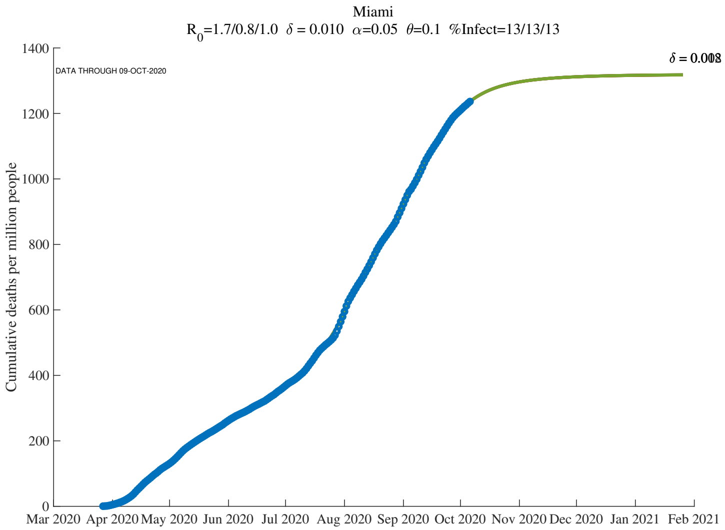
Miami: Cumulative Deaths per Million ($\delta = .01/.008/.012$)



Miami: Daily Deaths per Million People ($\delta = .01/.008/.012$)



Miami: Cumulative Deaths per Million ($\delta = .01/.008/.012$)



Reopening and Herd Immunity

– Black: assumes R_0(today) remains in place forever

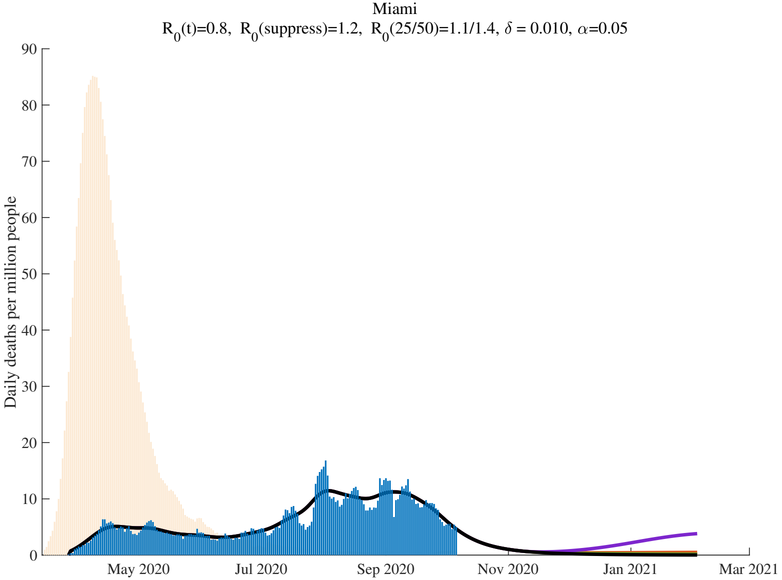
– Red: assumes R_0(suppress)$= 1/s(today)$

– Green: we move 25% of the way from R_0(today) back to initial R_0 = “normal”

– Purple: we move 50% of the way from R_0(today) back to initial R_0 = “normal”

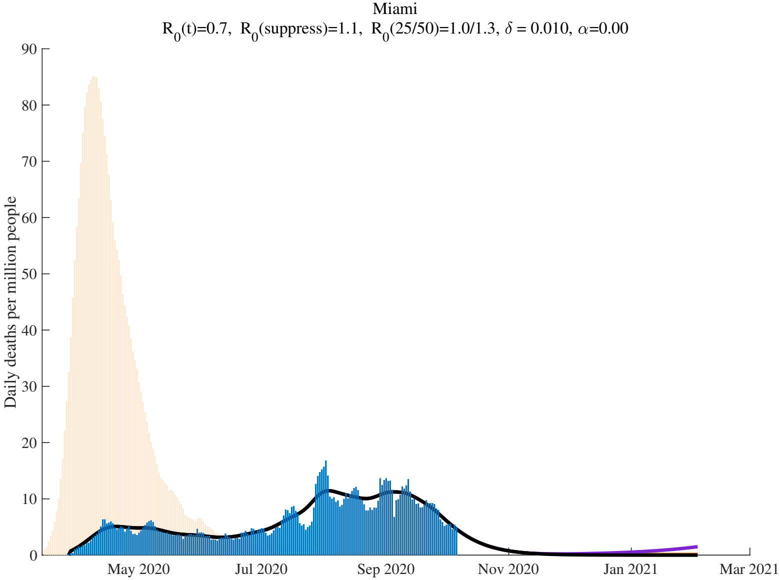
NOTE: Lines often cover each other up

Miami: Re-Opening ($\alpha = .05$)



(Light bars = New York City, for comparison)

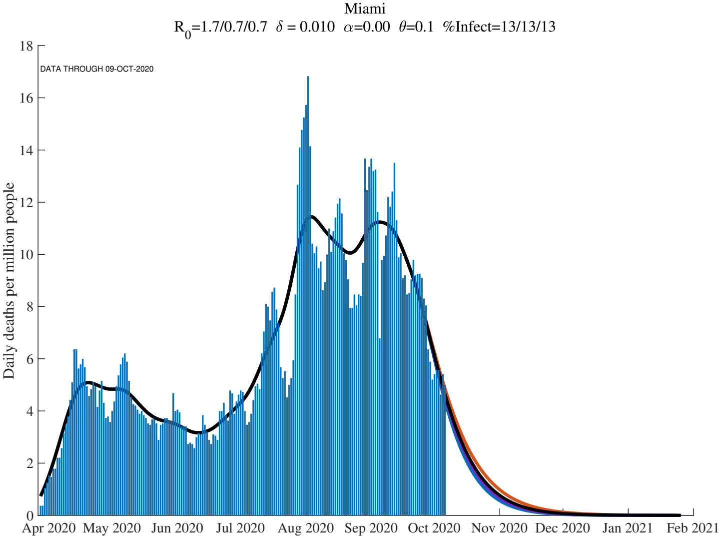
Miami: Re-Opening ($\alpha = 0$)



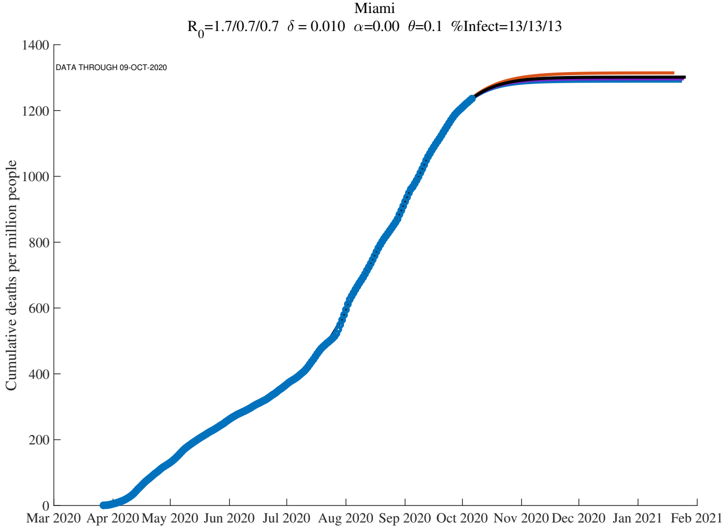
(Light bars = New York City, for comparison)

Results for alternative parameter values

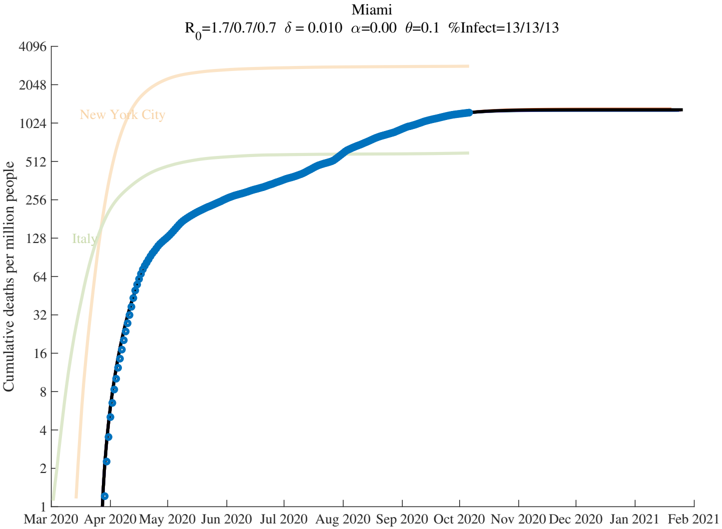
Miami (7 days): Daily Deaths per Million People ($\alpha = 0$)



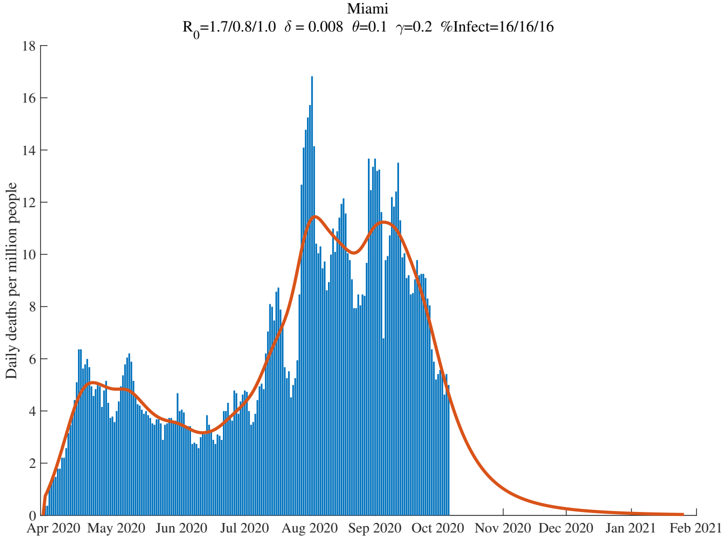
Miami (7 days): Cumulative Deaths per Million (Future, $\alpha = 0$)



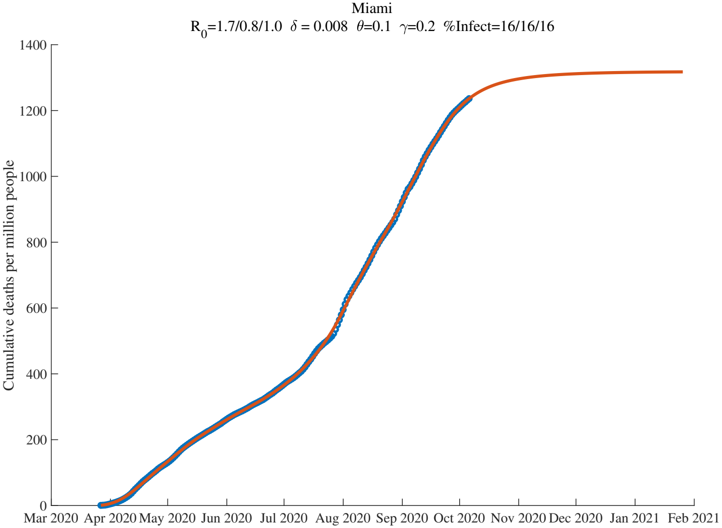
Miami (7 days): Cumulative Deaths per Million, Log Scale ($\alpha = 0$)



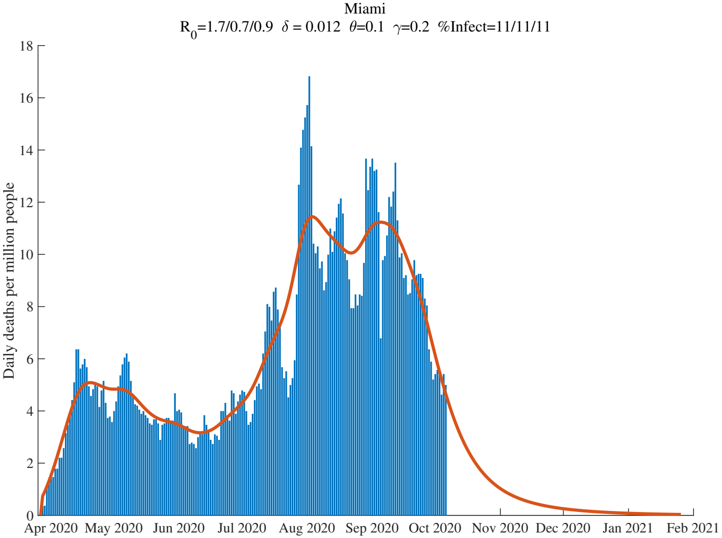
Miami: Daily Deaths per Million People ($\delta = 0.8\%$)



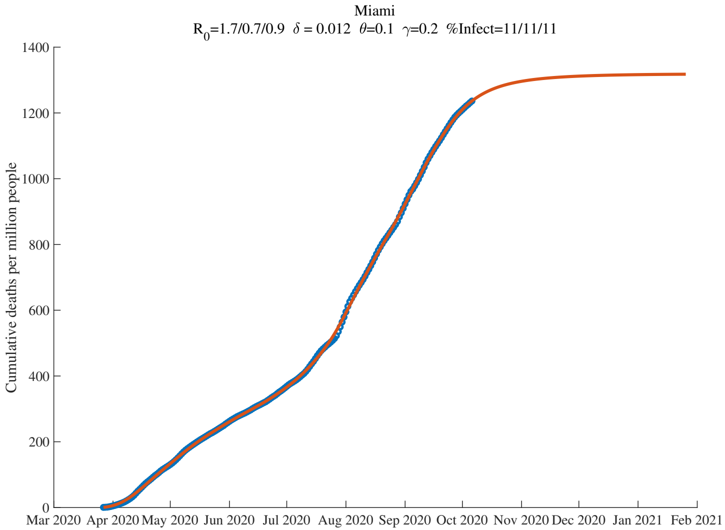
Miami: Cumulative Deaths per Million ($\delta = 0.8\%$)



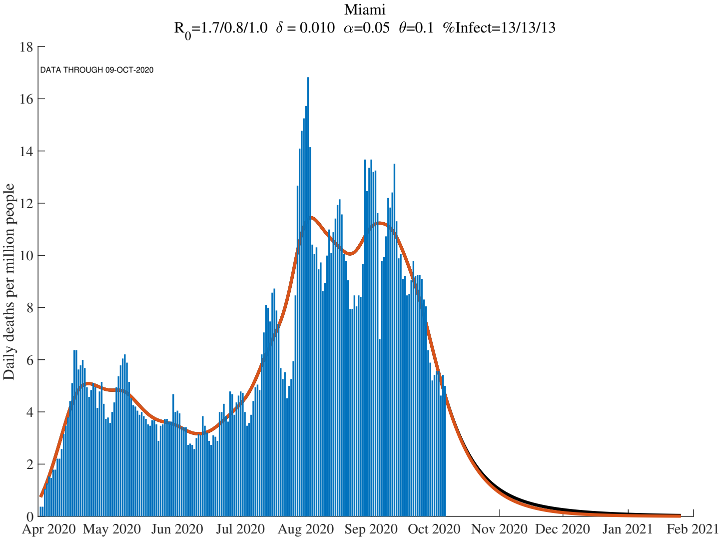
Miami: Daily Deaths per Million People ($\delta = 1.2\%$)



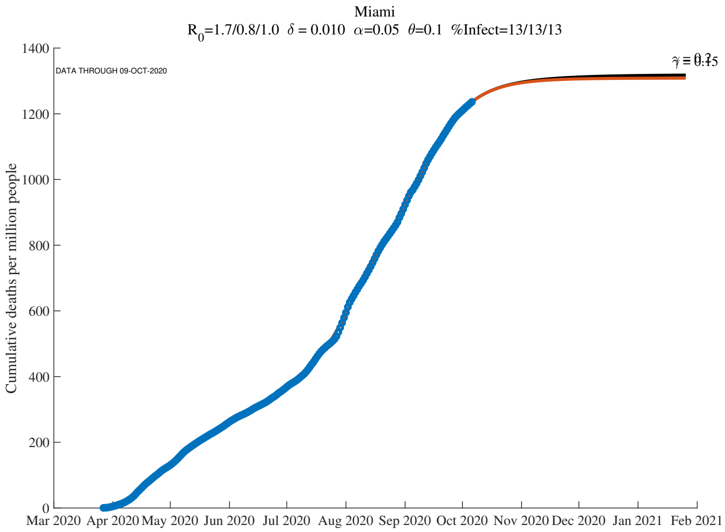
Miami: Cumulative Deaths per Million ($\delta = 1.2\%$)



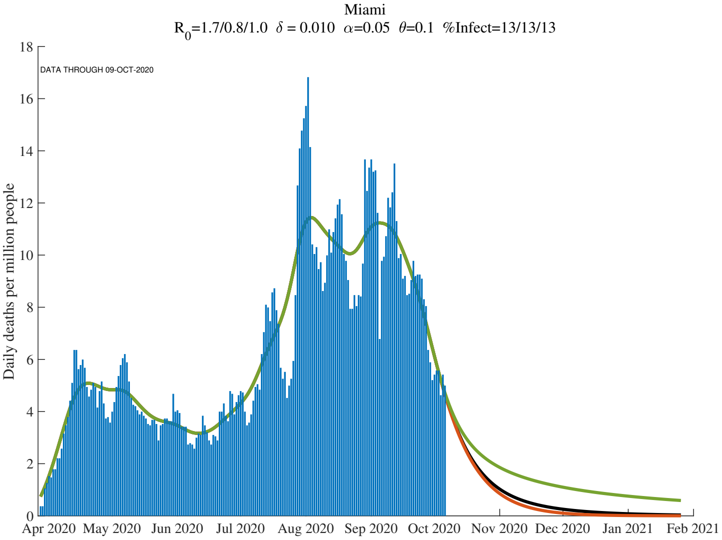
Miami: Daily Deaths per Million People ($\gamma = .2/.15$)



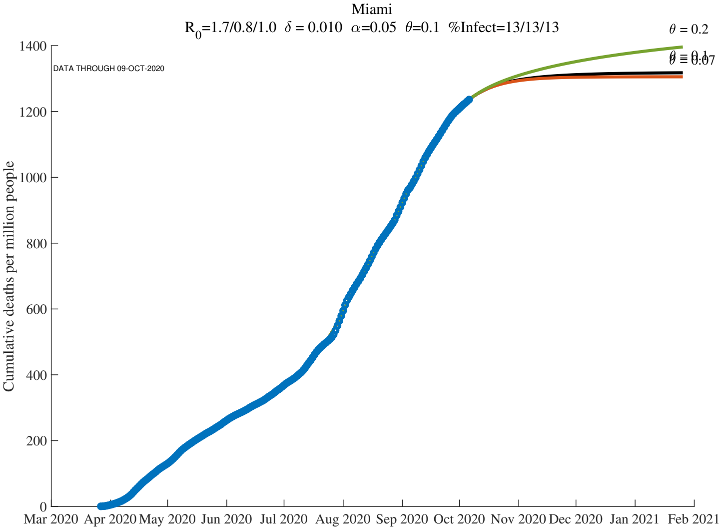
Miami: Cumulative Deaths per Million $\gamma = .2/.15$)



Miami: Daily Deaths per Million People ($\theta = .1/.07/.2$)



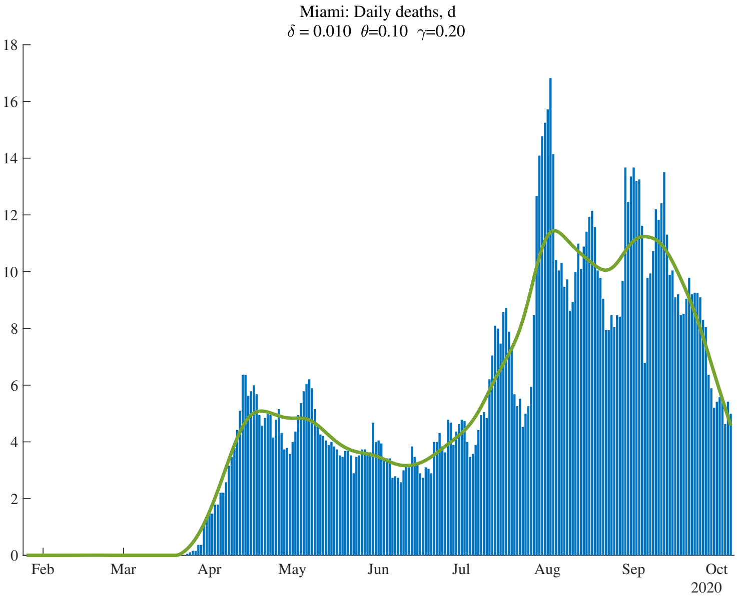
Miami: Cumulative Deaths per Million People ($\theta = .1/.07/.2$)



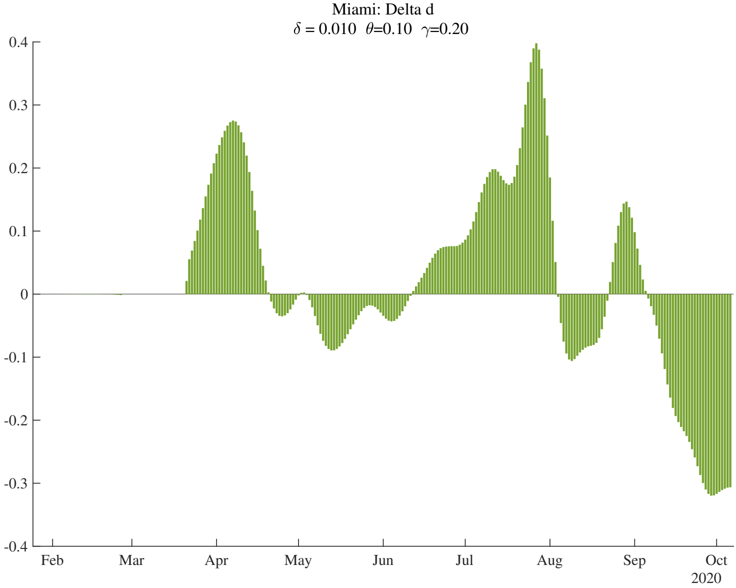
Data Underlying Estimates of Time-Varying R_0

– Inferred from daily deaths, and
– the change in daily deaths, and
– the change in (the change in daily deaths)

Miami: Daily Deaths, Actual and Smoothed



Miami: Change in Smoothed Daily Deaths



Miami: Change in (Change in Smoothed Daily Deaths)