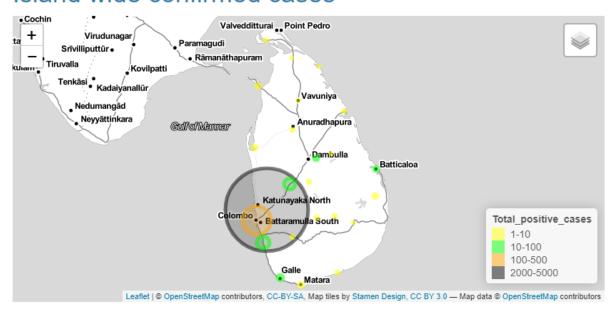
Shiny map implementation for covid-19 positive cases in sri lanka

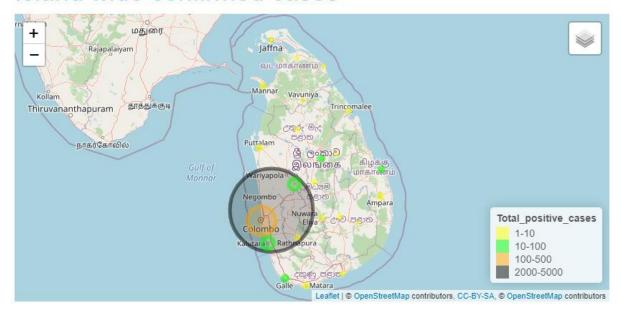
```
output$mymap<-renderLeaflet({
  iri<-read.csv(text =
getURL("https://raw.githubusercontent.com/ishancoderr/Covid 19 sri lanka/main/Distribution co
vid19.csv"))
  iri$Total positive cases=cut(iri$totalcase,
                 breaks = c(0,10,100,500,1000,2000,5000,10000), right = FALSE,
                 labels = c("1-10","10-100","100-500","500-1000","1000-2000","2000-
5000","5000-10000"))
  pal=colorFactor(palette = c("yellow", "green", "orange", "red", "brown", "black", "white"), domain =
iri$rangecase)
  leaflet(data = iri) %>%
   addTiles() %>%
   addProviderTiles(providers$Stamen.TonerLite, group = "Toner Lite") %>%
   addProviderTiles(providers$OpenStreetMap, group = "OSM") %>%
   addLayersControl(baseGroups = c( "Toner Lite", "OSM")) %>%
   setView(Ing=80.65167,lat =7.86,zoom = 7) %>%
   #addMarkers(Ing= ~LON,lat = ~ LAT)
   addCircleMarkers(Ing= ~LON,lat = ~ LAT,
            color =~pal(Total_positive_cases),
            radius = ~sqrt(totalcase)*1,
            label = paste("",iri$District,"POSITIVE:",iri$totalcase))%>%
   addLegend(position = "bottomright",pal = pal,values=~Total_positive_cases)
```

})

Island wide confirmed cases



Island wide confirmed cases



Full Shiny app

Link → https://ishanweerakoon11.shinyapps.io/covid-sri-lanka/

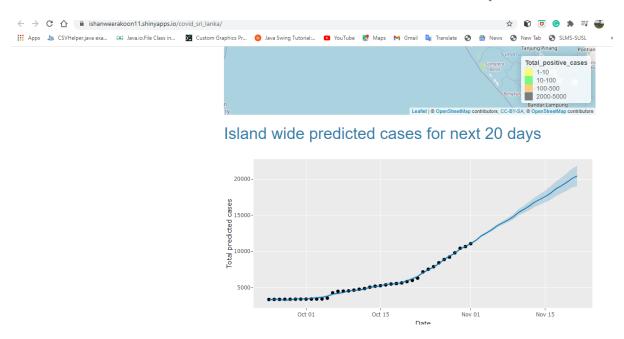
Analytics details → model_report-1.html

Data_analysis.html

Source code → covid_web_app.r



overseas cluster cases for next 20days



Susceptible, Exposed, Infected, Recovered model for COVID-19 Sri Lanka

S = Susceptible

I = Infective

R = Removed

Assumptions

- 1. Population remains constants
- 2. Rate of infective & contacts
- 3. Infective recover and die in constant rate

r = rate of contact

$$\frac{dS}{dt} = -rIS$$

$$\frac{dI}{dt} = rIS - aI$$

$$\frac{dR}{dt} = aI$$

Implementing these models, we can get these formulas

Here we wish to understand the dynamics of these categories through time t.

- Let Susceptible be denoted as S(t)
- Exposed be denoted as E(t)
- Infected be denoted as l(t)
- Recovered be denoted as R(t)
- Morality be denoted as D(t)

These are functions through time, hence the (t) notation.

Basically, these are "counts" of the number of people in each category.

For these simple models with two compartments S and I we can write the model as a system of two ordinary differential equation.

$$\frac{\partial S(t)}{\partial t} = -\partial S(t)t(t)$$

$$\frac{\partial l(t)}{\partial t} = -\partial S(t)l(t)$$

Let's do some math to make this simple!

$$\partial S(t) = \partial t \left(- \propto S(t) l(t) \right)$$

$$\partial l(t) = \partial t (\propto S(t)l(t))$$

Put in the fact that the ∂ means difference.

$$S(t) - S(t - \partial t) = \partial t (-\propto S(t)l(t))$$

$$l(t) - l(t - \partial t) = \partial t (\propto S(t)l(t))$$

Some more simplification:

$$S(t) - S(t - \partial t) = \partial t (-\propto S(t)l(t))$$

$$l(t) - l(t - \partial t) = \partial t (\propto S(t)l(t))$$

We for our application we know the difference in time ∂t is 1 day.

$$S(t) - S(t \sigma 1) = (-\infty S(t-1)l(t-1))$$

$$l(t) - l(t - 1) = (\propto S(t - 1)l(t - 1))$$

Isolate S(t) and l(t)

$$S(t) = S(t-1) - \propto S(t-1)l(t-1)$$

$$l(t) = l(t-1) + \propto S(t-1)l(t-1)$$

A little more complicated model is to add a recovered category.

$$\frac{\partial S(t)}{\partial t} = -\infty S(t)E(t)$$

$$\frac{\partial E(t)}{\partial t} = \propto S(t)E(t) - \beta E(t)$$

$$\frac{\partial I(\delta)}{\partial t} = \beta E(t) - \gamma R(t)$$

$$\frac{\partial R(t)}{\partial t} = \gamma I(t)$$

Which we will do the same process and turn it into:

$$S(t) = S(t-1) - \propto S(t-1)I(t-1)$$

$$E(t) = E(t-1) + \propto S(t-1)E(t-1) - \beta E(t)$$

$$I(t) = I(t-1) - \beta E(t) - \gamma I(t-1)$$

$$R(t) = R(t-1) - \gamma I(t-1)$$

According to these formulas we can implement model.

CODE→

#Suspectable, Infected and recovered Model

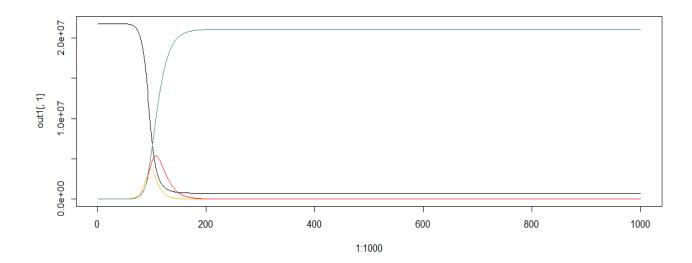
```
##sri lanka covid 19
```

lines(1:1000,out1[,3],col="red")

SO<- 21803000 #sri lanka population 10<- 1 #Initial infected RO<- 0 #intial recovers EO<- 10 #they have virus but they don't know about that #parameters alpha1<-1/100000000 alpha2<-1/100000000 beta1<-1/14 gamma1<-1/7 gamma2<-1/21 out1<-matrix(0,ncol = 4,nrow = 1000) for(i in 1:1000){ E0n<-E0 S0n<- S0 10n<- 10 R0n<- R0 SO<-max(0,SOn-alpha1*E0n*SOn-alpha2*SOn*IOn) EO<-max(0,E0n+alpha1*E0n*S0n+alpha2*S0n*I0n-gamma1*E0n-gamma2*E0n) IO<-max(0,I0n+gamma1*E0n-beta1*I0n) R0<-max(0,R0n+beta1*I0n+gamma2*E0n) out1[i,1] <- S0 out1[i,2] <- E0 out1[i,3] <- 10 out1[i,4] <- R0 } plot(1:1000,out1[,1],type="l",ylim=c(0,21803000)) lines(1:1000,out1[,2],col="orange")

lines(1:1000,out1[,4],col="seagreen")

Output→



Black line-many people are susceptible, and few people are infected, after 100 days it becomes large susceptive people come to low and infected people come to high.

In this programme beta1 variable represent how many people recover from the virus I am getting it as 1/14 every day

Green line-represent who are recovered people

Red line-who are the infected people

In this programme gamma1 variable number of people who are exposed to transition actually been sick

Orange line- exposed to virus and no go to sick they are recovered

In this programme gamma2 variable the come to recover group skipping infection group, they were no infection.

In this programme alpha2 variable some sick people actually interact with healthy people. quarantine valuable in this case. Add this variable we can the considerable difference in chart

Code file→SIR.r