Skeleton_of_Outputs

3/19/2020

This is a document for a code skeleton for the outputs section of the project.

Loading in relevant data sources for this:

```
#setwd('~/Dropbox/COVID-BaltimoreCity//')
require(socialmixr)
require(xlsx)
require(magrittr)
require(stringr)
require(reshape2)
require(dplyr)
require(ggplot2)
require(tidyr)
#load in data, subset incident cases/day (eventually should be symptomatic incident cases)
data_infected<- read.csv("SEIR_results_test.csv")</pre>
data_incident<- data_infected%>%
  select(incid1,incid2,incid3,incid4,incid5,incid6,incid7,incid8,incid9,incid10,incid11,
         incid12, incid13, incid14)
#consolidate age categories, 1-4 is now cat1, 5-6 is now cat2, 7-9 is now cat3, 10 is cat 4, 11 is cat5
data_incident<- data_incident%>%
  mutate(cat1=rowSums(.[1:4]))%>%
  mutate(cat2=rowSums(.[5:6]))%>%
 mutate(cat3=rowSums(.[7:9]))%>%
  mutate(cat4=rowSums(.[10]))%>%
  mutate(cat5=rowSums(.[11]))%>%
  mutate(cat6=rowSums(.[12]))%>%
  mutate(cat7=rowSums(.[13]))%>%
  mutate(cat8=rowSums(.[14]))
data_combined_cats<- data_incident%>%
  select(cat1,cat2,cat3,cat4,cat5,cat6,cat7,cat8)
#working probabilities for severe cases, hospitalization, ICU, vent and death.
#these are nested conditional probabilities, and they rely on the following assumptions:
#the percentage of severe cases in each categories is roughly an average of the percentage of people
#with diabetes, the percentage of people with cancer and the percentage of people with hypertension in
#we assume now that 100% of severe cases get hospitalized (this could be adjusted)
#we assume that globally, 24% of hospitalized cases get moved to ICU so this global percentage gets pro
#we assume that 25% of ICU cases are put on a vent again this gets projected down to age groups
# we assume that 50% of ventilated cases die, again this is projected down to age groups.
#HCW and homeless people have the same probabilities as people in category 4
prob_severe<- c(0.0318, 0.043, 0.197, 0.336,0.378,0.409,0.336,0.336)
```

Now calculate the number in of severe, hospitalized, ICU and vented infections at each timepoint (day) in each group (age groups 1-12 and healthcare workers and homeless).

```
#calculate number of incident cases that get classified as severe, of those,
#the number get hospitalized (currently 100%), of that then those that get classified as ICU and then o
#this is for each age group (+ HCW and homeless)
#note this categorization is NOT a disease progression, a person gets assigned based on the worst state
counts_array<- array(NA, dim=c(nrow(data_combined_cats),ncol(data_combined_cats),ncol(prob_matrix)))</pre>
for(j in 1:ncol(data combined cats)){
  for(i in 1:nrow(data_combined_cats)){
      counts_array[i,j,1]<- data_combined_cats[i,j]*prob_matrix[j,1]</pre>
      counts_array[i,j,2]<- counts_array[i,j,1]*prob_matrix[j,2]</pre>
      counts_array[i,j,3]<- counts_array[i,j,2]*prob_matrix[j,3]</pre>
      counts_array[i,j,4]<- counts_array[i,j,3]*prob_matrix[j,4]</pre>
  for(k in 1:nrow(data_combined_cats)){
    if(k<=11){
      counts_array[k,j,5]<- 0</pre>
    }
    else{
      counts_array[k,j,5]<- counts_array[k-11,j,4]*prob_matrix[j,5]</pre>
    }
  }
}
```

Now load in a dataset that gives estimates of how long people stay hospitalized, in ICU or on vents Compute a running total of each category.

```
#for now we are supposing we get a point estimate to determine how long people stay in each of the stat
length_of_stay<- c(NA,11,11,11,NA)

cumulative_array<- array(NA, dim=c(nrow(data_combined_cats),ncol(data_combined_cats),ncol(prob_matrix))

for(j in 1:ncol(data_combined_cats)){
   cumulative_array[1,j,]<- counts_array[1,j,]

for(i in 2:nrow(data_combined_cats)){
   cumulative_array[i,j,1]<-cumulative_array[i-1,j,1]+counts_array[i,j,1]</pre>
```

```
cumulative_array[i,j,5]<-cumulative_array[i-1,j,5]+counts_array[i,j,5]</pre>
    }
    for(k in 2:nrow(data_combined_cats)){
      if(k<=length_of_stay[2]){</pre>
      cumulative_array[k,j,2] <- cumulative_array[k-1,j,2] +counts_array[k,j,2]
      }
      else {
        cumulative\_array[k,j,2] < -cumulative\_array[k-1,j,2] +
          counts_array[k,j,2]-counts_array[k-length_of_stay[2],j,2]
      }
    }
  for(1 in 2:nrow(data_combined_cats)){
      if(1 <= length_of_stay[3]){</pre>
        cumulative_array[1,j,3]<- cumulative_array[1-1,j,3]+counts_array[1,j,3]
      }
    else{
      cumulative_array[1,j,3]<-cumulative_array[1-1,j,3]+</pre>
          counts_array[1,j,3]-counts_array[1-length_of_stay[3],j,3]
    }
  }
    for(m in 2:nrow(data_combined_cats)){
    if(m<=length of stay[4]){</pre>
      cumulative_array[m,j,4]<- cumulative_array[m-1,j,4]+counts_array[m,j,4]</pre>
    }
    else{
      cumulative_array[m,j,4]<-cumulative_array[m-1,j,4]+</pre>
          counts_array[m,j,4]-counts_array[m-length_of_stay[4],j,4]
      }
    }
}
cumulative_array[,2,3]
##
     [1]
           0.18705
                     0.41280
                                0.76755
                                          1.27065
                                                    1.85115
                                                               2.67675
                                                                         3.76680
           5.26965
                     7.30785 10.07490 13.46115 18.04710 23.76825
##
     [8]
                                                                        30.86325
   [15] 39.99645 50.77440 63.90660 79.05765 96.18885 116.06775 138.06225
##
  [22] 161.54025 187.06290 212.42430 237.40515 261.64425 284.54820 304.65930
##
  [29] 321.96465 336.46425 346.95195 353.26650 354.36945 352.00230 346.43595
##
   [36] 337.70265 326.33130 311.78655 296.60970 280.26540 263.51475 245.61600
  [43] 227.79465 212.19855 195.39630 180.07110 165.71340 151.53630 139.57155
##
                                                                        70.96935
## [50] 127.23915 116.25480 105.15435 95.90505 87.09435 78.59970
##
   [57] 63.79695 57.28890 51.83220
                                         46.54965
                                                   41.87985
                                                              37.53255
                                                                        33.88185
##
   [64] 30.23115 27.17385
                              24.40035
                                         21.97515
                                                   19.76925
                                                              17.76330
                                                                        15.67995
##
  [71]
         13.80945
                   12.51300
                              11.26815
                                        10.19745
                                                    9.17835
                                                              8.20440
                                                                         7.57230
                                                                         3.77325
  [78]
                     6.09525
                               5.56635
                                         5.02455
                                                    4.60530
                                                               4.27635
##
           6.88215
##
   [85]
           3.40560
                     2.95410
                               2.64450
                                          2.36070
                                                    2.06400
                                                               1.87050
                                                                         1.67055
##
    [92]
           1.49640
                     1.37385
                               1.17390
                                         1.12230
                                                    0.96750
                                                               0.89010
                                                                         0.82560
    [99]
           0.69015
```

Plot cumulative values over 99 days.

```
#return to long format for each state
#severe

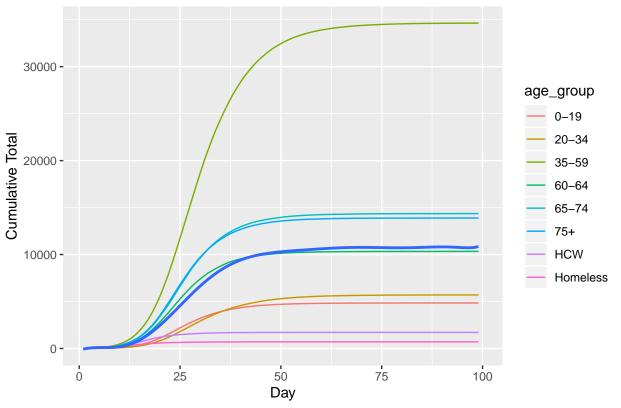
severe_wide<-as.data.frame(cbind(rep(1:100),cumulative_array[,,1]))
colnames(severe_wide)<- c("Day","0-19","20-34","35-59","60-64","65-74","75+","HCW","Homeless")

severe_long<- gather(severe_wide,age_group, cumulative_total, '0-19':Homeless,factor_key=TRUE)

plt_severe<- ggplot(severe_long)+
    geom_line(aes(y= cumulative_total, x=Day,color=age_group))+
    stat_smooth(aes(y= cumulative_total,x=Day), method=lm, formula=y~poly(x,10), se=FALSE)+
    labs(title="Trajectory of Severe Cases",y="Cumulative Total")

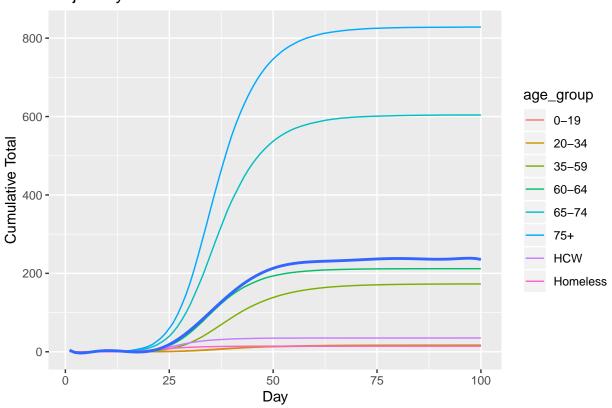
plt_severe</pre>
```

Trajectory of Severe Cases

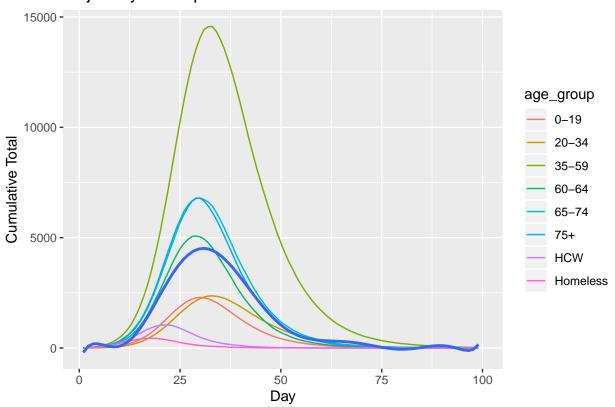


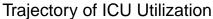
```
stat_smooth(aes(y= cumulative_total,x=Day), method=lm, formula=y~poly(x,10), se=FALSE)+
labs(title="Trajectory of Deaths",y="Cumulative Total")
plt_deaths
```

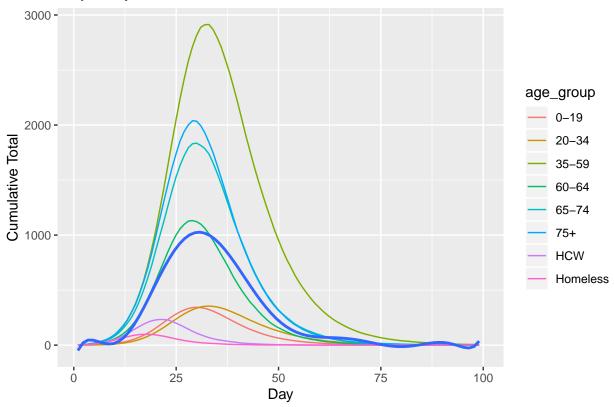
Trajectory of Deaths



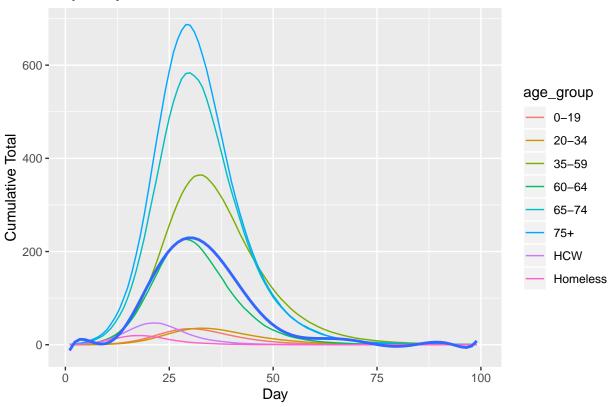




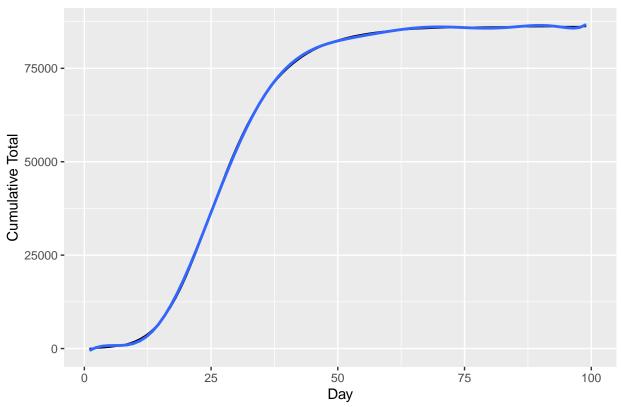


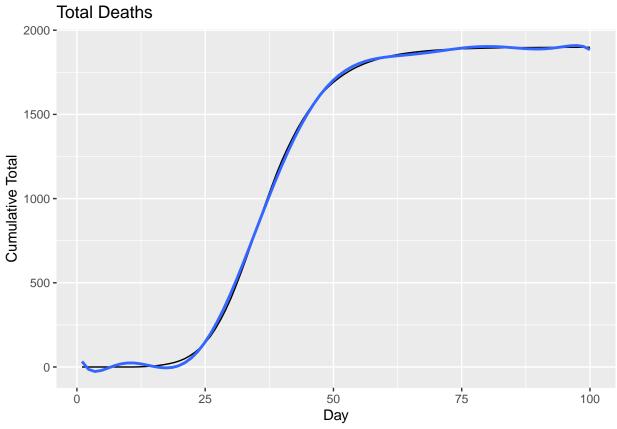


Trajectory of Vent Utilization

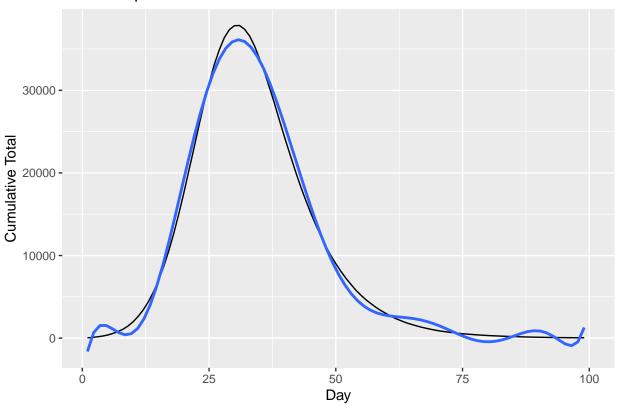


Total Severe Cases

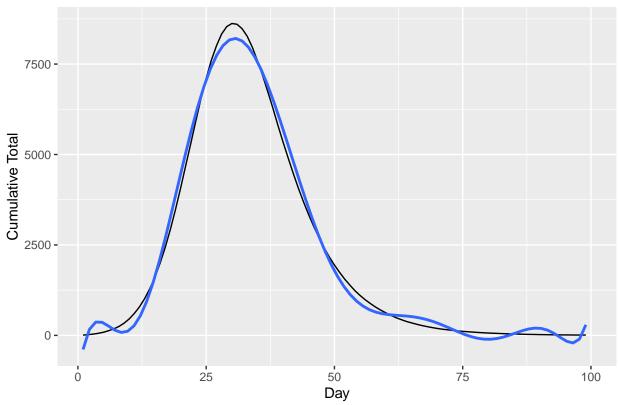


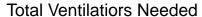


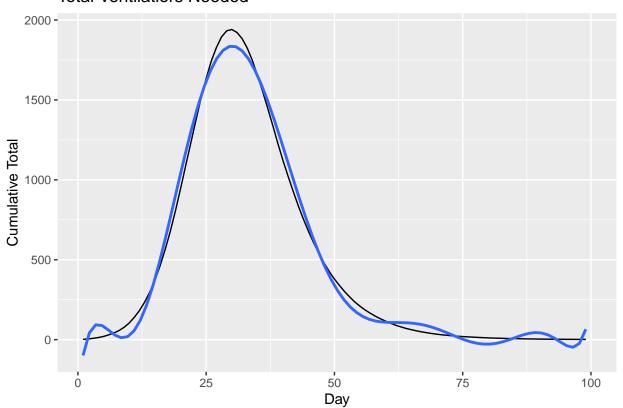
Total Hospitalizations



Total ICU Beds Needed







#number of severe cases total
cumulative_totals[99,2]

[1] 86185.61

#number of deaths total
cumulative_totals[100,6]

[1] 1898.571

#max number of hospital beds needed in one day (including ICU beds in this total)
max(cumulative_totals[,3],na.rm=TRUE)

[1] 37850.58

#max number of ICU beds needed in one day (including people who might also need a vent)
max(cumulative_totals[,4],na.rm=TRUE)

[1] 8621.777

#max number of vents needed in one day
max(cumulative_totals[,5],na.rm=TRUE)

[1] 1940.527