Applied Stochastic Process

Assignment 2 - Solutions

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An epidemic caused by a new virus is spreading in City A. You are asked to evaluate the effect of several strategies in controlling the epidemic. Suppose the basic reproduction number (R_0) is estimated to be 2.3 (new cases from each old case per week). Three strategies are proposed to deal with the situation. The effective reproduction numbers (R_e) in the presence of control efforts are predicted to be 0.8, 1.2 and 1.8 respectively. Model the epidemic as a branching process and answer the following questions by means of simulation.

To simulate the branching process, we first set seed and assign values to the parameters.

The function for simulating a branching process is defined below:

```
while (t <= T1 & NPath0[t]!=0) {</pre>
    \#NPath0[t+1] \leftarrow sum(rgeom(NPath0[t], p0))
    NPath0[t+1] <- rnbinom(1, NPath0[t], p0)</pre>
  }
  #Terminate if NPathO[t] == 0
  if (NPath0[t]==0) {
    NPath0[t:(T1+1)] <- 0</pre>
    NPath1[1:(T1+1), ] <- NPath0</pre>
    NPath1[(T1+2):(T1+T2+1), ] < 0
    return(NPath1)
  }
  #######
  #After controlling policies
  for (iR in 1:length(Re)) {
    NPath1[1:(T1+1), iR] <- NPath0</pre>
    t <- 1
    while (t <= T2 & NPath1[t+T1, iR]!=0) {
       \#NPath1[t+T1+1, iR] \leftarrow sum(rqeom(NPath1[t+T1, iR], pe[iR]))
      NPath1[t+T1+1, iR] <- rnbinom(1, NPath1[t+T1, iR], pe[iR])</pre>
      t <- t+1
    }
    if (NPath1[t+T1, iR]==0) {
      NPath1[(t+T1):(T1+T2+1), iR] < -0
    }
  }
  return(NPath1)
}
```

Assuming that the number of new cases from each old case Z follows a geometric distribution with probability of success p. Note that here we assume Z to be the number of failures before the first success (the definition in R) instead of the number of trials until the first success (the definition in lecture notes), i.e., $p(Z) = p(1-p)^Z$. Note that R_0 and R_e are the expected values of a geometrically distributed random variable, we can deduce from the probability of success by the relation $p_0 = (R_0 + 1)^{-1}$ and respectively for p_e .

With X_{n-1} cases in total at time n-1, the number of new cases at time n is

$$X_n = \sum_{i=1}^{X_{n-1}} Z_{n,i}.$$

Therefore, we can simply calculate X_n as the sum of X_{n-1} geometrically distributed random variable, which can be done by the code NPath0[t+1] <- sum(rgeom(NPath0[t], p0)). However, if the process explodes, rgeom(NPath0[t], p0) can run very slowly. Therefore, we can use the fact that independent sum of geometrically distributed random variables is negative binomially distributed, and write NPath0[t+1] <- rnbinom(1, NPath0[t], p0).

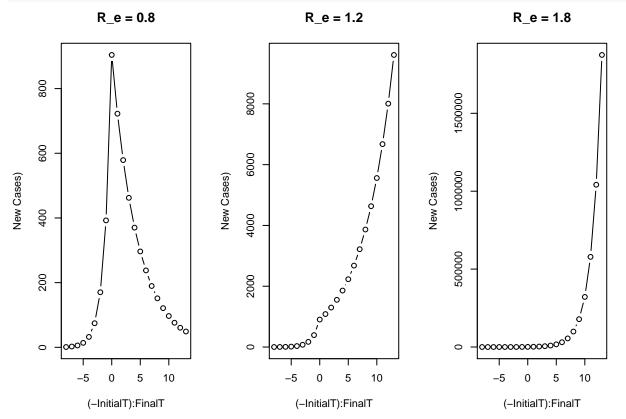
Finally, the number of new cases before and after the control policies can be obtained by running the above defined function.

```
Ncases <- replicate(NTrial, branching(RO, Re, InitialT, FinalT))</pre>
```

1. Suppose the epidemic has already been left to grow without intervention for 8 weeks since the first case. What is the mean and variance number of new cases this week?

[1] "The mean number of new cases today is 392.466"

- ## [1] "The variance of new cases today is 361283.319482966"
- 2. For each strategy, plot the mean number of new cases after 1-52 weeks.



3. For each strategy, what is the probability of zero new cases after 52 weeks?

Since the output of rnbinom() can be NA if NPathO[t] is too large, we only simulate up to 13 weeks after time 0 here.

- ## [1] "The probability of zero new cases under strategy 1 is: 5.38720538720539 %."
- ## [1] "The probability of zero new cases under strategy 2 is: 0.673400673400673 %."
- ## [1] "The probability of zero new cases under strategy 3 is: 0.336700336700337 %."