

Brief Analysis of Norovirus Data

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June 6, 2019

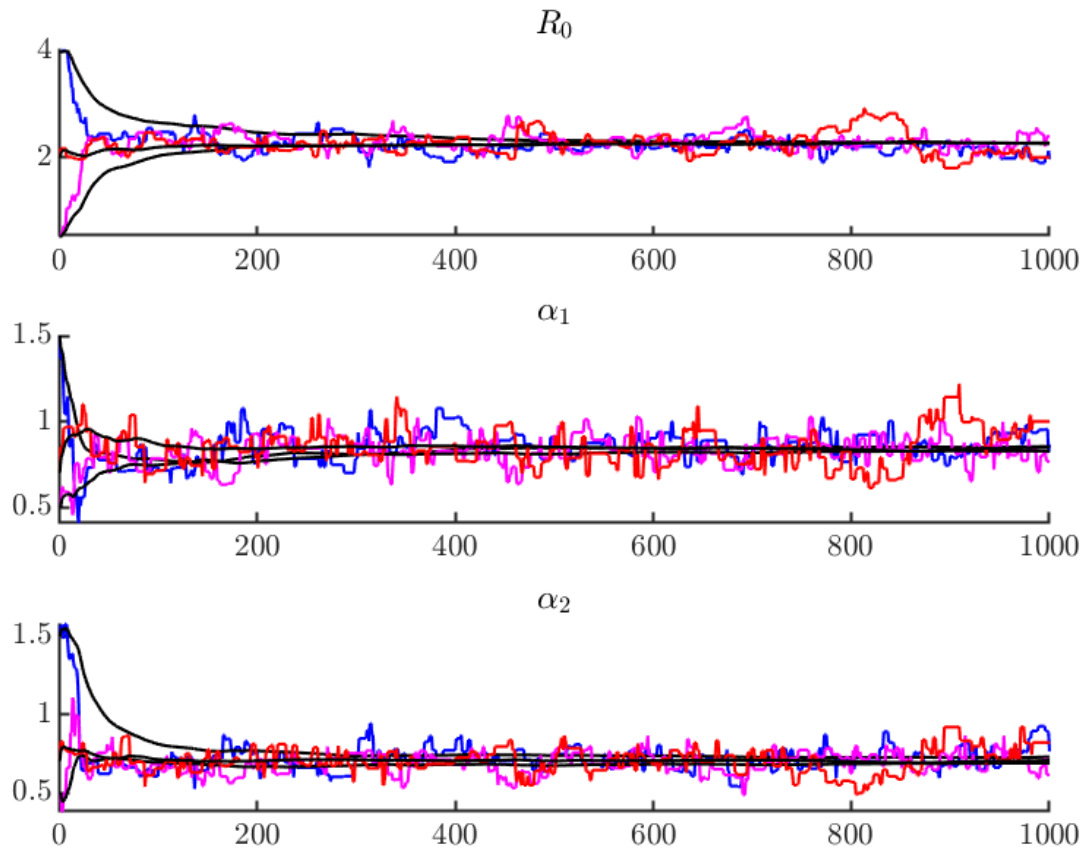


Figure 1: Caption here

Advise the government on effectiveness of interventions, and which should be implemented (if any).

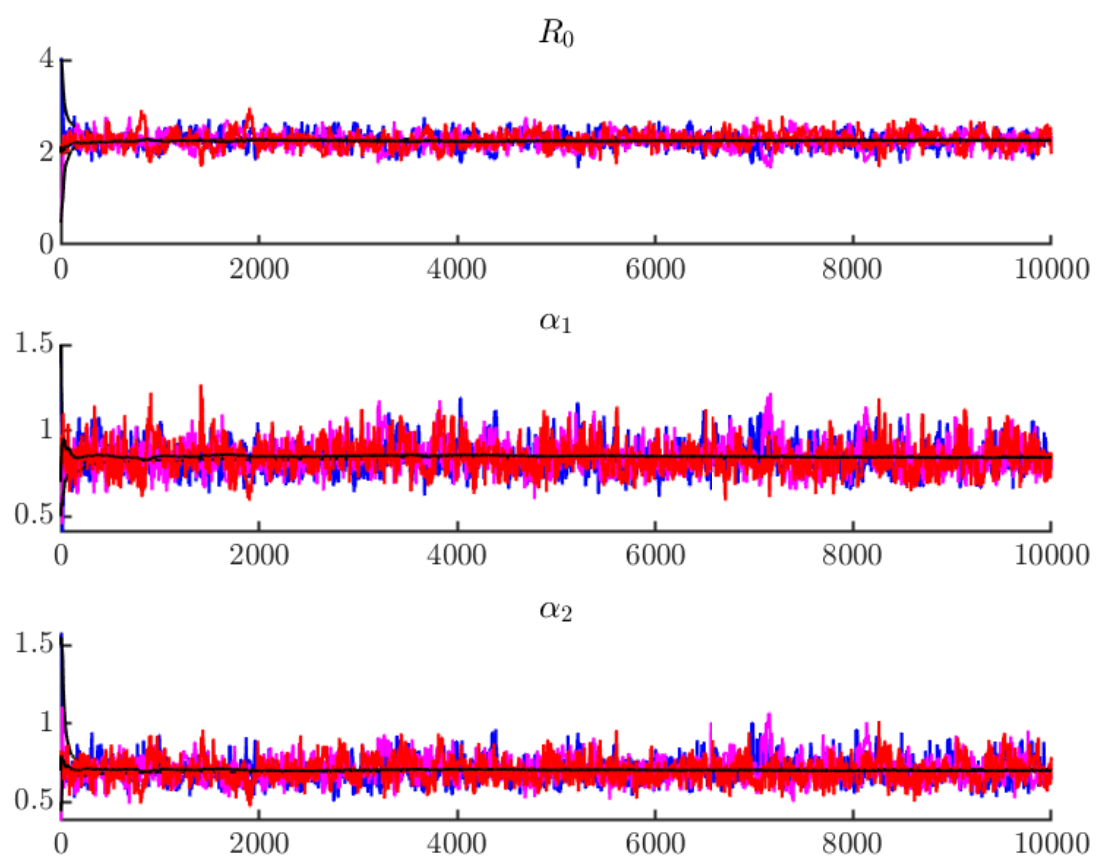


Figure 2: Caption here

.1 Code

.1.1 Main Script (R0Predict.m)

```
1 %Pretty plots
2 set(groot, 'DefaultLineLineWidth', 1, ...
3     'DefaultAxesLineWidth', 1, ...
4     'DefaultAxesFontSize', 12, ...
5     'DefaultTextFontSize', 12, ...
6     'DefaultTextInterpreter', 'latex', ...
7     'DefaultLegendInterpreter', 'latex', ...
8     'DefaultColorbarTickLabelInterpreter', 'latex', ...
9     'DefaultAxesTickLabelInterpreter', 'latex');
10
11
12
13 %Speeds up the code for multiple runs
14 if ~exist('dataMat','var')
15     dataMat = readmatrix('NorovirusDataA3.txt');
16     %print the first few rows
17     dataMat(1:5,:)
18 end
19
20 %Observation of dataset
21 %how many of them correspond to the disease effectively dying out?
22 amountOfData = length(dataMat(:,1))
23 propDiedOut = sum(dataMat(:,2) <= 0.1 * dataMat(:,1)) / amountOfData
24
25 %dataMat = [#people, #infected, #action]
26 %consistency
27 rng(1)
28
29
30
31 %since we are expecting approximately 66%
32 %getting the right prior
33 %we want the dist to be N'(2.5, sigma)
34 %where sigma is the variance to have approx 66% in (2-3).
35 %N' since we will drop values below 0 and above 5
36 %N' will still be symmetric
37 sigma = -(3-2.5)/norminv((1-0.66)/2);
38 variance = sigma * speye(3);
39
40 varR0 = 0.01;
41 distStruct.ProposalDist = @(x) mvnrnd(x, diag([varR0, 0.01, 0.01]));
42 distStruct.PriorPDF = @(x) normpdf(x, 2.5, sigma);
43 distStruct.LogLikelihoodFunc = @(params, data) LogLikelihood(params, data);
44 %constrain a_1, a_2 in [0, 2]
45 distStruct.ProposalConstraints = @(vars) ProposalConstraints(vars, [0, 0, 0], [5, 5, 5]);
46 %overshoot, undershoot and approximately close
```

```

47 startVars = [4,1.5,1.5;
48             0.5,0.5,0.5;
49             2,0.7,0.7];
50
51 vars = zeros(10000,3,3);
52 for index = 1:3
53     startPoint = startVars(index,:)
54     exitflag = 1;
55     while exitflag~=0
56         [varsTemp,accrate,exitflag]= MetropolisHastingsPassLikelihood(distStr
57         if exitflag==−1
58             %acceptance was too low
59             %variance is too high
60             varR0 = varR0 *2/3;
61             distStruct.ProposalDist = @(x) mvnrnd(x,diag([varR0,0.01,0.01]));
62             warning('retrying with decreased variance')
63         else if exitflag==1
64             %acceptance was too high
65             %variance is too low
66             varR0 = varR0*2;
67             distStruct.ProposalDist = @(x) mvnrnd(x,diag([varR0,0.01,0.01]));
68             warning('retrying with decreased variance')
69         end
70     end
71     vars(:, :, index) = varsTemp;
72 end
73
74
75 end
76 %%
77 cumVars = cumsum(vars,1)./(1:length(vars))';
78 tstr = ["$$R_0$$","$$\alpha_1$$","$$\alpha_2$$"];
79 for burnin = 0:1
80     figure
81     for i=1:3
82         hold on
83         subplot(3,1,i)
84         hold on
85         plot(vars(:,i,1),'b')
86         plot(cumVars(:,i,1),'k')
87         plot(vars(:,i,2),'m')
88         plot(cumVars(:,i,2),'k')
89         plot(vars(:,i,3),'r')
90         plot(cumVars(:,i,3),'k')
91         title(tstr(i))
92
93         if burnin
94             axis([0,1000,−inf,inf])
95         end
96     end

```

```

97         saveas(gcf,"MHplot"+num2str(burnin)+".eps","epsc")
98     end
99     axis([0,1000,-inf,inf])
100    xlabel('Iteration')
101
102    %%
103
104    %%Clean new data
105    %just take one set since they all converged
106    %and omit the burnin
107    varsClean = vars(500:end,:,3);
108
109    close all
110    %density plot
111    %and obtain estimates for R0, a1, a2
112
113    labs = ["$$R_0$$","$$\alpha_1$$","$$\alpha_2$$"];
114
115    est = [0,0,0];
116    for i=1:3
117        figure
118        [prob,val] = ksdensity(varsClean(:,i));
119        prob = prob./sum(prob);
120        plot(val,prob)
121        xlabel(labs(i))
122        ylabel("Probability")
123        title("Probability density for "+labs(i))
124        [~,ind] =max(prob);
125        %assuming there is no dependence
126        est(i) = val(ind);
127        saveas(gcf,"Probdensity"+num2str(i),"epsc")
128    end
129    est
130    est(2:3) * est(1)
131
132    for i=1:2
133        for j=i+1:3
134            figure
135            binscatter(varsClean(:,i),varsClean(:,j),60,'HandleVisibility','off')
136            hold on
137            scatter(est(i),est(j),'xr')
138            xlabel(labs(i))
139            ylabel(labs(j))
140            legend("Independent approximation")
141            colormap(gca,'parula')
142            saveas(gcf,"BinScatter"+num2str(i)+num2str(j),"epsc")
143        end
144    end

```

1.2 MetropolisHastingsPassLikelihood.m

```

1  function [vars, accRate, exitflag] = MetropolisHastingsPassLikelihood(distStruct, data, numIterations, startVars);
2  %Most generic MetropolisHastings
3  %distStruct has fields
4  %-priorDist - the prior to pull from
5  %-proposalDist - the proposed distribution
6
7  numAccepted = 0;
8  PriorPDF = distStruct.PriorPDF;
9  ProposalDist = distStruct.ProposalDist;
10 BreaksProposalConstraints = distStruct.ProposalConstraints;
11 LogLikelihoodFunc = distStruct.LogLikelihoodFunc;
12 vars = zeros(numIterations, length(startVars));
13 vars(1,:) = startVars;
14
15 for i=2:numIterations
16     proposal = ProposalDist(vars(i-1,:));
17     %if breaks constraints
18     if BreaksProposalConstraints(proposal)
19         vars(i) = vars(i-1);
20     else
21         candidateProbTop = LogLikelihoodFunc(proposal, data) + log(PriorPDF(proposal));
22         candidateProbBottom = LogLikelihoodFunc(vars(i-1,:), data) + log(PriorPDF(vars(i-1,:)));
23         candidateProb = candidateProbTop - candidateProbBottom;
24         acceptProb = log(rand);
25
26         if acceptProb < candidateProb
27             vars(i,:) = proposal;
28             numAccepted = numAccepted + 1;
29         else
30             vars(i,:) = vars(i-1,:);
31         end
32     end
33 end
34 accRate = numAccepted/numIterations;
35
36 if accRate < 0.2
37     warning("Bad acceptance rate - too low, accRate = "+num2str(accRate));
38     exitflag = -1;
39 else if accRate > 0.27
40     exitflag = 1;
41     warning("Bad acceptance rate - too high, accRate = "+num2str(accRate));
42 else
43     exitflag = 0;
44 end
45 end
46
47
48 end

```

.1.3 LogLikelihood.m

```

1 function logLikelihood = LogLikelihood(params,data)
2 %Amended SIR finalsize code from Josh
3 %Amended by Andrew Martin
4 %calculates log likelihood for a given R0,alpha1,alpha2
5 logLikelihood = 0;
6 % R0      = params(1);
7 % alpha1  = params(2);
8 % alpha2  = params(3);
9 NVec = data(:,1);
10 %[R0, a1R0, a2R0]
11 paramsModified = [params(1),params(1)*params(2),params(1)*params(3)];
12 %yay matlab uses 1 based indexing
13 R0index = data(:,3)+1;
14
15
16 for iterator=1:length(data)
17     N = NVec(iterator);
18     relevantParam = paramsModified(R0index(iterator));
19     q = zeros(N+1,1);
20     q(2) = 1;
21     %Proportions for each number of infection events
22     %could vectorise, but this is more meaningful
23     for Z2 = 0:N
24         for Z1 = Z2+1:N-1
25             %infection probability (jump prob)
26             infProb = 1 / ( 1 + ((N-1)/(relevantParam*(N-Z1))) );
27             q(Z1+2) = q(Z1+2) + q(Z1+1)*infProb;
28             q(Z1+1) = q(Z1+1)*(1-infProb);
29         end
30     end
31     %sum of the log likelihoods (product of likelihoods)
32     logLikelihood= logLikelihood + log(q(data(iterator,2)+1));
33 end
34
35
36 end

```

.1.4 ProposalConstraints.m

```

1 function boolean = ProposalConstraints(vals,min,max)
2 %
3 %OUTPUT:
4 %boolean - true if the boundary constraints are broken
5 boolean = any(vals < min | vals > max);
6
7
8 % %hardcoded way
9 % boolean =vals(1) < 0 || vals(1) > 5 ...
10 %         || vals(2) < min || vals(2) > max ...
11 %         || vals(3) < min || vals(3) > max;
12

```


Assignment III

Worth 15% of course assessment; due by 3pm on Friday 7th June, 2019.

Most-relevant lectures: Lectures 21 – 26.

The total marks for this assignment is 45.

Please provide code where appropriate.

Report to the Government on the effectiveness of interventions.

[45 marks]

The CSV file `NorovirusDataA3` contains information regarding independent outbreaks in a set of 125 hospital wards of varying sizes. The first column contains the number of occupied beds in the ward, the second column contains the number of those patients which succumbed to norovirus during the outbreak, and the third column indicates the control action implemented; in the last column, 0 corresponds to standard practices, 1 corresponds to a trial intervention strategy, and 2 corresponds to different trial intervention strategy.

You are to analyse this data to advise the Government on the effectiveness of their interventions, and to advise them on which intervention should be adopted (if any).

You may assume that the interventions work to reduce the effective transmission rate parameter. The Chief Medical Officer has said that R_0 for norovirus in typical hospital settings is between 2 and 3 with (approximately) 66% probability; you may use this expert opinion as prior knowledge.

You are to prepare two reports. One is for the Government. The second is to provide detail on how you have performed the statistical analysis, including the model(s) used, any assumptions you have made and providing evidence that your approach/algorithms are working correctly, for example through the use of simulated data of similar form to the ‘real’ data (e.g., using trace plots from multiple independent chains, and kernel density estimators and box plots). This assignment is deliberately vague – you need to make decisions, but feel free to ask for feedback as you make progress.