

This Maple worksheet is denoted sirs3c.

It is used to study an SIRS model with demography, for $R_0 > 1$ and α_1 large.

The deterministic version of the model then shows damped oscillations toward an endemic infection level.

The particular model dealt with here has a state space with 3 variables: S, I, and R, and the infection rate is "conventional", with N in the denominator.

Two things are done here.

First we derive an expression for the angular frequency of the deterministic model oscillations, and after that we give a derivation of the moments of a diffusion approximation.

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```
> restart;
with(LinearAlgebra,Transpose,Eigenvalues,
CharacteristicPolynomial);
with(VectorCalculus,Jacobian);
interface(imaginaryunit=II);
      [Transpose,Eigenvalues,CharacteristicPolynomial]
      [Jacobian]
      /
```

(1)

The reason for changing the notation used for the imaginary unit is that "I" will be used below to denote the number of infected individuals.

The original transition rates are stored in the table transA:

```
> transA:=table([[1,0,0]=mu*N,[1,0,-1]=delta*R,[-1,1,0]=beta*S*
I/N,[-1,0,0]=mu*S,[0,-1,1]=gamma*I,[0,-1,0]=mu*I,[0,0,-1]=mu*R]
);
```

After scaling: $x_1 = S/N$, $x_2 = I/N$, $x_3 = R/N$, and reparametrization $R_0 = \beta/(\gamma + \mu)$, $\alpha_1 = (\gamma + \mu)/\mu$, $\alpha_2 = (\delta + \mu)/\mu$, we get: $S = x_1 * N$, $I = x_2 * N$, $R = x_3 * N$, $\beta = \mu * \alpha_1 * R_0$, $\gamma = \mu * (\alpha_1 - 1)$, $\delta = \mu * (\alpha_2 - 1)$.

The Maple procedure "scale" is used to change the transition rates accordingly:

```
> scale:=proc(tab)
  local xA,n,xB,xC;
  xA:=op(2,eval(tab));
  n:=nops(xA);
  xB:=subs(S=x1*N,I=x2*N,R=x3*N,beta=mu*alpha1*R0,gamma=mu*
(alpha1-1),delta=mu*(alpha2-1),xA);
  xC:=[seq(lhs(op(i,xB))=simplify(rhs(op(i,xB)/N)),i=1..n)];
  table(xC);
end proc;
```

Apply the scaling and reparametrization described above to get the table of transition rates "trans":

```
> trans:=scale(transA);
trans:=table([[0,0,-1]=mu*x3,[0,-1,0]=mu*x2,[0,-1,1]=mu*(alpha1-1)*x2,[-1,0,0]=mu*x1,[-1,1,0]=mu*alpha1*R0*x1*x2,[1,0,-1]=mu*(alpha2-1)*x3,[1,0,0]=mu])
```

(2)

Next is a procedure that determines the right-hand sides of the deterministic ODEs

for the scaled variables x_1, x_2, x_3 from the table of transition rates:

```
> equ:=proc(i,tab)
  local x,n;
  x:=op(2,eval(tab));
  add(lhs(x[n])[i]*rhs(x[n]),n=1..nops(x));
end proc;
```

The 3 right-hand sides are as follows:

```
> eq1:=equ(1,trans);
eq2:=simplify(equ(2,trans));
eq3:=simplify(equ(3,trans));
```

$$\begin{aligned} eq1 &:= -\mu x_1 - \mu \alpha_1 R_0 x_1 x_2 + \mu (\alpha_2 - 1) x_3 + \mu \\ eq2 &:= -\mu x_2 \alpha_1 + \mu \alpha_1 R_0 x_1 x_2 \\ eq3 &:= \mu x_2 \alpha_1 - \mu x_2 - \mu x_3 \alpha_2 \end{aligned} \quad (3)$$

The critical points:

```
> crit:=solve({eq1,eq2,eq3},{x1,x2,x3});
```

$$\begin{aligned} crit &:= \{x_1 = 1, x_2 = 0, x_3 = 0\}, \left\{ x_1 = \frac{1}{R_0}, x_2 = \frac{\alpha_2 (-1 + R_0)}{R_0 (\alpha_2 + \alpha_1 - 1)}, x_3 \right. \\ &= \left. \frac{(-1 + R_0) (\alpha_1 - 1)}{R_0 (\alpha_2 + \alpha_1 - 1)} \right\} \end{aligned} \quad (4)$$

The point corresponding to an endemic infection level is termed (x_{10}, x_{20}, x_{30}) :

```
> x10:=rhs(crit[2][1]);
x20:=rhs(crit[2][2]);
x30:=rhs(crit[2][3]);
```

$$\begin{aligned} x_{10} &:= \frac{1}{R_0} \\ x_{20} &:= \frac{\alpha_2 (-1 + R_0)}{R_0 (\alpha_2 + \alpha_1 - 1)} \\ x_{30} &:= \frac{(-1 + R_0) (\alpha_1 - 1)}{R_0 (\alpha_2 + \alpha_1 - 1)} \end{aligned} \quad (5)$$

The Jacobian of the system of ODEs is denoted B_x :

```
> Bx:=Jacobian([eq1,eq2,eq3],[x1,x2,x3]);
```

$$B_x := \begin{bmatrix} -\mu - \mu x_2 \alpha_1 R_0 & -\mu \alpha_1 R_0 x_1 & \mu (\alpha_2 - 1) \\ \mu x_2 \alpha_1 R_0 & -\alpha_1 \mu + \mu \alpha_1 R_0 x_1 & 0 \\ 0 & \alpha_1 \mu - \mu & -\mu \alpha_2 \end{bmatrix} \quad (6)$$

Evaluate the Jacobian at the critical point:

```
> B:=simplify(subs(x1=x10,x2=x20,x3=x30,Bx));
```

(7)

$$B := \begin{bmatrix} -\frac{\mu(\alpha_2 + \alpha_1 - 1 - \alpha_1 \alpha_2 + \alpha_1 \alpha_2 R_0)}{\alpha_2 + \alpha_1 - 1} & -\alpha_1 \mu & \mu(\alpha_2 - 1) \\ \frac{\mu \alpha_2 (-1 + R_0) \alpha_1}{\alpha_2 + \alpha_1 - 1} & 0 & 0 \\ 0 & \alpha_1 \mu - \mu & -\mu \alpha_2 \end{bmatrix} \quad (7)$$

We proceed to determine the eigenvalues of the matrix B.

We use first the command "Eigenvalues" to show that one of the eigenvalues equals $-\mu$.

After that, we use the command "CharacteristicPolynomial" and the knowledge that one eigenvalue equals $-\mu$ to derive a quadratic equation for the two remaining eigenvalues.

```
> eig:=Eigenvalues(B);
```

$$eig := \left[\left[-\mu \right], \right] \quad (8)$$

$$\left[\frac{1}{2} \frac{1}{\alpha_2 + \alpha_1 - 1} \left(\left(-\alpha_2^2 - \alpha_1 \alpha_2 R_0 + \alpha_2 \right. \right. \right. \\ \left. \left. \left. + (\alpha_2^4 - 2 \alpha_1 \alpha_2^3 R_0 - 2 \alpha_2^3 + \alpha_1^2 \alpha_2^2 R_0^2 + 6 \alpha_1 \alpha_2^2 R_0 + \alpha_2^2 \right. \right. \right. \\ \left. \left. \left. - 8 \alpha_2^2 \alpha_1 + 4 \alpha_2^3 \alpha_1 + 8 \alpha_2^2 \alpha_1^2 - 8 \alpha_1^2 \alpha_2^2 R_0 - 8 \alpha_2 \alpha_1^2 + 4 \alpha_2 \alpha_1^3 \right. \right. \right. \\ \left. \left. \left. - 4 \alpha_2 \alpha_1^3 R_0 + 8 \alpha_1^2 \alpha_2 R_0 + 4 \alpha_1 \alpha_2 - 4 \alpha_1 \alpha_2 R_0 \right)^{1/2} \right) \mu \right], \\ \left[-\frac{1}{2} \frac{1}{\alpha_2 + \alpha_1 - 1} \left(\left(\alpha_2^2 + \alpha_1 \alpha_2 R_0 - \alpha_2 \right. \right. \right. \\ \left. \left. \left. + (\alpha_2^4 - 2 \alpha_1 \alpha_2^3 R_0 - 2 \alpha_2^3 + \alpha_1^2 \alpha_2^2 R_0^2 + 6 \alpha_1 \alpha_2^2 R_0 + \alpha_2^2 \right. \right. \right. \\ \left. \left. \left. - 8 \alpha_2^2 \alpha_1 + 4 \alpha_2^3 \alpha_1 + 8 \alpha_2^2 \alpha_1^2 - 8 \alpha_1^2 \alpha_2^2 R_0 - 8 \alpha_2 \alpha_1^2 + 4 \alpha_2 \alpha_1^3 \right. \right. \right. \\ \left. \left. \left. - 4 \alpha_2 \alpha_1^3 R_0 + 8 \alpha_1^2 \alpha_2 R_0 + 4 \alpha_1 \alpha_2 - 4 \alpha_1 \alpha_2 R_0 \right)^{1/2} \right) \mu \right] \right]$$

One of the eigenvalues is thus seen to be equal to $-\mu$.

Next we determine the characteristic polynomial p:

```
> p:=CharacteristicPolynomial(B,lambda);
```

$$p := \lambda^3 + \frac{\mu(\alpha_2^2 + \alpha_1 - 1 + \alpha_1 \alpha_2 R_0) \lambda^2}{\alpha_2 + \alpha_1 - 1} \quad (9) \\ + \frac{\alpha_2 \mu^2 (\alpha_2 + \alpha_1 - 1 - \alpha_1 \alpha_2 + \alpha_1 \alpha_2 R_0 - \alpha_1^2 + \alpha_1^2 R_0) \lambda}{\alpha_2 + \alpha_1 - 1} + \mu^3 \alpha_2 (-1 \\ + R_0) \alpha_1$$

To proceed, we derive a quadratic equation for the remaining two eigenvalues by dividing p by $\lambda + \mu$, and simplifying:

```
> p1:=map(simplify,collect(simplify(p/(lambda+mu)),lambda));
```

$$p1 := \lambda^2 + \frac{\mu \alpha_2 (\alpha_2 + \alpha_1 R_0 - 1) \lambda}{\alpha_2 + \alpha_1 - 1} + \alpha_1 (-1 + R_0) \alpha_2 \mu^2 \quad (10)$$

The two roots of $p_1=0$, where $p_1 = \lambda^2 + a \cdot \lambda + b$, can be written $-a/2 \pm i \cdot \Omega$, with Ω

$$= \sqrt{b - \left(\frac{a}{2}\right)^2} = \mu \cdot \sqrt{\alpha_1 \cdot \alpha_2 \cdot (R_0 - 1) - \left(\frac{\alpha_2 \cdot R_1}{2}\right)^2}, \text{ where } R_1 = \frac{\alpha_1 \cdot R_0 + \alpha_2 - 1}{\alpha_1 + \alpha_2 - 1}.$$

This concludes the derivation of an expression for the angular frequency.

Next we turn toward a derivation of a diffusion approximation.

Covariances of $x[i]x[j]$ are determined by cov1:

```
> cov1:=proc(i,j,tab)
  local x,n;
  x:=op(2,eval(tab));
  add(lhs(x[n])[i]*lhs(x[n])[j]*rhs(x[n]),n=1..nops(x));
end proc;
```

The local covariance matrix S is determined by the procedure cov:

```
> cov:=proc(tab)
  local i,j,d,S;
  d:=nops(lhs(op(2,eval(tab))[1]));
  for i from 1 to d do
    for j from 1 to d do
      S[i,j]:=cov1(i,j,tab);
    od;
  od;
  S:=Matrix(d,S);
end proc;
```

By using the table of transition rates in "trans", we get

```
> Sx:=simplify(cov(trans));
```

$$Sx := \begin{bmatrix} [\mu x_1 + \mu \alpha_1 R_0 x_1 x_2 + \mu x_3 \alpha_2 - \mu x_3 + \mu, -\mu \alpha_1 R_0 x_1 x_2, -\mu (\alpha_2 - 1) x_3], \\ [-\mu \alpha_1 R_0 x_1 x_2, \mu x_2 \alpha_1 + \mu \alpha_1 R_0 x_1 x_2, -\mu (\alpha_1 - 1) x_2], \\ [-\mu (\alpha_2 - 1) x_3, -\mu (\alpha_1 - 1) x_2, \mu x_2 \alpha_1 - \mu x_2 + \mu x_3 \alpha_2] \end{bmatrix} \quad (11)$$

Evaluate the local covariance matrix at the critical point:

```
> S:=simplify(subs(x1=x10,x2=x20,x3=x30,Sx));
```

$$S := \begin{bmatrix} \left[\frac{2 \mu (\alpha_2 + \alpha_1 - 1 - \alpha_1 \alpha_2 + \alpha_1 \alpha_2 R_0)}{R_0 (\alpha_2 + \alpha_1 - 1)}, -\frac{\mu \alpha_2 (-1 + R_0) \alpha_1}{R_0 (\alpha_2 + \alpha_1 - 1)}, \right. \\ \left. -\frac{\mu (\alpha_2 - 1) (-1 + R_0) (\alpha_1 - 1)}{R_0 (\alpha_2 + \alpha_1 - 1)} \right], \\ \left[-\frac{\mu \alpha_2 (-1 + R_0) \alpha_1}{R_0 (\alpha_2 + \alpha_1 - 1)}, \frac{2 \mu \alpha_2 (-1 + R_0) \alpha_1}{R_0 (\alpha_2 + \alpha_1 - 1)}, -\frac{\mu (\alpha_1 - 1) \alpha_2 (-1 + R_0)}{R_0 (\alpha_2 + \alpha_1 - 1)} \right], \\ \left[-\frac{\mu (\alpha_2 - 1) (-1 + R_0) (\alpha_1 - 1)}{R_0 (\alpha_2 + \alpha_1 - 1)}, -\frac{\mu (\alpha_1 - 1) \alpha_2 (-1 + R_0)}{R_0 (\alpha_2 + \alpha_1 - 1)} \right], \end{bmatrix} \quad (12)$$

$$\left[\frac{2\mu(\alpha_1 - 1)\alpha_2(-1 + R_0)}{R_0(\alpha_2 + \alpha_1 - 1)} \right]$$

Now proceed to solve $A = -S$, where $A = B \cdot \text{SIG} + \text{SIG} \cdot \text{BT}$, and where $\text{BT} = \text{Transpose}(B)$.
First introduce notation for the elements of the matrix SIG:

```
> SIG:=Matrix(3,[s11,s12,s13,s21,s22,s23,s31,s32,s33]);
```

$$\text{SIG} := \begin{bmatrix} s_{11} & s_{12} & s_{13} \\ s_{21} & s_{22} & s_{23} \\ s_{31} & s_{32} & s_{33} \end{bmatrix} \quad (13)$$

Next determine the matrix A:

```
> A:=Matrix(evalm(B*SIG+SIG*Transpose(B))):
```

Solve the 9 scalar equations found from the matrix equation $A + S = 0$ for the 9 unknowns in the matrix SIG:

```
> solve(convert(A+S,set),convert(SIG,set)):
```

```
> assign(%);
```

Determine the first term of the asymptotic approximation of each of the elements of the matrix SIG as α_1 becomes large.

```
> s11a:=simplify(op(1,asympt(s11,alpha1)));
s12:=s12;
s13a:=simplify(op(1,asympt(s13,alpha1)));
s21:=s21;
s22a:=simplify(op(1,asympt(s22,alpha1)));
s23a:=simplify(op(1,asympt(s23,alpha1)));
s31a:=simplify(op(1,asympt(s31,alpha1)));
s32a:=simplify(op(1,asympt(s32,alpha1)));
s33a:=simplify(op(1,asympt(s33,alpha1)));
```

$$s_{11a} := \frac{\alpha_1}{\alpha_2 R_0^2}$$

$$s_{12} := -\frac{1}{R_0}$$

$$s_{13a} := -\frac{\alpha_1}{\alpha_2 R_0^2}$$

$$s_{21} := -\frac{1}{R_0}$$

$$s_{22a} := \frac{-1 + R_0}{R_0^2}$$

$$s_{23a} := \frac{1}{R_0^2}$$

$$s_{31a} := -\frac{\alpha_1}{\alpha_2 R_0^2}$$

$$s32a := \frac{1}{R0^2}$$

$$s33a := \frac{\alpha1}{\alpha2 R0^2} \quad (14)$$

The expectation of S+I+R, divided by N, is denoted Expsum:

```
> Expsum:=simplify(x10+x20+x30);
```

$$Expsum := 1 \quad (15)$$

The variance of S+I+R, divided by N, is denoted Varsum:

```
> Varsum:=simplify(s11+s22+s33+2*s12+2*s13+2*s23);
```

$$Varsum := 1 \quad (16)$$

$\mu_I = x20*N$ divided by $\sigma_I = \sqrt{N*s22}$ is called ρ_I . Determine an approximation of ρ_I for large $\alpha1$:

```
> x20;
```

```
x20a:=op(1,asympt(x20,alpha1));
```

```
rhoI:=simplify(x20a*N/sqrt(s22a*N)) assuming R0>1;
```

$$x20a := \frac{\alpha2 (-1 + R0)}{R0 (\alpha2 + \alpha1 - 1)}$$

$$rhoI := \frac{\alpha2 \sqrt{-1 + R0} \sqrt{N}}{\alpha1} \quad (17)$$

```
> x10;
```

```
rhoS:=simplify(x10*N/sqrt(s11a*N)) assuming R0>1,alpha1>1,
```

```
alpha2>1;
```

$$rhoS := \frac{1}{R0} \frac{\sqrt{N} \sqrt{\alpha2}}{\sqrt{\alpha1}} \quad (18)$$

```
> x30;
```

$$\frac{(-1 + R0) (\alpha1 - 1)}{R0 (\alpha2 + \alpha1 - 1)} \quad (19)$$

```
> x30a:=op(1,asympt(x30,alpha1));
```

$$x30a := \frac{-1 + R0}{R0} \quad (20)$$

```
> rhoR:=simplify(x30a*N/sqrt(s33a*N)) assuming R0>1,N>0,alpha1>1,
```

```
alpha2>0;
```

$$rhoR := \frac{(-1 + R0) \sqrt{N} \sqrt{\alpha2}}{\sqrt{\alpha1}} \quad (21)$$

We determine also the first two terms in the asymptotic approximations of the elements of SIG for large alpha:

```
> s11b:=map(simplify,convert(asympt(s11,alpha1,1),polynom));
s12:=s12;
s13b:=map(simplify,convert(asympt(s13,alpha1,1),polynom));
s21:=s21;
s22b:=map(simplify,convert(asympt(s22,alpha1,2),polynom));
s23b:=map(simplify,convert(asympt(s23,alpha1,2),polynom));
s31b:=map(simplify,convert(asympt(s31,alpha1,1),polynom));
s32b:=map(simplify,convert(asympt(s32,alpha1,2),polynom));
s33b:=map(simplify,convert(asympt(s33,alpha1,1),polynom));
```

$$\begin{aligned}
 s11b &:= \frac{\alpha 1}{\alpha 2 R 0^2} + \frac{\alpha 2 R 0^2 + \alpha 2 R 0 - R 0 - \alpha 2 + 1}{\alpha 2 R 0^3} \\
 s12 &:= -\frac{1}{R 0} \\
 s13b &:= -\frac{\alpha 1}{\alpha 2 R 0^2} - \frac{-\alpha 2 + 1 + \alpha 2 R 0 - R 0}{\alpha 2 R 0^3} \\
 s21 &:= -\frac{1}{R 0} \\
 s22b &:= \frac{-1 + R 0}{R 0^2} + \frac{R 0^3 \alpha 2 - \alpha 2 R 0^2 + R 0^2 + \alpha 2 - 1}{\alpha 1 R 0^3} \\
 s23b &:= \frac{1}{R 0^2} + \frac{\alpha 2 R 0^2 - \alpha 2 + 1}{\alpha 1 R 0^3} \\
 s31b &:= -\frac{\alpha 1}{\alpha 2 R 0^2} - \frac{-\alpha 2 + 1 + \alpha 2 R 0 - R 0}{\alpha 2 R 0^3} \\
 s32b &:= \frac{1}{R 0^2} + \frac{\alpha 2 R 0^2 - \alpha 2 + 1}{\alpha 1 R 0^3} \\
 s33b &:= \frac{\alpha 1}{\alpha 2 R 0^2} + \frac{R 0^3 \alpha 2 - R 0 - \alpha 2 + 1}{\alpha 2 R 0^3}
 \end{aligned} \tag{22}$$