

This Maple worksheet is denoted sirs3pr.

It is used to study an SIRS model with demography, for  $R_0 > 1$  and  $\alpha_1$  large.

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

The particular model dealt here has a state space with 3 variables: S, I, and R, and the infection rate is "proper", with  $S+I+R-1$  in the denominator.

Two things are done here.

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

Ingemar Nåsell, KTH, Stockholm, 2012-10-03.

```
> 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/
(S+I+R-1),[-1,0,0]=mu*S,[0,-1,1]=gamma*I,[0,-1,0]=mu*I,[0,0,-1]
=mu*R]);
```

$$\text{transA} := \text{table} \left( \left[ \begin{array}{l} [0, 0, -1] = \mu R, [0, -1, 0] = \mu I, [0, -1, 1] = \gamma I, [-1, 0, 0] = \mu S, \\ [-1, 1, 0] = \frac{\beta SI}{S+I+R-1}, [1, 0, -1] = \delta R, [1, 0, 0] = \mu N \end{array} \right] \right) \quad (2)$$

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:

```
> 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);
```

$$\begin{aligned}
 \text{trans} := \text{table} \left( \left[ \begin{array}{l}
 [0, 0, -1] = \mu x_3, [0, -1, 0] = \mu x_2, [0, -1, 1] = \mu (\alpha_1 - 1) x_2, [ \\
 -1, 0, 0] = \mu x_1, [-1, 1, 0] = \frac{N\mu \alpha_1 R_0 x_1 x_2}{x_1 N + x_2 N + x_3 N - 1}, [1, 0, -1] = \mu (\alpha_2 \\
 -1) x_3, [1, 0, 0] = \mu \end{array} \right] \right) \quad (3)
 \end{aligned}$$

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:=equ(3,trans);

```

$$eq1 := -\mu x_1 - \frac{N\mu \alpha_1 R_0 x_1 x_2}{x_1 N + x_2 N + x_3 N - 1} + \mu (\alpha_2 - 1) x_3 + \mu$$

$$eq2 := \frac{\mu x_2 \alpha_1 (-x_1 N - x_2 N - x_3 N + 1 + N R_0 x_1)}{x_1 N + x_2 N + x_3 N - 1}$$

$$eq3 := -\mu x_3 + \mu (\alpha_1 - 1) x_2 - \mu (\alpha_2 - 1) x_3 \quad (4)$$

Critical points:

```

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

```

$$\begin{aligned}
 \text{crit} := \{x_1 = 1, x_2 = 0, x_3 = 0\}, \left\{ x_1 = \frac{N-1}{NR_0}, x_2 = \frac{\alpha_2 (-N + NR_0 + 1)}{NR_0 (\alpha_2 + \alpha_1 - 1)}, x_3 \right. \\
 \left. = \frac{(\alpha_1 - 1) (-N + NR_0 + 1)}{NR_0 (\alpha_2 + \alpha_1 - 1)} \right\} \quad (5)
 \end{aligned}$$

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:=map(factor,rhs(crit[2][3]));

```

$$x_{10} := \frac{N-1}{NR_0}$$

$$x_{20} := \frac{\alpha_2 (-N + NR_0 + 1)}{NR_0 (\alpha_2 + \alpha_1 - 1)}$$

$$x_{30} := \frac{(\alpha_1 - 1) (-N + NR_0 + 1)}{NR_0 (\alpha_2 + \alpha_1 - 1)} \quad (6)$$

The Jacobian of the system of ODEs is denoted  $B_x$ :

```

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

```

$$\begin{aligned}
 Bx := & \left[ \left[ -\mu - \frac{N\mu\alpha_1 R_0 x_2}{x_1 N + x_2 N + x_3 N - 1} + \frac{N^2 \mu \alpha_1 R_0 x_1 x_2}{(x_1 N + x_2 N + x_3 N - 1)^2}, \right. \right. \\
 & - \frac{N\mu\alpha_1 R_0 x_1}{x_1 N + x_2 N + x_3 N - 1} + \frac{N^2 \mu \alpha_1 R_0 x_1 x_2}{(x_1 N + x_2 N + x_3 N - 1)^2}, \\
 & \left. \left. \frac{N^2 \mu \alpha_1 R_0 x_1 x_2}{(x_1 N + x_2 N + x_3 N - 1)^2} + \mu(\alpha_2 - 1) \right], \right. \\
 & \left[ \frac{\mu x_2 \alpha_1 (-N + NR_0)}{x_1 N + x_2 N + x_3 N - 1} - \frac{\mu x_2 \alpha_1 (-x_1 N - x_2 N - x_3 N + 1 + NR_0 x_1) N}{(x_1 N + x_2 N + x_3 N - 1)^2}, \right. \\
 & \frac{\mu \alpha_1 (-x_1 N - x_2 N - x_3 N + 1 + NR_0 x_1)}{x_1 N + x_2 N + x_3 N - 1} - \frac{\mu x_2 \alpha_1 N}{x_1 N + x_2 N + x_3 N - 1} \\
 & - \frac{\mu x_2 \alpha_1 (-x_1 N - x_2 N - x_3 N + 1 + NR_0 x_1) N}{(x_1 N + x_2 N + x_3 N - 1)^2}, - \frac{\mu x_2 \alpha_1 N}{x_1 N + x_2 N + x_3 N - 1} \\
 & \left. \left. - \frac{\mu x_2 \alpha_1 (-x_1 N - x_2 N - x_3 N + 1 + NR_0 x_1) N}{(x_1 N + x_2 N + x_3 N - 1)^2} \right], \right. \\
 & \left. \left[ 0, \mu(\alpha_1 - 1), -\mu - \mu(\alpha_2 - 1) \right] \right]
 \end{aligned} \tag{7}$$

Evaluate the Jacobian at the critical point:

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

$$\begin{aligned}
 B := & \left[ \left[ -\frac{1}{R_0(\alpha_2 N + \alpha_1 N - N - \alpha_2 - \alpha_1 + 1)} (\mu(N\alpha_2\alpha_1 - 2N\alpha_1 R_0\alpha_2 \right. \right. \\
 & + \alpha_1 NR_0 + R_0^2\alpha_2\alpha_1 N - NR_0 + NR_0\alpha_2 - \alpha_2\alpha_1 - R_0\alpha_1 + R_0\alpha_2\alpha_1 + R_0 \\
 & - \alpha_2 R_0)), - \frac{\mu\alpha_1(\alpha_1 NR_0 - NR_0 - \alpha_2 R_0 - R_0\alpha_1 + R_0 + \alpha_2 N - \alpha_2)}{R_0(\alpha_2 N + \alpha_1 N - N - \alpha_2 - \alpha_1 + 1)}, \\
 & \frac{1}{R_0(\alpha_2 N + \alpha_1 N - N - \alpha_2 - \alpha_1 + 1)} (\mu(2N\alpha_1 R_0\alpha_2 - \alpha_1 NR_0 - N\alpha_2\alpha_1 \\
 & + NR_0 - 2NR_0\alpha_2 + NR_0\alpha_2^2 + \alpha_2\alpha_1 + R_0\alpha_1 - R_0\alpha_2\alpha_1 - R_0 + 2\alpha_2 R_0 \\
 & - \alpha_2^2 R_0)), \left. \right], \\
 & \left[ \frac{\mu\alpha_2(-N + NR_0 + 1)\alpha_1(-1 + R_0)}{R_0(\alpha_2 N + \alpha_1 N - N - \alpha_2 - \alpha_1 + 1)}, \right. \\
 & - \frac{\mu\alpha_2(-N + NR_0 + 1)\alpha_1}{R_0(\alpha_2 N + \alpha_1 N - N - \alpha_2 - \alpha_1 + 1)}, - \frac{\mu\alpha_2(-N + NR_0 + 1)\alpha_1}{R_0(\alpha_2 N + \alpha_1 N - N - \alpha_2 - \alpha_1 + 1)} \\
 & \left. \right],
 \end{aligned} \tag{8}$$

$$\begin{bmatrix} 0, \mu (\alpha 1 - 1), -\mu \alpha 2 \end{bmatrix}$$

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 remaining two eigenvalues.

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

$$eig := \begin{bmatrix} -\mu \end{bmatrix},$$

(9)

$$\begin{aligned} & \left[ -\frac{1}{2} \frac{1}{\alpha 2 N + \alpha 1 N - N - \alpha 2 - \alpha 1 + 1} \left( \left( N \alpha 2^2 - \alpha 2^2 + N \alpha 1 R 0 \alpha 2 - \alpha 2 N \right. \right. \right. \\ & + \alpha 2 \\ & - \left( \alpha 2^2 + 16 N \alpha 2^2 \alpha 1 + 16 N \alpha 2 \alpha 1^2 - 8 N \alpha 2 \alpha 1 - 8 \alpha 2 \alpha 1^2 N^2 \right. \\ & - 8 \alpha 2^2 \alpha 1 N^2 - 2 N \alpha 2^2 - 8 \alpha 2^2 \alpha 1 - 8 \alpha 2 \alpha 1^2 + 4 \alpha 2 \alpha 1 + \alpha 2^2 N^2 \\ & + 4 N \alpha 1 R 0 \alpha 2 - 4 N^2 R 0 \alpha 2 \alpha 1 + 6 \alpha 2^2 \alpha 1 N^2 R 0 + 8 \alpha 2 \alpha 1^2 N^2 R 0 \\ & - 6 \alpha 2^2 \alpha 1 N R 0 - 8 \alpha 2 \alpha 1^2 N R 0 + 4 \alpha 2 N^2 \alpha 1 + 4 N \alpha 2^3 - 2 \alpha 2^3 N^2 - 2 \alpha 2^3 \\ & - 8 N^2 \alpha 2^2 \alpha 1^2 R 0 - 2 N^2 \alpha 2^3 \alpha 1 R 0 + 8 N \alpha 2^2 \alpha 1^2 R 0 + 2 N \alpha 2^3 \alpha 1 R 0 \\ & + N^2 \alpha 1^2 R 0^2 \alpha 2^2 - 4 \alpha 2 \alpha 1^3 N^2 R 0 + 4 \alpha 2 \alpha 1^3 N R 0 + 4 N^2 \alpha 2^3 \alpha 1 - 8 N \alpha 2^3 \alpha 1 \\ & + 4 \alpha 2 \alpha 1^3 N^2 - 8 \alpha 2 \alpha 1^3 N + 8 \alpha 2^2 \alpha 1^2 + 8 N^2 \alpha 2^2 \alpha 1^2 - 16 N \alpha 2^2 \alpha 1^2 \\ & \left. \left. \left. + 4 \alpha 2^3 \alpha 1 + 4 \alpha 2 \alpha 1^3 + \alpha 2^4 + N^2 \alpha 2^4 - 2 N \alpha 2^4 \right)^{1/2} \right) \mu \right], \\ & \left[ -\frac{1}{2} \frac{1}{\alpha 2 N + \alpha 1 N - N - \alpha 2 - \alpha 1 + 1} \left( \left( N \alpha 2^2 - \alpha 2^2 + N \alpha 1 R 0 \alpha 2 - \alpha 2 N \right. \right. \right. \\ & + \alpha 2 \\ & + \left( \alpha 2^2 + 16 N \alpha 2^2 \alpha 1 + 16 N \alpha 2 \alpha 1^2 - 8 N \alpha 2 \alpha 1 - 8 \alpha 2 \alpha 1^2 N^2 \right. \\ & - 8 \alpha 2^2 \alpha 1 N^2 - 2 N \alpha 2^2 - 8 \alpha 2^2 \alpha 1 - 8 \alpha 2 \alpha 1^2 + 4 \alpha 2 \alpha 1 + \alpha 2^2 N^2 \\ & + 4 N \alpha 1 R 0 \alpha 2 - 4 N^2 R 0 \alpha 2 \alpha 1 + 6 \alpha 2^2 \alpha 1 N^2 R 0 + 8 \alpha 2 \alpha 1^2 N^2 R 0 \\ & - 6 \alpha 2^2 \alpha 1 N R 0 - 8 \alpha 2 \alpha 1^2 N R 0 + 4 \alpha 2 N^2 \alpha 1 + 4 N \alpha 2^3 - 2 \alpha 2^3 N^2 - 2 \alpha 2^3 \\ & - 8 N^2 \alpha 2^2 \alpha 1^2 R 0 - 2 N^2 \alpha 2^3 \alpha 1 R 0 + 8 N \alpha 2^2 \alpha 1^2 R 0 + 2 N \alpha 2^3 \alpha 1 R 0 \end{aligned}$$

$$\begin{aligned}
& + N^2 \alpha 1^2 R 0^2 \alpha 2^2 - 4 \alpha 2 \alpha 1^3 N^2 R 0 + 4 \alpha 2 \alpha 1^3 N R 0 + 4 N^2 \alpha 2^3 \alpha 1 - 8 N \alpha 2^3 \alpha 1 \\
& + 4 \alpha 2 \alpha 1^3 N^2 - 8 \alpha 2 \alpha 1^3 N + 8 \alpha 2^2 \alpha 1^2 + 8 N^2 \alpha 2^2 \alpha 1^2 - 16 N \alpha 2^2 \alpha 1^2 \\
& + 4 \alpha 2^3 \alpha 1 + 4 \alpha 2 \alpha 1^3 + \alpha 2^4 + N^2 \alpha 2^4 - 2 N \alpha 2^4)^{1/2}) \mu) \Big] \Big]
\end{aligned}$$

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

Next we determine the characteristic polynomial:

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

$$\begin{aligned}
p := & \lambda^3 + \frac{\mu (N \alpha 2^2 - \alpha 2^2 + \alpha 1 N + N \alpha 1 R 0 \alpha 2 - N - \alpha 1 + 1) \lambda^2}{\alpha 2 N + \alpha 1 N - N - \alpha 2 - \alpha 1 + 1} \\
& + \frac{1}{\alpha 2 N + \alpha 1 N - N - \alpha 2 - \alpha 1 + 1} (\alpha 2 \mu^2 (-N \alpha 1^2 + \alpha 1 N - N \alpha 2 \alpha 1 \\
& + N \alpha 1 R 0 \alpha 2 - N + \alpha 2 N - \alpha 1 + \alpha 2 \alpha 1 + 1 - \alpha 2 + R 0 N \alpha 1^2 + \alpha 1^2) \lambda) \\
& + \frac{\mu^3 \alpha 2 (-N + N R 0 + 1) \alpha 1}{N - 1}
\end{aligned} \tag{10}$$

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));**

$$p1 := \lambda^2 + \frac{\mu \alpha 2 (\alpha 2 N - \alpha 2 + \alpha 1 N R 0 - N + 1) \lambda}{\alpha 2 N + \alpha 1 N - N - \alpha 2 - \alpha 1 + 1} + \frac{\mu^2 \alpha 2 (-N + N R 0 + 1) \alpha 1}{N - 1} \tag{11}$$

The two roots of  $p1 = 0$ , where  $p1 = \lambda^2 + a \cdot \lambda + b$ , can be written  $-\frac{a}{2} \pm i \cdot \Omega$ , where

$$\Omega = \sqrt{b - \left(\frac{a}{2}\right)^2}.$$

We note that  $\mu$  is a factor in  $\Omega$ , so we can write  $\Omega = \mu \cdot \sqrt{b1 - \left(\frac{a1}{2}\right)^2}$ , where  $b1 = \frac{b}{\mu^2}$ ,

and  $a1 = \frac{a}{\mu}$ .

Thus,

**> a1:=op(2,p1)/mu/lambda;**  
**b1:=op(3,p1)/mu^2;**

$$\begin{aligned}
a1 := & \frac{\alpha 2 (\alpha 2 N - \alpha 2 + \alpha 1 N R 0 - N + 1)}{\alpha 2 N + \alpha 1 N - N - \alpha 2 - \alpha 1 + 1} \\
b1 := & \frac{\alpha 2 (-N + N R 0 + 1) \alpha 1}{N - 1}
\end{aligned} \tag{12}$$

Consider the quantity  $C = b1 - \left(\frac{a1}{2}\right)^2$

**> C:=b1-(a1/2)^2;**

**(13)**

$$C := \frac{\alpha_2 (-N + NR_0 + 1) \alpha_1}{N - 1} - \frac{1}{4} \frac{\alpha_2^2 (\alpha_2 N - \alpha_2 + \alpha_1 NR_0 - N + 1)^2}{(\alpha_2 N + \alpha_1 N - N - \alpha_2 - \alpha_1 + 1)^2} \quad (13)$$

C is seen to depend on N.

We determine the asymptotic approximation of C as N becomes large.

The one-term asymptotic approximation of C, denoted C1, equals the limit of C as N approaches infinity.

```
> C1:=limit(b1,N=infinity)-limit((a1/2)^2,N=infinity);
```

$$C1 := \alpha_2 (-1 + R_0) \alpha_1 - \frac{1}{4} \frac{\alpha_2^2 (\alpha_2 + R_0 \alpha_1 - 1)^2}{(\alpha_2 + \alpha_1 - 1)^2} \quad (14)$$

This can be written

```
> C2:=op(1,C1)-(alpha2*R1/2)^2;
```

$$C2 := \alpha_2 (-1 + R_0) \alpha_1 - \frac{1}{4} \alpha_2^2 R1^2 \quad (15)$$

where

```
> R1=(alpha1*R0+alpha2-1)/(alpha1+alpha2-1);
```

$$R1 = \frac{\alpha_2 + R_0 \alpha_1 - 1}{\alpha_2 + \alpha_1 - 1} \quad (16)$$

We summarize: The limit of  $\Omega$  as N approaches infinity can be written  $\Omega_1 = \mu^*$

$$\sqrt{\alpha_1 \cdot \alpha_2 \cdot (R_0 - 1) - \left( \alpha_2 \cdot \frac{R1}{2} \right)^2}$$

This finishes the study of the eigenvalues.

The rest of the worksheet is used to derive approximations of the matrix of covariances for the 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));
```

(17)

$$\begin{aligned}
 Sx := & \left[ \left[ \frac{1}{x_1 N + x_2 N + x_3 N - 1} (\mu (x_1^2 N + x_1 x_2 N - x_1 + N \alpha_1 R_0 x_1 x_2 \right. \right. \\
 & + x_3 \alpha_2 x_1 N + x_3 \alpha_2 x_2 N + \alpha_2 x_3^2 N - x_3 \alpha_2 - x_3 x_2 N - x_3^2 N + x_3 + x_1 N \\
 & + x_2 N + x_3 N - 1)), -\frac{N \mu \alpha_1 R_0 x_1 x_2}{x_1 N + x_2 N + x_3 N - 1}, -\mu (\alpha_2 - 1) x_3 \Big], \\
 & \left[ -\frac{N \mu \alpha_1 R_0 x_1 x_2}{x_1 N + x_2 N + x_3 N - 1}, \frac{\mu x_2 \alpha_1 (x_1 N + x_2 N + x_3 N - 1 + N R_0 x_1)}{x_1 N + x_2 N + x_3 N - 1}, \right. \\
 & \left. -\mu (\alpha_1 - 1) x_2 \right], \\
 & \left[ -\mu (\alpha_2 - 1) x_3, -\mu (\alpha_1 - 1) x_2, \alpha_1 \mu x_2 - \mu x_2 + \mu x_3 \alpha_2 \right] \Big]
 \end{aligned} \tag{17}$$

Evaluate the local covariance matrix at the critical point:

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

$$\begin{aligned}
 S := & \left[ \left[ \frac{2 \mu (-N \alpha_2 \alpha_1 + \alpha_1 N + N \alpha_1 R_0 \alpha_2 - N + \alpha_2 N - \alpha_1 + \alpha_2 \alpha_1 + 1 - \alpha_2)}{R_0 (\alpha_2 + \alpha_1 - 1) N}, \right. \right. \\
 & \left. -\frac{\alpha_1 \mu \alpha_2 (-N + N R_0 + 1)}{N R_0 (\alpha_2 + \alpha_1 - 1)}, -\frac{\mu (\alpha_2 - 1) (\alpha_1 - 1) (-N + N R_0 + 1)}{N R_0 (\alpha_2 + \alpha_1 - 1)} \right], \\
 & \left[ -\frac{\alpha_1 \mu \alpha_2 (-N + N R_0 + 1)}{N R_0 (\alpha_2 + \alpha_1 - 1)}, \frac{2 \alpha_1 \mu \alpha_2 (-N + N R_0 + 1)}{N R_0 (\alpha_2 + \alpha_1 - 1)}, \right. \\
 & \left. -\frac{\mu (\alpha_1 - 1) \alpha_2 (-N + N R_0 + 1)}{N R_0 (\alpha_2 + \alpha_1 - 1)} \right], \\
 & \left[ -\frac{\mu (\alpha_2 - 1) (\alpha_1 - 1) (-N + N R_0 + 1)}{N R_0 (\alpha_2 + \alpha_1 - 1)}, -\frac{\mu (\alpha_1 - 1) \alpha_2 (-N + N R_0 + 1)}{N R_0 (\alpha_2 + \alpha_1 - 1)}, \right. \\
 & \left. \frac{2 \mu (\alpha_1 - 1) \alpha_2 (-N + N R_0 + 1)}{N R_0 (\alpha_2 + \alpha_1 - 1)} \right] \Big]
 \end{aligned} \tag{18}$$

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]);
```

$$\text{SIG} := \begin{bmatrix} s_{11} & s_{12} & s_{13} \\ s_{21} & s_{22} & s_{23} \\ s_{31} & s_{32} & s_{33} \end{bmatrix} \tag{19}$$

Next, evaluate the matrix A:

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

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

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

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

All the elements of the matrix SIG depend on both N and on  $\alpha_1$ .

We determine the one-term asymptotic approximations of each of them as both N and  $\alpha_1$  become large:

```
> s11a:=simplify(op(1,asymp(op(1,asymp(s11,alpha1)),N)));
s12a:=simplify(op(1,asymp(op(1,asymp(s12,alpha1)),N)));
s13a:=simplify(op(1,asymp(op(1,asymp(s13,alpha1)),N)));
s21a:=simplify(op(1,asymp(op(1,asymp(s21,alpha1)),N)));
s22a:=simplify(op(1,asymp(op(1,asymp(s22,alpha1)),N)));
s23a:=simplify(op(1,asymp(op(1,asymp(s23,alpha1)),N)));
s31a:=simplify(op(1,asymp(op(1,asymp(s31,alpha1)),N)));
s32a:=simplify(op(1,asymp(op(1,asymp(s32,alpha1)),N)));
s33a:=simplify(op(1,asymp(op(1,asymp(s33,alpha1)),N)));
```

$$s11a := \frac{\alpha_1}{\alpha_2 R_0^2}$$

$$s12a := -\frac{1}{R_0}$$

$$s13a := -\frac{\alpha_1}{\alpha_2 R_0^2}$$

$$s21a := -\frac{1}{R_0}$$

$$s22a := \frac{-1 + R_0}{R_0^2}$$

$$s23a := \frac{1}{R_0^2}$$

$$s31a := -\frac{\alpha_1}{\alpha_2 R_0^2}$$

$$s32a := \frac{1}{R_0^2}$$

$$s33a := \frac{\alpha_1}{\alpha_2 R_0^2} \quad (20)$$

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

```
> Expsum:=simplify(x10+x20+x30);
Expsum:= 1 \quad (21)
```

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 (22)
```

I define  $\rho_I$  as the first term in the asymptotic approximation of the ratio  $x_{20} \cdot N / \sqrt{N \cdot s_{22}}$  for large  $\alpha_1$  and N.

```
> x20;
x20a:=op(1,asymp(op(1,asymp(x20,alpha1)),N));
```

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

$$\frac{\alpha^2 (-N + NR_0 + 1)}{NR_0 (\alpha^2 + \alpha - 1)}$$

$$x20a := \frac{\alpha^2 (-1 + R_0)}{R_0 \alpha}$$

$$rhoI := \frac{\alpha^2 \sqrt{-1 + R_0} \sqrt{N}}{\alpha}$$

**(23)**

This expression for rhoI is the same as for sirs2c and sirs3c.