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> #Maple code for example 1 of paper Parameter Redundancy and Identifiability in Hidden Markov
  Models by D.J. Cole
> with(LinearAlgebra) :
> Dmat := proc(se, pars)
  local DD1, i, j;
  description "Form the derivative matrix";
  with(LinearAlgebra) :
  DD1 := Matrix(1 .. Dimension(pars), 1 .. Dimension(se)) :
  for i from 1 to Dimension(pars) do
    for j from 1 to Dimension(se) do
      DD1[i, j] := diff(se[j], pars[i])
    end do
  end do;
  DD1;
end proc:
> Hybrid := proc(D1, pars, minpars, maxpars, ret)
  local results, j, numpars, D1rand, ans, roll :
  description "This procedure finds the rank and alpha for the hybrid-symbolic-numeric method.
  If ret = 1 returns full results. Otherwise returns model rank.";
  results := Matrix(5, 2) :
  for j from 1 to 5 do
    roll := rand(minpars .. maxpars) :
    numpars := seq(pars[i] = evalf(roll( )), i = 1 .. Dimension(pars)) :
    D1rand := eval(D1, {numpars});
    results[j, 1] := Rank(D1rand);
    results[j, 2] := NullSpace(Transpose(D1rand)) :
  end do:
  if ret = 1 then
    ans := results :
  else
    ans := max(results[1 .. 5, 1]) :
  end if:
  ans :
end proc:
> Estpar := proc(DD1, pars, ret)
  local r, d, alphapre, alpha, PDE, FF, i, ans;
  description "Finds the estimable set of parameters for derivative matrix DD1. If ret = 1 returns
  alpha, PDEs, estimable parameter combinations. Otherwise returns estimable parameter
  combinations";
  with(LinearAlgebra) :
  r := Rank(DD1); d := Dimension(pars) - r :
  alphapre := NullSpace(Transpose(DD1)) : alpha := Matrix(d, Dimension(pars)) : PDE :=
  Vector(d) :
  FF := f(seq(pars[i], i = 1 .. Dimension(pars))) :
  for i from 1 to d do
    alpha[i, 1 .. Dimension(pars)] := alphapre[i] :
    PDE[i] := add(diff(FF, pars[j]) * alpha[i, j], j = 1 .. Dimension(pars)) :
  end do:
  if ret = 1 then
    ans := <pdsolve({seq(PDE[i] = 0, i = 1 .. d)}), {alpha}, {PDE}> :
  elif ret = 2 then

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    ans := ⟨{alpha}, {PDE}⟩ :
else
    ans := pdsolve( {seq(PDE[i]=0, i=1 ..d)} ) :
end if:
ans :
end proc:

```

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> HMMexsum := proc(deltavector, Gammamatrix, Pmatrix, st, T)
local i, m, Prodkappa, kappa

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description "This procedure finds the exhaustive summary for a HMM. Inputs: vector and matrices of HMM, st = 1 if stationary, T no. data points";

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m := Dimension(deltavector) :

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kappa := Vector(T) :

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if st = 1 then

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```

    Prodkappa := deltavector • eval(Gammamatrix, t = 1).eval(Pmatrix, t = 1);

```

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else

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```

    Prodkappa := deltavector • eval(Pmatrix, t = 1);

```

```

end if:

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kappa[1] := (Prodkappa • Matrix(m, 1, 1))[1];

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for i from 2 to T do

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    Prodkappa := Prodkappa.eval(Gammamatrix, t = i).eval(Pmatrix, t = i) :

```

```

    kappa[i] := (Prodkappa • Matrix(m, 1, 1))[1];

```

```

end do:

```

```

kappa

```

```

end proc:

```

```

> # Setting up matrices and vector that define the model

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> m := 2 :

```

```

> deltavector := ⟨delta[1]|1 - delta[1]⟩;

```

$$\text{deltavector} := \begin{bmatrix} \delta_1 & 1 - \delta_1 \end{bmatrix} \quad (1)$$

```

> Gammamatrix := Matrix(m, m) :

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for i from 1 to m do

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    for j from 1 to m - 1 do

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        Gammamatrix[i, j] := gamma[i, j] :

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```

    end do:

```

```

        Gammamatrix[i, m] := 1 - add(gamma[i, j], j = 1 ..m - 1) :

```

```

end do:

```

```

> Gammamatrix

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$$\begin{bmatrix} \gamma_{1,1} & 1 - \gamma_{1,1} \\ \gamma_{2,1} & 1 - \gamma_{2,1} \end{bmatrix} \quad (2)$$

```

> Pyvector := ⟨seq( (exp(-lambda[j]) • lambda[j]^x[t]) / x[t]!, j = 1 ..m) ⟩ :

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> Pmatrix := DiagonalMatrix(Pyvector);

```

$$Pmatrix := \begin{bmatrix} \frac{e^{-\lambda_1} \lambda_1^{x_t}}{x_t!} & 0 \\ 0 & \frac{e^{-\lambda_2} \lambda_2^{x_t}}{x_t!} \end{bmatrix} \quad (3)$$

> # When there are 3 data points:

>

> T := 3 :

> kappa := HMMexsum(deltavector, Gammamatrix, Pmatrix, 1, T);

$$\kappa := \begin{bmatrix} \left[\frac{(\delta_1 \gamma_{1,1} + (1 - \delta_1) \gamma_{2,1}) e^{-\lambda_1} \lambda_1^{x_1}}{x_1!} + \frac{(\delta_1 (1 - \gamma_{1,1}) + (1 - \delta_1) (1 - \gamma_{2,1})) e^{-\lambda_2} \lambda_2^{x_1}}{x_1!} \right] \right. \\ \left. \left[\frac{1}{x_2!} \left(\left(\frac{(\delta_1 \gamma_{1,1} + (1 - \delta_1) \gamma_{2,1}) e^{-\lambda_1} \lambda_1^{x_1} \gamma_{1,1}}{x_1!} + \frac{(\delta_1 (1 - \gamma_{1,1}) + (1 - \delta_1) (1 - \gamma_{2,1})) e^{-\lambda_2} \lambda_2^{x_1} \gamma_{2,1}}{x_1!} \right) e^{-\lambda_1} \lambda_1^{x_2} \right) \right. \right. \\ \left. \left. + \frac{1}{x_2!} \left(\left(\frac{(\delta_1 \gamma_{1,1} + (1 - \delta_1) \gamma_{2,1}) e^{-\lambda_1} \lambda_1^{x_1} (1 - \gamma_{1,1})}{x_1!} + \frac{(\delta_1 (1 - \gamma_{1,1}) + (1 - \delta_1) (1 - \gamma_{2,1})) e^{-\lambda_2} \lambda_2^{x_1} (1 - \gamma_{2,1})}{x_1!} \right) e^{-\lambda_2} \lambda_2^{x_2} \right) \right] \right. \\ \left. \left[\frac{1}{x_3!} \left(\left(\frac{1}{x_2!} \left(\left(\frac{(\delta_1 \gamma_{1,1} + (1 - \delta_1) \gamma_{2,1}) e^{-\lambda_1} \lambda_1^{x_1} \gamma_{1,1}}{x_1!} \right) \right) \right) \right) \right] \right. \end{bmatrix} \quad (4)$$

$$\begin{aligned}
& + \frac{(\delta_1 (1 - \gamma_{1,1}) + (1 - \delta_1) (1 - \gamma_{2,1})) e^{-\lambda_2 \lambda_1^{x_1} \gamma_{2,1}}}{x_1!} \left. e^{-\lambda_1 \lambda_1^{x_2} \gamma_{1,1}} \right) \\
& + \frac{1}{x_2!} \left(\left(\frac{(\delta_1 \gamma_{1,1} + (1 - \delta_1) \gamma_{2,1}) e^{-\lambda_1 \lambda_1^{x_1} (1 - \gamma_{1,1})}}{x_1!} \right. \right. \\
& + \left. \left. \frac{(\delta_1 (1 - \gamma_{1,1}) + (1 - \delta_1) (1 - \gamma_{2,1})) e^{-\lambda_2 \lambda_1^{x_1} (1 - \gamma_{2,1})}}{x_1!} \right) e^{-\lambda_2 \lambda_2^{x_2} \gamma_{2,1}} \right) e^{-\lambda_1 \lambda_1^{x_3}} \\
& + \frac{1}{x_3!} \left(\left(\frac{1}{x_2!} \left(\left(\frac{(\delta_1 \gamma_{1,1} + (1 - \delta_1) \gamma_{2,1}) e^{-\lambda_1 \lambda_1^{x_1} \gamma_{1,1}}}{x_1!} \right. \right. \right. \\
& + \left. \left. \frac{(\delta_1 (1 - \gamma_{1,1}) + (1 - \delta_1) (1 - \gamma_{2,1})) e^{-\lambda_2 \lambda_1^{x_1} \gamma_{2,1}}}{x_1!} \right) e^{-\lambda_1 \lambda_1^{x_2} (1 - \gamma_{1,1})} \right) \\
& + \frac{1}{x_2!} \left(\left(\frac{(\delta_1 \gamma_{1,1} + (1 - \delta_1) \gamma_{2,1}) e^{-\lambda_1 \lambda_1^{x_1} (1 - \gamma_{1,1})}}{x_1!} \right. \right. \\
& + \left. \left. \frac{(\delta_1 (1 - \gamma_{1,1}) + (1 - \delta_1) (1 - \gamma_{2,1})) e^{-\lambda_2 \lambda_1^{x_1} (1 - \gamma_{2,1})}}{x_1!} \right) e^{-\lambda_2 \lambda_2^{x_2} (1 - \gamma_{2,1})} \right) \\
& \left. \left. \left. e^{-\lambda_2 \lambda_2^{x_3}} \right) \right) \right)
\end{aligned}$$

>

> $\text{pars} := \langle \gamma_{1,1}, \gamma_{2,1}, \lambda_1, \lambda_2 \rangle;$

$$pars := \begin{bmatrix} \gamma_{1,1} \\ \gamma_{2,1} \\ \lambda_1 \\ \lambda_2 \end{bmatrix} \quad (5)$$

> $D1 := Dmat(kappa, pars) :$

> $r := Rank(D1); d := Dimension(pars) - r;$
 $r := 3$
 $d := 1$ (6)

> $T := 4 :$

> $kappa := HMMexsum(deltavector, Gammamatrix, Pmatrix, 1, T) :$

> $D1 := Dmat(kappa, pars) :$

> $r := Rank(D1); d := Dimension(pars) - r;$
 $r := 4$
 $d := 0$ (7)

> # Local identifiability too complex to solve

> # solve({seq(kappa[i] = k[i], i = 1 ..4)}, {seq(pars[i], i = 1 ..4)});

> $m := 3 :$

> $deltavector := \langle \delta_1 \delta_2 | 1 - \delta_1 - \delta_2 \rangle;$

$$deltavector := \begin{bmatrix} \delta_1 & \delta_2 & 1 - \delta_1 - \delta_2 \end{bmatrix} \quad (8)$$

> $Gammamatrix := Matrix(m, m) :$

for i from 1 to m do

for j from 1 to m - 1 do

$Gammamatrix[i, j] := gamma[i, j] :$

end do:

$Gammamatrix[i, m] := 1 - add(gamma[i, j], j = 1 ..m - 1) :$

end do:

> $Gammamatrix$

$$\begin{bmatrix} \gamma_{1,1} & \gamma_{1,2} & 1 - \gamma_{1,1} - \gamma_{1,2} \\ \gamma_{2,1} & \gamma_{2,2} & 1 - \gamma_{2,1} - \gamma_{2,2} \\ \gamma_{3,1} & \gamma_{3,2} & 1 - \gamma_{3,1} - \gamma_{3,2} \end{bmatrix} \quad (9)$$

> $Pyvector := \left\langle seq\left(\frac{\exp(-\lambda[j]) \cdot \lambda[j]^{x[t]}}{x[t]!}, j = 1 ..m\right) \right\rangle :$

> $Pmatrix := DiagonalMatrix(Pyvector);$

$$Pmatrix := \begin{bmatrix} \frac{e^{-\lambda_1} \lambda_1^{x_t}}{x_t!} & 0 & 0 \\ 0 & \frac{e^{-\lambda_2} \lambda_2^{x_t}}{x_t!} & 0 \\ 0 & 0 & \frac{e^{-\lambda_3} \lambda_3^{x_t}}{x_t!} \end{bmatrix} \quad (10)$$

> $T := 8 :$

> $kappa := eval(HMMexsum(deltavector, Gammamatrix, Pmatrix, 1, T), \{x[1]=1, x[2]=2, x[3]=1, x[4]=2, x[5]=1, x[6]=1, x[7]=1, x[8]=1\}) :$

> $indets(kappa)$

$$\{\delta_1, \delta_2, \gamma_{1,1}, \gamma_{1,2}, \gamma_{2,1}, \gamma_{2,2}, \gamma_{3,1}, \gamma_{3,2}, \lambda_1, \lambda_2, \lambda_3, e^{-\lambda_1}, e^{-\lambda_2}, e^{-\lambda_3}\} \quad (11)$$

> $pars := \langle \gamma_{1,1}, \gamma_{1,2}, \gamma_{2,1}, \gamma_{2,2}, \gamma_{3,1}, \gamma_{3,2}, \lambda_1, \lambda_2, \lambda_3 \rangle :$

> $D1 := Dmat(kappa, pars) :$

>

> $indetpars := \langle \delta_1, \delta_2, \gamma_{1,1}, \gamma_{1,2}, \gamma_{2,1}, \gamma_{2,2}, \gamma_{3,1}, \gamma_{3,2}, \lambda_1, \lambda_2, \lambda_3 \rangle :$

> $Hybrid(D1, indetpars, 0.0, 1.0, 1)$

$$\begin{array}{c}
8 \\
\left\{ \begin{array}{l}
\left[\begin{array}{l}
-0.0710036094303968 \\
-0.259613765489724 \\
0.878434532541952 \\
-0.0525664381999851 \\
0.213494380567544 \\
-0.321742326441289 \\
-0.0448803938078468 \\
0.0301987837126429 \\
0.0335340183250897
\end{array} \right]
\end{array} \right. \\
7 \\
\left\{ \begin{array}{l}
\left[\begin{array}{l}
-0.240210944722951 \\
0.131340349510768 \\
-0.265675348649239 \\
0.358236409408899 \\
0.361523646941555 \\
-0.419070196651291 \\
-0.0472510500910235 \\
0.00105395426216690 \\
0.646203334211498
\end{array} \right] , \left[\begin{array}{l}
0.377739285499986 \\
-0.383169934550079 \\
0.346694185304037 \\
-0.381742706304043 \\
0.467834688209981 \\
-0.472801842878501 \\
0.0414571842441634 \\
0.0197330669391829 \\
0.00710713224621105
\end{array} \right]
\end{array} \right. \\
8 \\
\left\{ \left[\begin{array}{l}
0.325707227015121 \\
-0.276643510834436 \\
0.449827173127426 \\
-0.292492990172497 \\
-0.343932228786572 \\
0.635550407030166 \\
-0.0427947743955878 \\
-0.0371410560593432 \\
0.0637337160722460
\end{array} \right] \right. \\
8 \\
\left\{ \left[\begin{array}{l}
0.361174493038883 \\
-0.531185076532133 \\
0.106639486683413 \\
-0.311495563901396 \\
0.414490819268559 \\
-0.458609301489467 \\
0.0611094835309849
\end{array} \right] \right.
\end{array}$$

(12)

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>
> T := 9 :
> kappa := eval(HMMexsum(deltavector, Gammamatrix, Pmatrix, 1, T), {x[1]=1, x[2]=2, x[3]
= 1, x[4]=2, x[5]=1, x[6]=1, x[7]=1, x[8]=1, x[9]=1}) :
> indets(kappa)

```

$$\{\delta_1, \delta_2, \gamma_{1,1}, \gamma_{1,2}, \gamma_{2,1}, \gamma_{2,2}, \gamma_{3,1}, \gamma_{3,2}, \lambda_1, \lambda_2, \lambda_3, e^{-\lambda_1}, e^{-\lambda_2}, e^{-\lambda_3}\}$$

(13)

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> pars := <\gamma_{1,1}, \gamma_{1,2}, \gamma_{2,1}, \gamma_{2,2}, \gamma_{3,1}, \gamma_{3,2}, \lambda_1, \lambda_2, \lambda_3> :
> D1 := Dmat(kappa, pars) :
>
>
> indetpars := <\delta_1, \delta_2, \gamma_{1,1}, \gamma_{1,2}, \gamma_{2,1}, \gamma_{2,2}, \gamma_{3,1}, \gamma_{3,2}, \lambda_1, \lambda_2, \lambda_3> :
> Hybrid(D1, indetpars, 0.0, 1.0, 1)

```

$$\begin{array}{c}
8 \\
9 \\
8 \\
9 \\
9
\end{array}
\left\{
\begin{array}{c}
\left[\begin{array}{c}
-0.0302880862737410 \\
-0.293803354961962 \\
-0.194504522870175 \\
-0.742910050099361 \\
-0.126494471217161 \\
-0.488484014945349 \\
0.0337356773472763 \\
-0.0636263568925426 \\
-0.251417929281599
\end{array} \right] \\
\emptyset \\
\left[\begin{array}{c}
0.0129185307853816 \\
-0.0651981497258576 \\
0.178687754086673 \\
-0.424313561663699 \\
-0.404233387851155 \\
0.670964402619706 \\
-0.0496659457670418 \\
-0.0410341591469710 \\
-0.407262361310401
\end{array} \right] \\
\emptyset \\
\emptyset
\end{array}
\right.$$

(14)

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>
> # Setting up matrices and vector that define the model
> m := 2 :
> deltavector := <delta[1]1 - delta[1]>;

```

$$deltavector := \begin{bmatrix} \delta_1 & 1 - \delta_1 \end{bmatrix} \quad (15)$$

```

> Gammamatrix := Matrix(m, m) :
for i from 1 to m do
  for j from 1 to m - 1 do
    Gammamatrix[i, j] := gamma[i, j] :
  end do:
  Gammamatrix[i, m] := 1 - add(gamma[i, j], j = 1 .. m - 1) :
end do:

```

```

> Gammamatrix := eval(Gammamatrix, {gamma1,1 = 1 - c*pi[1, 2], gamma2,1 = c*pi[2, 1]});

```

$$Gammamatrix := \begin{bmatrix} -c\pi_{1,2} + 1 & c\pi_{1,2} \\ c\pi_{2,1} & -c\pi_{2,1} + 1 \end{bmatrix} \quad (16)$$

```

> Pyvector := <seq(exp(-lambda[j]) * lambda[j]x[t], j = 1 .. m)> :

```

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> Pmatrix := DiagonalMatrix(Pyvector);

```

$$Pmatrix := \begin{bmatrix} e^{-\lambda_1} \lambda_1^{x_t} & 0 \\ 0 & e^{-\lambda_2} \lambda_2^{x_t} \end{bmatrix} \quad (17)$$

```

> # When there are 3 data points:

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```

> T := 3 :

```

```

> kappa := HMMexsum(deltavector, Gammamatrix, Pmatrix, 1, T);

```

$$\kappa := \begin{bmatrix} \left((\delta_1 (-c\pi_{1,2} + 1) + (1 - \delta_1) c\pi_{2,1}) e^{-\lambda_1} \lambda_1^{x_1} + (\delta_1 c\pi_{1,2} + (1 - \delta_1) (-c\pi_{2,1} + 1)) e^{-\lambda_2} \lambda_2^{x_1} \right) \\ \left[\left((\delta_1 (-c\pi_{1,2} + 1) + (1 - \delta_1) c\pi_{2,1}) e^{-\lambda_1} \lambda_1^{x_1} (-c\pi_{1,2} + 1) + (\delta_1 c\pi_{1,2} + (1 - \delta_1) (-c\pi_{2,1} + 1)) e^{-\lambda_2} \lambda_2^{x_1} c\pi_{2,1} \right) e^{-\lambda_1} \lambda_1^{x_2} + \left((\delta_1 (-c\pi_{1,2} + 1) + (1 - \delta_1) c\pi_{2,1}) e^{-\lambda_1} \lambda_1^{x_1} c\pi_{1,2} + (\delta_1 c\pi_{1,2} + (1 - \delta_1) (-c\pi_{2,1} + 1)) e^{-\lambda_2} \lambda_2^{x_1} (-c\pi_{2,1} + 1) \right) e^{-\lambda_2} \lambda_2^{x_2} \right] \\ \left[\left((\delta_1 (-c\pi_{1,2} + 1) + (1 - \delta_1) c\pi_{2,1}) e^{-\lambda_1} \lambda_1^{x_1} (-c\pi_{1,2} + 1) + (\delta_1 c\pi_{1,2} + (1 - \delta_1) (-c\pi_{2,1} + 1)) e^{-\lambda_2} \lambda_2^{x_1} c\pi_{2,1} \right) e^{-\lambda_1} \lambda_1^{x_2} (-c\pi_{1,2} + 1) + \left((\delta_1 (-c\pi_{1,2} + 1) + (1 - \delta_1) c\pi_{2,1}) e^{-\lambda_1} \lambda_1^{x_1} c\pi_{1,2} + (\delta_1 c\pi_{1,2} + (1 - \delta_1) (-c\pi_{2,1} + 1)) e^{-\lambda_2} \lambda_2^{x_1} (-c\pi_{2,1} + 1) \right) e^{-\lambda_2} \lambda_2^{x_2} c\pi_{2,1} \right] e^{-\lambda_1} \lambda_1^{x_3} + \left[\left((\delta_1 (-c\pi_{1,2} + 1) + (1 - \delta_1) c\pi_{2,1}) e^{-\lambda_1} \lambda_1^{x_1} (-c\pi_{1,2} + 1) + (\delta_1 c\pi_{1,2} + (1 - \delta_1) (-c\pi_{2,1} + 1)) e^{-\lambda_2} \lambda_2^{x_1} c\pi_{2,1} \right) e^{-\lambda_1} \lambda_1^{x_2} (-c\pi_{1,2} + 1) + \left((\delta_1 (-c\pi_{1,2} + 1) + (1 - \delta_1) c\pi_{2,1}) e^{-\lambda_1} \lambda_1^{x_1} c\pi_{1,2} + (\delta_1 c\pi_{1,2} + (1 - \delta_1) (-c\pi_{2,1} + 1)) e^{-\lambda_2} \lambda_2^{x_1} (-c\pi_{2,1} + 1) \right) e^{-\lambda_2} \lambda_2^{x_2} c\pi_{2,1} \right] e^{-\lambda_1} \lambda_1^{x_3} \end{bmatrix} \quad (18)$$

$$\begin{aligned}
& -c\pi_{1,2} + 1) + (\delta_1 c\pi_{1,2} + (1 - \delta_1)(-c\pi_{2,1} + 1)) e^{-\lambda_2 x_1} \lambda_2^x c\pi_{2,1} \Big) e^{-\lambda_1 x_2} \lambda_1^x c\pi_{1,2} \\
& + \left((\delta_1(-c\pi_{1,2} + 1) + (1 - \delta_1)c\pi_{2,1}) e^{-\lambda_1 x_1} \lambda_1^x c\pi_{1,2} + (\delta_1 c\pi_{1,2} + (1 - \delta_1)(-c\pi_{2,1} \right. \\
& \left. + 1)) e^{-\lambda_2 x_1} \lambda_2^x (-c\pi_{2,1} + 1) \Big) e^{-\lambda_2 x_2} \lambda_2^x (-c\pi_{2,1} + 1) \Big) e^{-\lambda_2 x_3} \lambda_2^x \Big]
\end{aligned}$$

> *indets(kappa)*

$$\left\{ c, \delta_1, \lambda_1, \lambda_2, \pi_{1,2}, \pi_{2,1}, x_1, x_2, x_3, \lambda_1^{x_1}, \lambda_1^{x_2}, \lambda_1^{x_3}, \lambda_2^{x_1}, \lambda_2^{x_2}, \lambda_2^{x_3}, e^{-\lambda_1}, e^{-\lambda_2} \right\} \quad (19)$$

> *pars := <c, λ₁, λ₂, pi_{1,2}, pi_{2,1}>;*

$$\text{pars} := \begin{bmatrix} c \\ \lambda_1 \\ \lambda_2 \\ \pi_{1,2} \\ \pi_{2,1} \end{bmatrix} \quad (20)$$

> *D1 := Dmat(kappa, pars) :*

> *r := Rank(D1); d := Dimension(pars) - r;*

$$r := 3$$

$$d := 2$$

(21)

> *T := 4 :*

> *kappa := eval(HMMexsum(deltavector, Gammamatrix, Pmatrix, 1, T), delta[1] = 1/2) :*

> *D1 := Dmat(kappa, pars) :*

> *r := Rank(D1); d := Dimension(pars) - r;*

$$r := 4$$

$$d := 1$$

(22)

> *Estpar(D1, pars, 1);*

$$\left[\left\{ f(c, \lambda_1, \lambda_2, \pi_{1,2}, \pi_{2,1}) = -FI(\lambda_1, \lambda_2, c\pi_{1,2}, c\pi_{2,1}) \right\} \right], \quad (23)$$

$$\left[\left[\left[\begin{array}{cccc} -\frac{c}{\pi_{2,1}} & 0 & 0 & \frac{\pi_{1,2}}{\pi_{2,1}} & 1 \end{array} \right] \right] \right],$$

$$\left[\left[\left[\left[-\frac{\left(\frac{\partial}{\partial c} f(c, \lambda_1, \lambda_2, \pi_{1,2}, \pi_{2,1}) \right) c}{\pi_{2,1}} + \frac{\left(\frac{\partial}{\partial \pi_{1,2}} f(c, \lambda_1, \lambda_2, \pi_{1,2}, \pi_{2,1}) \right) \pi_{1,2}}{\pi_{2,1}} \right. \right. \right. \right. \\ \left. \left. \left. \left. + \frac{\partial}{\partial \pi_{2,1}} f(c, \lambda_1, \lambda_2, \pi_{1,2}, \pi_{2,1}) \right] \right] \right] \right]$$

> # Reparameterisation:

> $m := 2$:

> $deltavector := \langle \text{delta}[1] | 1 - \text{delta}[1] \rangle$;

$$deltavector := \begin{bmatrix} \delta_1 & 1 - \delta_1 \end{bmatrix} \quad (24)$$

> $Gammamatrix := \text{Matrix}(m, m)$:

for i **from** 1 **to** m **do**

for j **from** 1 **to** $m - 1$ **do**

$Gammamatrix[i, j] := \text{gamma}[i, j]$:

end do:

$Gammamatrix[i, m] := 1 - \text{add}(\text{gamma}[i, j], j = 1 .. m - 1)$:

end do:

> $Gammamatrix := \text{eval}(Gammamatrix, \{\gamma_{1,1} = 1 - \text{beta}[1], \gamma_{2,1} = \text{beta}[2]\})$;

$$Gammamatrix := \begin{bmatrix} 1 - \beta_1 & \beta_1 \\ \beta_2 & 1 - \beta_2 \end{bmatrix} \quad (25)$$

> $Pyvector := \langle \text{seq}(\exp(-\text{lambda}[j]) \cdot \text{lambda}[j]^{x[t]}, j = 1 .. m) \rangle$:

> $Pmatrix := \text{DiagonalMatrix}(Pyvector)$;

$$Pmatrix := \begin{bmatrix} e^{-\lambda_1} \lambda_1^{x_t} & 0 \\ 0 & e^{-\lambda_2} \lambda_2^{x_t} \end{bmatrix} \quad (26)$$

> $T := 4$:

> $\text{kappa} := \text{eval}\left(\text{HMMexsum}(deltavector, Gammamatrix, Pmatrix, 1, T), \text{delta}[1] = \frac{1}{2}\right)$:

> $\text{pars} := \langle \beta_1, \beta_2, \lambda_1, \lambda_2 \rangle$:

> $D1 := \text{Dmat}(\text{kappa}, \text{pars})$:

> $r := \text{Rank}(D1)$; $d := \text{Dimension}(\text{pars}) - r$;

$r := 4$

$d := 0$

(27)

> $\text{kappa}[1]$

(28)

$$\left(\frac{1}{2} - \frac{\beta_1}{2} + \frac{\beta_2}{2} \right) e^{-\lambda_1 x_1} + \left(\frac{\beta_1}{2} + \frac{1}{2} - \frac{\beta_2}{2} \right) e^{-\lambda_2 x_2} \quad (28)$$

> kappa[2]

$$\left(\left(\frac{1}{2} - \frac{\beta_1}{2} + \frac{\beta_2}{2} \right) e^{-\lambda_1 x_1} (1 - \beta_1) + \left(\frac{\beta_1}{2} + \frac{1}{2} - \frac{\beta_2}{2} \right) e^{-\lambda_2 x_1} \beta_2 \right) e^{-\lambda_1 x_2} + \left(\frac{1}{2} - \frac{\beta_1}{2} + \frac{\beta_2}{2} \right) e^{-\lambda_1 x_1} \beta_1 + \left(\frac{\beta_1}{2} + \frac{1}{2} - \frac{\beta_2}{2} \right) e^{-\lambda_2 x_1} (1 - \beta_2) \right) e^{-\lambda_2 x_2} \quad (29)$$