Computer algebra derives normal forms of general stochastic and non-autonomous differential equations

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Abstract

Modelling stochastic systems has many important applications. Stochastic coordinate transforms to a normal form is a powerful way of disentangling emergent long term dynamics. Since the analysis involves classic calculus, then the approach also applies to a wide class of non-autonomous dynamical systems. Further, cater for deterministic autonomous systems by simply omitting the time dependence in the system. For generality, this approach now caters for unstable modes, and for differential equation systems with a rational right-hand side. Use this code via the website[1].

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[1] http://www.maths.adelaide.edu.au/anthony.roberts/sdenf.php

1 Introduction

Construct stochastic normal form of a wide class of non-autonomous or stochastic differential equations (SDEs), based upon earlier research (Cox & Roberts 1991, Chao & Roberts 1996, Roberts 2008). Interpret all SDEs in the Stratonovich sense so the analysis applies to deterministic differential equations, both non-autonomous and autonomous. To construct normal forms of deterministic autonomous differential equations, simply omit specifying any noise. This article documents code designed for an interactive web site (Roberts 2009) that is available to all to use.

In the following, assign `thecase:=webpage;` for the web service (or to read a system from file `sdeb.red`), otherwise assign `thecase` to be any of the example dynamical systems in set `thecases`.

```
%% Execute in reduce with  in_tex "cadnfgsde.tex"$
%% See cadnfgsde.pdf for detailed explanation.
thecase:=sdemmh;
thecases:={sdesimple, oderat, sdemulti, sdebif, sdelac,
sdemom, sderadek, sdehyper, sdeduan, sdehifour, sdeMona1,
sdeMona2, sdeMona3, sdeMona4, sdeMonaSS, sdeMajda3m,
sdeMajda3a, sdePRLorenz, sdehe, sdehebc, sdeheqr, sdemmh,
```

```
sdePRKdV }$
```

Define default parameters for the iteration: `maxiter_` is the maximum number of allowed iterations; `toosmall` is the order of errors in the analysis in terms of the parameter `small`. Specific problems may override these defaults. The code cannot handle any higher order in noise amplitude `sigma`.

```
maxiter_:=29$
factor small;
toosmall:=3$
let sigma^3=>0;
```

2 Choose the SDEs

For an SDE with m slow modes, n_y fast stable modes (quickly decaying), and/or n_z fast unstable modes (quickly growing), you must denote the slow modes by `x(1)` through to `x(m)`, the stable fast modes by `y(1)` through to `y(ny)`, and the unstable fast modes by `z(1)` through to `z(nz)`. Each Stratonovich white noise, derivative of a Stratonovich Wiener process, must be denoted by `w(.)` where the dot denotes almost any label you care to choose: simple numbers such as `w(1)` and/or `w(2)` are the usual choices; but other labels for the noise can be used. The SDEs must be linearly diagonalised.[2]

Load the reduce to LaTeX package so we can also generate a nicer version of the output via LaTeX. Load now so that a canny definition of the SDEs can invoke some of the options.

```
load_package rlfi;
```

First, define the operators to be used in the specification of the SDEs.

```
operator x;
operator y;
operator z;
operator w;
```

[2] Although a Jordan form is also acceptable, there are issues in the error control.

Cater for rational function SDEs by allowing time dependence in these variables at specification. Then users must multiply each SDE by a common denominator, and put on the right-hand side the nonlinear terms involving the time derivative: see examples in sections 2.2 and 2.11.2.

```
depend x,t;
depend y,t;
depend z,t;
```

Use the RHS part of an SDE to specify an SDE. Form the RHSs into lists to specify the system of SDEs. Set trivial defaults for the SDEs in case I forget.

```
xrhs:=yrhs:=zrhs:={}$
```

In the case `webpage`, get the SDEs from the file `sdeb.red` which is where the web script writes a user specified system.

```
if thecase=webpage then in "sdeb.red"$
```

2.1 Simple pair of SDEs

A classically simple pair of fast/slow SDEs is

$$\dot{x} = -xy \quad \text{and} \quad \dot{y} = -y + x^2 - 2y^2 + \sigma w(t), \tag{1}$$

where lowercase $w(t)$ denotes the formal derivative dW/dt of a Stratonovich Wiener process $W(t, \omega)$. Parameter σ controls the strength of the noise. Use `x(1)` to denote variable x, `y(1)` to denote variable y, and `w(1)` to denote Stratonovich noise w. Alternatively, $w(t)$ could denote some non-autonomous deterministic forcing, control or other extrinsic input.

Specify the slow x SDE by allocating the one element list of its right-hand side to the variable `xrhs`.

```
if thecase=sdesimple then begin
xrhs:={-x(1)*y(1)};
```

Specify the nonlinear and noise terms of the y SDE as the one element list assigned to `yrhs`.

```
yrhs:={-y(1)+x(1)^2-2*y(1)^2+w(1)};
```

There are no unstable modes.

```
zrhs:={};
```

The code automatically multiplies the noise factors by a parameter `sigma` so there is no need include the parameter σ in the specification of the problem (unless you particularly want to), as it will be done for you. The code uses the parameter `small` to control truncation in nonlinearity; just ignore parameter `small` wherever it appears.

```
factor small,sigma;
end;
```

The stochastic coordinate transform

$$y_1 = Y_1 + X_1^2 + 2Y_1^2 + \sigma e^{-t}\star w_1\,(1 + 4Y_1)$$

$$x_1 = X_1 + X_1 Y_1 + \sigma e^{-t}\star w_1\, X_1$$

Result normal form SDEs

$$\dot{Y}_1 = -Y_1 - 2X_1^2 Y_1 - 4\sigma w_1 Y_1 + 8\sigma^2 w_1 e^{-t}\star w_1\, Y_1$$

$$\dot{X}_1 = -X_1^3 - \sigma w_1 X_1 + 2\sigma^2 w_1 e^{-t}\star w_1\, X_1$$

Easy stochastic bifurcation For example, modifying the x equation to $\dot{x} = ax - xy$ induces a stochastic bifurcation as parameter a crosses zero. Construct the stochastic normal form for this by simply changing to

```
    xrhs:={small*a*x(1)-x(1)*y(1)};
```

The extra factor of `small` causes only low powers of parameter a to be retained, as appropriate for such a bifurcation. Some may like to consider it analogous to scaling a with ϵ^2 and scaling x with ϵ.

2.2 Simple rational ODEs

A simple system of fast/slow ODEs in rational functions is

$$\dot{x} = -\frac{xy}{1+z}, \quad \dot{y} = -\frac{y}{1+2y} + x^2, \quad \dot{z} = 2\frac{z}{1+3x}. \tag{2}$$

Use `x(1)` to denote variable x, `y(1)` to denote variable y, and `z(1)` to denote z. Multiply each ODE by the denominator for the ODE and shift the nonlinear `d/dt` terms to the right-hand side.

```
if thecase=oderat then begin
xrhs:={-x(1)*y(1)-z(1)*df(x(1),t)};
yrhs:={-y(1)+x(1)^2*(1+2*y(1))-2*y(1)*df(y(1),t)};
zrhs:={2*z(1)-3*x(1)*df(z(1),t)};
```

Truncate to one higher order because it is a simple system.

```
toosmall:=4;
end;
```

Time dependent coordinate transform

$$z_1 = 6X_1Y_1Z_1\varepsilon^2 + Z_1$$

$$y_1 = 2X_1^4\varepsilon^2 - 4X_1^2Y_1^2\varepsilon^2 + X_1^2\varepsilon + 6Y_1^3\varepsilon^2 - 2Y_1^2\varepsilon + Y_1$$

$$x_1 = 2X_1^3Y_1\varepsilon^2 - 1/2X_1Y_1^2\varepsilon^2 + X_1Y_1Z_1\varepsilon^2 + X_1Y_1\varepsilon + X_1$$

Result normal form DEs For example, from this form we see the slow manifold is $Y_1 = Z_1 = 0$.

$$\dot{Z}_1 = -54X_1^3Z_1\varepsilon^3 + 18X_1^2Z_1\varepsilon^2 - 6X_1Z_1\varepsilon + 2Z_1$$

$$\dot{Y}_1 = 8X_1^4Y_1\varepsilon^3 + 4X_1^2Y_1\varepsilon^2 + 2X_1^2Y_1\varepsilon - Y_1$$

$$\dot{X}_1 = -2X_1^5\varepsilon^3 - X_1^3\varepsilon^2 - 2X_1Y_1^2Z_1\varepsilon^3$$

2.3 Future noise in the transform

An interesting pair of fast/slow SDEs derived from stochastic advection/dispersion is

$$\dot{x} = -\sigma y w(t) \quad \text{and} \quad \dot{y} = -y + \sigma x w(t)\,, \tag{3}$$

where lowercase $w(t)$ denotes the formal derivative dW/dt of a Stratonovich Wiener process $W(t,\omega)$. Parameter σ controls the strength of the noise. In stochastic advection/dispersion parameter σ represents the lateral wavenumber of the concentration profile.

Use `x(1)` to denote variable x, `y(1)` to denote variable y, and `w(1)` to denote Stratonovich noise w. Specify the slow x SDE by allocating the one element list of its right-hand side to the variable `xrhs`. Specify the nonlinear and noise terms of the y SDE as the one element list assigned to `yrhs`.

```
if thecase=sdemulti then begin
factor small,sigma;
xrhs:={-y(1)*w(1)};
yrhs:={-y(1)+x(1)*w(1)};
zrhs:={};
end;
```

The code automatically multiplies the noise factors by a parameter `sigma` so there is no need include the parameter σ in the specification of the problem (unless you particularly want to), as it will be done for you.

Via the coordinate transform

$$x \approx X + \sigma Y e^{t}\star w \quad \text{and} \quad y \approx Y + \sigma X e^{-t}\star w\,,$$

the resultant normal form is

$$\dot{X} \approx -\sigma^2 X w e^{-t}\star w \quad \text{and} \quad \dot{Y} \approx -Y + \sigma^2 Y w e^{t}\star w\,.$$

One of the interesting aspects of this example is the quickness with which we could go to higher order noise interactions, higher orders in σ. However, I do not compute such higher order terms in this code.

2.4 Other methodologies fail

Consider for small parameter ϵ

$$\begin{aligned} \text{slow mode} \quad \dot x &= \epsilon x + x^3 - (1-\sigma w)xy\,, && (4)\\ \text{fast mode} \quad \dot y &= -y + x^2 + y^2 + \sigma y w\,. && (5) \end{aligned}$$

Deterministically, there is a bifurcation to two equilibria for small $\epsilon > 0$. The noise w affects this bifurcation somehow.

Why is this tricky? Cross-sectional averaging is simply projection onto the slow space $y = 0$ which predicts instability of subcritical bifurcation $\dot x = \epsilon x + x^3$. Whereas adiabatic approximation, singular perturbation and multiple scales set $\dot y = 0$ whence $y \approx x^2$ and thus predict linear growth of $\dot x = \epsilon x$. Normal form transforms get the deterministic dynamics correctly. But what happens for stochastic dynamics?

Multiply a nonlinear term in the x SDE in order to get cancellation when the right-hand sides are multiplied by `small`. Multiply the bifurcation parameter by `small^2` in order to make small.

```
if thecase=sdebif then begin
xrhs:={small*eps*x(1)+small*x(1)^3-x(1)*y(1)*(1-small*w(1))};
```

Insert the noise in a rather special way so that its dominant effects cancel.

```
yrhs:={-y(1)+x(1)^2+y(1)^2+y(1)*w(1)};
```

There are no unstable modes.

```
zrhs:={};
```

Truncate to higher order in the amplitudes in order to discern the subtle bifurcation.

```
toosmall:=5;
factor small,sigma;
end;
```

The coordinate transform is very messy, but dominantly (dropping subscripts for simplicity)

$$\begin{aligned} x &\approx X + XY + 2X^3Y \\ &\quad + \sigma\Big[(-XY^2 + 3X^3Y)e^{+t}\star + (XY^2 + XY^3)e^{2t}\star \\ &\qquad + X^3e^{-t}\star - XY^3e^{3t}\star\Big]w\,, \qquad (6) \\ y &\approx Y - Y^2 + Y^3 - Y^4 + X^2 - 7X^2Y^2 + X^4 \\ &\quad + \sigma\Big[(-Y + 2Y^3 - 3Y^4 - 10X^2Y^2 - 4X^2Y^2e^{+t}\star)e^{+t}\star \\ &\qquad + (X^2 - 2X^2Y + 3X^2Y^2 + X^4 + 4X^4e^{-t}\star)e^{-t}\star + 2X^2Y^2e^{2t}\star\Big]w\,. \qquad (7) \end{aligned}$$

In these coordinates the slow mode SDE becomes the normal form

$$\dot X \approx \epsilon X - X^5 - 2\sigma X^5 w - 3\sigma^2X^5we^{-t}\star w\,, \qquad (8)$$

$$\dot Y \approx (-1 + 4X^2 + 6X^4)Y + \sigma(1 + 2X^2 + 22X^4)Yw\,. \qquad (9)$$

The 'drift' of the quadratic noise in $\dot X$ should also be nonlinearly stabilising.

2.5 Levy area contraction: off-diagonal example

Pavliotis & Stuart (2008) assert the following system of five coupled SDEs are interesting.

$$dx_1 = \epsilon y_1\,dt\,, \qquad (10)$$

$$dx_2 = \epsilon y_2\,dt\,, \qquad (11)$$

$$dx_3 = \epsilon(x_1y_2 - x_2y_1)dt\,, \qquad (12)$$

$$dy_1 = (-y_1 - \alpha y_2)dt + dW_1\,, \qquad (13)$$

$$dy_2 = (+\alpha y_1 - y_2)dt + dW_2\,. \qquad (14)$$

This stochastic system has two noise sources. I presume $W_i(t,\omega)$ are Stratonovich Wiener processes. Use `x(i)` to denote variable x_i, `y(i)` to denote variable y_i, and `w(i)` to denote noise dW_i/dt.

Let `eps` denote parameter ϵ. Thus specify the slow dynamics via this three component list allocated to `xrhs`.

```
if thecase=sdelac then begin
factor small,sigma,eps;
toosmall:=4;
xrhs:={eps*y(1),eps*y(2),eps*(x(1)*y(2)-x(2)*y(1))};
```

For the fast modes, specify the linear parts separately from the rest of the SDE. Here the linear part has an off-diagonal component parametrised by α. This code cannot exactly analyse such systems. Thus analyse with the α moderated terms when treated as a perturbation of the decay at rate one. Specify the dynamics of the y SDE as the two element list assigned to `yrhs`.

```
yrhs:={-y(1)-a*y(2)+w(1),-y(2)+a*y(1)+w(2)};
```

There are no unstable modes.

```
zrhs:={};
end;
```

The stochastic normal form is

$$\begin{aligned}
\dot X_1 &\approx \epsilon\sigma w_1 - \epsilon\sigma w_2 a\,,\\
\dot X_2 &\approx \epsilon\sigma w_2 + \epsilon\sigma w_1 a\,,\\
\dot X_3 &\approx \epsilon\sigma(-w_1X_2 + w_2X_1) + \epsilon\sigma(w_1X_1a + w_2X_2a)\\
&\quad + \epsilon^2\sigma^2(w_1e^{-t}\star w_2 - w_2e^{-t}\star w_1)\,,\\
\dot Y_1 &\approx -Y_1 - Y_2a\,,\\
\dot Y_2 &\approx -Y_2 + Y_1a\,.
\end{aligned}$$

2.6 Position-momentum: the Jordan form

Suppose you want to analyse the semi-mechanical system of SDEs

$$\ddot x = -xy \quad\text{and}\quad \dot y = -2y + x^2 + \dot x^2 + \sigma w(t)\,, \tag{15}$$

where $w(t)$ denotes the formal derivative dW/dt of a Stratonovich Wiener process $W(t,\omega)$. Parameter σ controls the strength of the noise. Introduce $x_1 = x$, $x_2 = \dot x$ and $y_1 = y$ to convert to the system of three coupled SDEs

$$\dot x_1 = x_2\,, \tag{16}$$

$$\dot{x}_2 = -x_1 y_1\,, \tag{17}$$
$$\dot{y}_1 = -2y_1 + x_1^2 + x_2^2 + \sigma w(t)\,. \tag{18}$$

Specify the slow x SDEs by allocating the two element list of its right-hand side to the variable `xrhs`. Divide the x_1 term by `small` on the right-hand side of $\dot{x}_2$ in order to overcome the automatic multiplication of the right-hand side by `small`. Iteration still works for this Jordan form system: but I am not responsible for anyone who divides by `small` or `sigma`.

```
if thecase=sdemom then begin
xrhs:={x(2)/small,-x(1)*y(1)};
yrhs:={-2*y(1)+x(1)^2+x(2)^2+w(y)};
factor small,sigma;
```

Note: the code automatically multiply the noise factors by a parameter `sigma` so there is no need include the parameter σ in the specification of the problem.

There are no unstable modes.

```
zrhs:={};
end;
```

The resultant normal form is

$$\begin{aligned}
\dot{X}_1 &\approx X_2 + \sigma(\tfrac{1}{4}wX_1 + \tfrac{1}{4}wX_2) \\
&\quad + \sigma^2(-\tfrac{3}{32}wX_1e^{-2t}\star w - \tfrac{3}{64}wX_2e^{-2t}\star w) \\
\dot{X}_2 &\approx \sigma(-\tfrac{1}{2}wX_1 - \tfrac{1}{4}wX_2) + (-\tfrac{1}{2}X_1^3 + \tfrac{1}{2}X_2X_1^2 - \tfrac{3}{4}X_2^2X_1) \\
&\quad + \sigma^2(\tfrac{1}{8}wX_1e^{-2t}\star w + \tfrac{3}{32}wX_2e^{-2t}\star w) \\
\dot{Y}_1 &\approx -2Y_1 + (-\tfrac{1}{2}X_1^2Y_1 + \tfrac{1}{2}X_2X_1Y_1 + \tfrac{1}{2}X_2^2Y_1)
\end{aligned}$$

2.7 Radek's slow oscillation with fast noise

Consider Radek's system

$$\dot{x} = -\epsilon xz\,, \quad \dot{y} = +\epsilon yz \quad \text{and} \quad \dot{z} = -(z-1) + \sigma w(t)\,.$$

Transform to our standard form by

$$x = x_1\,, \quad y = x_2 \quad \text{and} \quad z = 1 + y_1\,.$$

Then obtain the stochastic normal form with the following code.

```
if thecase=sderadek then begin
yrhs:={-y(1)+w(1)};
xrhs:={-eps*x(2)*(1+y(1)),eps*x(1)*(1+y(1))};
factor small,sigma,eps;
zrhs:={};
end;
```

The normal form equations are

$$\begin{aligned}\dot X_1 &\approx -\epsilon(1+\sigma w)X_2\,,\\ \dot X_2 &\approx +\epsilon(1+\sigma w)X_1\,,\\ \dot Y_1 &\approx -Y_1\,.\end{aligned}$$

The dynamics clearly oscillate in (X_1, X_2) with phase angle $\theta = \epsilon(t + \sigma W(t,\omega))$. This normal form arises from the coordinate transform

$$\begin{aligned}x_1 &\approx X_1 + \epsilon X_2 Y_1 + \epsilon\sigma X_2 e^{-t}\star w - \tfrac12\epsilon^2 X_1 Y_1^2\,,\\ x_2 &\approx X_2 - \epsilon X_1 Y_1 - \epsilon\sigma X_1 e^{-t}\star w - \tfrac12\epsilon^2 X_2 Y_1^2\,,\\ y_1 &= Y_1 + \sigma e^{-t}\star w\,.\end{aligned}$$

Different analysis has to be done for the case of fast oscillation in this system.

2.8 Simple hyperbolic system

$$\begin{aligned}\dot y_1 &= w_1 z_1 \sigma - y_1\\ \dot z_1 &= w_1 y_1 \sigma + z_1\end{aligned}$$

```
if thecase=sdehyper then begin
yrhs:={-y(1)+z(1)*w(1)};
xrhs:={};
zrhs:={+z(1)+y(1)*w(1)};
end;
```

The stochastic coordinate transform $z_1 = -e^{2t}{\star}w_1\, Y_1\sigma + Z_1$

$y_1 = e^{-2t}{\star}w_1\, Z_1\sigma + Y_1$

Result normal form SDEs $\dot Z_1 = e^{-2t}{\star}w_1\, w_1 Z_1\sigma^2 + Z_1$

$\dot Y_1 = -e^{2t}{\star}w_1\, w_1 Y_1\sigma^2 - Y_1$

2.9 Duan's hyperbolic system for foliation

Used as an example by Sun et al. (2011).

$\dot y_1 = w_1 y_1\sigma - y_1$

$\dot z_1 = w_1 z_1\sigma + y_1^2\varepsilon + z_1$

```
if thecase=sdeduan then begin
yrhs:={-y(1)+y(1)*w(1)};
xrhs:={};
zrhs:={+z(1)+y(1)^2+z(1)*w(1)};
end;
```

The stochastic coordinate transform

$z_1 = -1/3e^{3t}{\star}w_1\, Y_1^2\sigma - 1/3Y_1^2 + Z_1$

$y_1 = Y_1$

Result normal form SDEs

$\dot Z_1 = w_1 Z_1\sigma + Z_1$

$\dot Y_1 = w_1 Y_1\sigma - Y_1$

2.10 A four mode hyperbolic system

$$\dot{x}_1 = w_2 y_1 \sigma + w_1 z_1 \sigma + y_1 z_2 \varepsilon$$

$$\dot{y}_1 = w_1 z_1 \sigma - y_1$$

$$\dot{z}_1 = w_1 y_1 \sigma + z_1$$

$$\dot{z}_2 = w_2 y_1 \sigma + 2z_2$$

```
if thecase=sdehifour then begin
yrhs:={-y(1)+z(1)*w(1)};
xrhs:={y(1)*w(2)+z(1)*w(1)+y(1)*z(2)};
zrhs:={+z(1)+y(1)*w(1)
,+2*z(2)+y(1)*w(2)};
end;
```

The stochastic coordinate transform

$$z_1 = -e^{2t}\star w_1 Y_1 \sigma + Z_1$$

$$z_2 = -e^{3t}\star w_2 Y_1 \sigma + Z_2$$

$$y_1 = e^{-2t}\star w_1 Z_1 \sigma + Y_1$$

$$x_1 = -e^{3t}\star w_2 Y_1^2 \sigma + e^{2t}\star w_2 Y_1^2 \sigma - e^{t}\star w_2 Y_1 \sigma + e^{-1t}\star w_1 Z_1 \sigma + e^{-2t}\star w_1 Z_2 Z_1 \sigma - e^{-3t}\star w_1 Z_2 Z_1 \sigma + X_1 + Y_1 Z_2$$

Result normal form SDEs

$$\dot{Z}_1 = e^{-2t}\star w_1 w_1 Z_1 \sigma^2 + Z_1$$

$$\dot{Z}_2 = 2Z_2$$

$$\dot{Y}_1 = -e^{2t}\star w_1 w_1 Y_1 \sigma^2 - Y_1$$

$$\dot{X}_1 = e^{3t}\star w_2 w_1 Y_1 Z_1 \sigma^2 - 2e^{-2t}\star w_1 w_2 Y_1 Z_1 \sigma^2$$

2.11 Monahan's four examples

Monahan & Culina (2011) discuss stochastic averaging and give four examples in an appendix which I also analyse here. They also give a couple of interesting examples in the body of the article which I may also explore at some time. I contend that they really need my approach as "a large separation often does not exist in atmosphere or ocean dynamics" between the fast and slow time scales.

2.11.1 Example four: 'three' time scales

Monahan & Culina (2011) comment that this linear system has three time scales. But I do not see that: I only see varying strength interactions. And as a linear system, it is the simplest. They consider

$$\frac{dx}{dt} = -x + \frac{a}{\sqrt{\tau}}y \quad \text{and} \quad \frac{dy}{dt} = \frac{1}{\sqrt{\tau}}x - \frac{1}{\tau}y + \frac{b}{\sqrt{\tau}}\dot{W}.$$

Let $\tau = \epsilon^2$ and rescale time, $t = \tau t' = \epsilon^2 t'$ so that $d/dt = \frac{1}{\tau}d/dt'$ and $\dot{W} = \frac{1}{\sqrt{\tau}}dW/dt'$. Then, dropping dashes, the SDE system is

$$\frac{dx}{dt} = -\epsilon^2 x + a\epsilon y \quad \text{and} \quad \frac{dy}{dt} = \epsilon x - y + b\dot{W}.$$

Using the default inbuilt parametrisation of noise by `sigma` to represent parameter b, and using `small` in the x-SDE so that it counts the numbers of small ϵ, code these as the following.

```
if thecase=sdeMona4 then begin
xrhs:={eps*a*y(1)-eps^2*small*x(1)};
yrhs:={eps*x(1)-y(1)+w(1)};
zrhs:={ };
toosmall:=4;
factor small,sigma,eps,yy,y,w,ou;
end;
```

The stochastic coordinate transform

$$y_1 = -e^{-1t}\star e^{-1t}\star w_1\ \epsilon^2\sigma a - e^{-1t}\star w_1\ \epsilon^2\sigma a + e^{-1t}\star w_1\ \sigma + Y_1 + \epsilon X_1$$

$$x_1 = -e^{-1t}\star w_1\ \epsilon\sigma a - Y_1\epsilon a + X_1$$

Result normal form SDEs

$$\dot{Y}_1 = -Y_1\epsilon^2 a - Y_1$$

$$\dot{X}_1 = w_1\epsilon^3\sigma\big(-2a^2 + a\big) + w_1\epsilon\sigma a + \epsilon^2\big(X_1 a - X_1\big)$$

Monahan & Culina (2011) derive the last two terms in the X-equation, but not the first as it is too small for their averaging analysis. They comment that $a > 1$ is some sort of difficulty; but I have no problem with $a > 1$ (until X growth invalidates the linearity), especially as the decay rate to the stochastic slow manifold, the Y-SDE, is $(1 + \epsilon a)$ which gets stronger with parameter a.

2.11.2 Example one: simple nonlinear

Monahan & Culina (2011) first consider the example

$$\frac{dx}{dt} = -x + \Sigma(x)y \quad \text{and} \quad \frac{dy}{dt} = -\frac{1}{\tau}y + \frac{1}{\sqrt{\tau}}\dot{W},$$

for general smooth functions $\Sigma(x)$. Rescale time, $t = \tau t'$ so that $d/dt = \frac{1}{\tau}d/dt'$ and $\dot{W} = \frac{1}{\sqrt{\tau}}dW/dt'$. Then, dropping dashes, the SDE is

$$\frac{dx}{dt} = -\tau x + \tau\Sigma(x)y \quad \text{and} \quad \frac{dy}{dt} = -y + \dot{W}.$$

Restricting to a rational function $\Sigma = (a_0 + a_1x + a_2x^2)/(1 + b_1x + b_2x^2)$, code these as the following (multiply through by the denominator).

```
if thecase=sdeMona1 then begin
operator a; defindex a(down);
operator b; defindex b(down);
```

```
xrhs:={-tau*x(1)*(1+b(1)*x(1)+b(2)*x(1)^2)
  -(b(1)*x(1)+b(2)*x(1)^2)*df(x(1),t)
  +tau*y(1)*(a(0)+a(1)*x(1)+a(2)*x(1)^2) };
yrhs:={-y(1)+w(1)};
zrhs:={ };
toosmall:=3;
factor small,sigma,tau,yy,y,w,ou;
end;
```

The stochastic coordinate transform In the following expressions, recall that σ parametrises the noise; for comparison with the modelling of Monahan & Culina (2011), take $\sigma = 1$. Also recall that a rational represents the function Σ.

$$y_1 = e^{-1t}\star w_1\,\sigma + Y_1$$

$$x_1 = e^{-1t}\star w_1\,\sigma\varepsilon\tau\big(-a_2X_1^2 - a_1X_1 - a_0\big) + Y_1\varepsilon\tau\big(-a_2X_1^2 - a_1X_1 - a_0\big) + X_1$$

Result normal form SDEs

$$\dot Y_1 = -Y_1$$

$$\dot X_1 = w_1\sigma\varepsilon^2\tau^2\big(a_2b_2X_1^4 - a_2X_1^2 + 2a_1b_2X_1^3 + a_1b_1X_1^2 + 3a_0b_2X_1^2 + 2a_0b_1X_1 + a_0\big) + w_1\sigma\varepsilon^2\tau\big(-a_2b_2X_1^4 - a_2b_1X_1^3 - a_1b_2X_1^3 - a_1b_1X_1^2 - a_0b_2X_1^2 - a_0b_1X_1\big) + w_1\sigma\varepsilon\tau\big(a_2X_1^2 + a_1X_1 + a_0\big) + \varepsilon^2\tau\big(b_2^2X_1^5 + 2b_2b_1X_1^4 + b_2X_1^3 + b_1^2X_1^3 + b_1X_1^2\big) + \varepsilon\tau\big(-b_2X_1^3 - b_1X_1^2 - X_1\big)$$

Monahan & Culina (2011) derive some of this X equation. The others here are higher order terms that become significant at finite parameter values. For example, the next correction to their analysis, $w_1\tau^2(-3a_4X_1^4 - 2a_3X_1^3 - a_2X_1^2 + a_0)$, is probably derivable as $\tau^2(\Sigma - x\Sigma')\dot W$ (when rescaled).

2.11.3 Example three: many fast modes

Monahan & Culina (2011) third considered the example

$$\frac{dx}{dt} = -x + \Sigma(x)\|\vec{y}\| \quad \text{and} \quad \frac{d\vec{y}}{dt} = -\frac{1}{\tau}\vec{y} + \sqrt{\frac{2}{\tau}}\dot{\vec{W}},$$

for general smooth functions $\Sigma(x)$. As before, rescale time, $t = \tau t'$ so that $d/dt = \frac{1}{\tau}d/dt'$ and $\dot{W} = \frac{1}{\sqrt{\tau}}dW/dt'$. Here I also cheat: they have $\|\vec{y}\|$ in the slow equation; but $\|\vec{y}\|$ is not a smooth multinomial and so my generic program cannot apply; instead I replace $\|\vec{y}\|$ with $\|\vec{y}\|^2$ which has the same symmetry but is multinomial. Then, upon the rescaling of time, and dropping dashes, the SDE is

$$\frac{dx}{dt} = -\tau x + \tau\Sigma(x)\|\vec{y}\|^2 \quad \text{and} \quad \frac{d\vec{y}}{dt} = -\vec{y} + \sigma\dot{\vec{W}}.$$

Restricting to the general quartic $\Sigma = a_0 + a_1 x + \cdots + a_4 x^4$, code these as the following (the generic program automatically inserts the σ in the noise). Currently restrict to just a two component $\vec{y}$ as I do not see any reason for any more and Monahan & Culina (2011) do not appear to specify.

```
if thecase=sdeMona3 then begin
operator a; defindex a(down);
xrhs:={-tau*x(1)+tau*(y(1)^2+y(2)^2)
*(a(0)+a(1)*x(1)+a(2)*x(1)^2+a(3)*x(1)^3+a(4)*x(1)^4) };
yrhs:={-y(1)+w(1),-y(2)+w(2)};
zrhs:={ };
toosmall:=3;
factor small,sigma,tau,yy,y,w,ou;
end;
```

The stochastic coordinate transform In the following expressions, recall that σ parametrises the noise; for comparison with the modelling of Monahan & Culina (2011), you might take $\sigma = \sqrt{2}$ but as I changed the SDE somewhat an exact comparison is not possible. Also recall that a general quartic represents the function Σ.

$y_1 = e^{-1t}\star w_1\,\sigma + Y_1$

$y_2 = e^{-1t}\star w_2\,\sigma + Y_2$

$x_1 = e^{t}\star w_2\,Y_2\sigma\tau\big(-a_4X_1^4 - a_3X_1^3 - a_2X_1^2 - a_1X_1 - a_0\big) + e^{-1t}\star w_2\,Y_2\sigma\tau\big(-a_4X_1^4 - a_3X_1^3 - a_2X_1^2 - a_1X_1 - a_0\big) + e^{t}\star w_1\,Y_1\sigma\tau\big(-a_4X_1^4 - a_3X_1^3 - a_2X_1^2 - a_1X_1 - a_0\big) + e^{-1t}\star w_1\,Y_1\sigma\tau\big(-a_4X_1^4 - a_3X_1^3 - a_2X_1^2 - a_1X_1 - a_0\big) + Y_2^2\tau\big(-1/2a_4X_1^4 - 1/2a_3X_1^3 - 1/2a_2X_1^2 - 1/2a_1X_1 - 1/2a_0\big) + Y_1^2\tau\big(-1/2a_4X_1^4 - 1/2a_3X_1^3 - 1/2a_2X_1^2 - 1/2a_1X_1 - 1/2a_0\big) + X_1$

Result normal form SDEs

$\dot{Y}_1 = -Y_1$

$\dot{Y}_2 = -Y_2$

$\dot{X}_1 = e^{-1t}\star w_2\,w_2\sigma^2\tau^2\big(-3/2a_4X_1^4 - a_3X_1^3 - 1/2a_2X_1^2 + 1/2a_0\big) + e^{-1t}\star w_2\,w_2\sigma^2\tau\big(a_4X_1^4 + a_3X_1^3 + a_2X_1^2 + a_1X_1 + a_0\big) + e^{-1t}\star w_1\,w_1\sigma^2\tau^2\big(-3/2a_4X_1^4 - a_3X_1^3 - 1/2a_2X_1^2 + 1/2a_0\big) + e^{-1t}\star w_1\,w_1\sigma^2\tau\big(a_4X_1^4 + a_3X_1^3 + a_2X_1^2 + a_1X_1 + a_0\big) - \tau X_1$

In this modelling for X, the three terms linear in τ are the leading order, and rewritten are

$$\dot{X} \approx -\tau X + \tau\Sigma(X)\sigma^2(w_1e^{-t}\star w_1 + w_2e^{-t}\star w_2)$$

Such quadratic terms in the noise generate both fluctuations and mean drift (Chao & Roberts 1996): the mean drift effect from each is $\sigma^2/2$, so their mean sum is just σ^2. Hence the mean part of this model reduces to the form of (A33) (Monahan & Culina 2011).

The fluctuations in $w_1e^{-t}\star w_1$ are skewed, on finite times, and so should contribute to the skewness commented on by Monahan & Culina (2011). A generalisation of the Fokker–Planck analysis of Chao & Roberts (1996) suggests that such skewness decays algebraically in the scale separation τ: such algebraic decay in τ makes skewness much more noticeable at finite τ than other modelling approximations which decay exponentially in the scale separation.

2.11.4 Example two: irregular slow manifold

Monahan & Culina (2011) second consider the example

$$\frac{dx}{dt} = x - x^3 + \Sigma(x)y \quad \text{and} \quad \frac{dy}{dt} = -\frac{1}{x\tau}y + \frac{1}{\sqrt{\tau}}\dot{W},$$

for general smooth functions $\Sigma(x)$. Since the y-dynamics are at least exponentially unstable for negative x, we restrict attention to $x > 0$. Even for positive x the system is singular as $x \to 0$ so the slow manifold is irregular in some sense (although 'singular' in a good way in that the scale separation between fast and slow becomes infinite). Here I think we have to be more sophisticated in rescaling time: let's choose the new fast time t' so that $dt = x\tau\,dt'$; that is, $t' = \int(x\tau)^{-1}dt$ which would not be explicitly known until after a solution $x(t')$ is found. I presume that the noise then transforms as $\dot{W} = \frac{1}{\sqrt{x\tau}}dW/dt'$ (needs checking). Then, dropping dashes, the SDE is

$$\frac{dx}{dt} = \tau\left[x^2 - x^4 + x\Sigma(x)y\right] \quad \text{and} \quad \frac{dy}{dt} = -y + \sqrt{x}\dot{W}.$$

Now the $\sqrt{x}$ is a problem in my generic computer algebra which requires multinomial systems so transform to $x = \xi^2$ so that $2\xi\,d\xi = dx$. Then the SDE system takes on a multinomial form

$$\frac{d\xi}{dt} = \frac{1}{2}\tau\left[\xi^3 - \xi^7 + \xi\Sigma(\xi^2)y\right] \quad \text{and} \quad \frac{dy}{dt} = -y + \xi\dot{W}.$$

The Stratonovich and Ito versions of the above SDE are still the same (an unresolved question is whether the non-uniform time scaling introduces a difference). Restricting to the general linear $\Sigma = a_0 + a_1 x$, code the SDE system as the following (remember $\mathtt{x(1)} = \xi = \sqrt{x}$).

```
if thecase=sdeMona2 then begin
operator a; defindex a(down);
xrhs := {1/2*tau*(x(1)^3-x(1)^7+x(1)*(a(0)+a(1)*x(1)^2)*y(1))}$
yrhs := { -y(1)+x(1)*w(1) }$
zrhs := {}$
factor small,sigma,tau,yy,y,w,ou;
end;
```

The stochastic coordinate transform In the following expressions, recall that σ parametrises the noise; for comparison with the modelling of Monahan & Culina (2011), take $\sigma = 1$.

$$y_1 = e^{-1t}\star e^{-1t}\star w_1\ \sigma\tau\big(1/2X_1^7 - 1/2X_1^3\big) + e^{-1t}\star w_1\,\sigma X_1 + Y_1$$

$$x_1 = e^{-1t}\star w_1\ \sigma\tau\big(-1/2a_1X_1^4 - 1/2a_0X_1^2\big) + Y_1\tau\big(-1/2a_1X_1^3 - 1/2a_0X_1\big) + X_1$$

Result normal form SDEs

$$\dot Y_1 = e^{t}\star w_1\, w_1 Y_1\sigma^2\tau^2\big(1/4a_1^2X_1^6 + 1/2a_1a_0X_1^4 + 1/4a_0^2X_1^2\big) + e^{-1t}\star w_1\, w_1Y_1\sigma^2\tau^2\big(3/4a_1^2X_1^6 + a_1a_0X_1^4 + 1/4a_0^2X_1^2\big) + w_1Y_1\sigma\tau^2\big(-a_1X_1^9 - 3/2a_0X_1^7 + 1/2a_0X_1^3\big) + w_1Y_1\sigma\tau\big(-1/2a_1X_1^3 - 1/2a_0X_1\big) - Y_1$$

$$\dot X_1 = e^{-1t}\star w_1\, w_1\sigma^2\tau^2\big(-1/4a_1^2X_1^7 - 1/2a_1a_0X_1^5 - 1/4a_0^2X_1^3\big) + w_1\sigma\tau^2\big(a_1X_1^{10} + 3/2a_0X_1^8 - 1/2a_0X_1^4\big) + w_1\sigma\tau\big(1/2a_1X_1^4 + 1/2a_0X_1^2\big) + \tau\big(-1/2X_1^7 + 1/2X_1^3\big)$$

Using just the leading order terms, the ones linear in τ, and recalling $X_1 \approx \xi = \sqrt{x}$, the last SDE gives the model

$$\frac{dx}{dt'} \approx \tau\left[x^2 - x^4 + x^{3/2}\Sigma(x)\sigma\frac{dW}{dt'}\right].$$

But recall that $dt' = dt/(x\tau)$ (although one should be more careful as $X_1 \approx \sqrt{x}$, not exact equality) and similarly $dW/dt' = \sqrt{x\tau}\dot W$ so that this model becomes

$$\frac{dx}{dt} \approx x - x^3 + \sqrt{\tau}x\Sigma(x)\sigma\frac{dW}{dt}.$$

This agrees with the Stratonovich part of (A28) by Monahan & Culina (2011). But again, the above derivation has the systematic higher order corrections that are needed for finite scale separation τ (such corrections appear to be of the same order as the difference between the Ito and Stratonovich versions of this model).

2.11.5 Idealised Stommel-like model of meridional overturning circulation

Monahan & Culina (2011) also analyse the Idealised Stommel-like model

$$\frac{dx}{dt} = \mu - |y - x|x + \sigma_A \dot{W}_1 ,$$
$$\frac{dy}{dt} = +\frac{1}{\tau}(1 - y) - |y - x|y + \sqrt{\frac{2}{\tau}}\sigma_M \dot{W}_2 .$$

The mod-functions do not fit into my generic computer algebra so replace them with squares to at least preserve the symmetry. As before, introduce $\epsilon^2 = \tau$ and rescale time, $t = \tau t' = \epsilon^2 t'$ so that $d/dt = \frac{1}{\tau} d/dt'$ and $\dot{W}_j = \frac{1}{\sqrt{\tau}} dW_j/dt' = \frac{1}{\epsilon} dW_j/dt'$. Since for small τ, the fast variable y is strongly attracted to one, change the reference point for y by setting $y = 1 + y'(t)$. Then the SDEs becomes akin to

$$\frac{dx}{dt'} = \epsilon^2 \left[\mu - (1 + y' - x)^2 x\right] + \epsilon\sigma_A \frac{dW_1}{dt'} ,$$
$$\frac{dy'}{dt'} = -y' - \epsilon^2(1 + y' - x)^2(1 + y') + \sqrt{2}\sigma_M \frac{dW_2}{dt'} .$$

Let $\rho = \sigma_A/(\sqrt{2}\sigma_M)$, use the inbuilt $\sigma = \sqrt{2}\sigma_M$, and invoke `small` to correctly count the number of small ϵs in the analysis. Code the above dynamics as the following.

```
if thecase=sdeMonaSS then begin
xrhs := {small*eps^2*(mu-(1+y(1)-x(1))^2*x(1))
    +small*eps*rho*w(1)}$
yrhs := { -y(1)-small*eps^2*(1+y(1)-x(1))^2*(1+y(1))+w(2) }$
zrhs := {}$
factor small,sigma,eps,rho,yy,y,w,ou;
toosmall:=4;
end;
```

The stochastic coordinate transform

$y_1 = e^{-1t}\star e^{-1t}\star w_2\ \epsilon^2\sigma\big(-X_1^2+4X_1-3\big)+3/2e^{t}\star w_2\,Y_1^2\epsilon^2\sigma+3/2e^{-1t}\star w_2\,Y_1^2\epsilon^2\sigma+e^{-1t}\star w_2\,Y_1\epsilon^2\sigma\big(-4X_1+6\big)+e^{-1t}\star w_2\,\sigma+1/2Y_1^3\epsilon^2+Y_1^2\epsilon^2\big(-2X_1+3\big)+Y_1+\epsilon^2\big(-X_1^2+2X_1-1\big)$

$x_1 = e^{t}\star w_2\,Y_1\epsilon^2\sigma X_1+e^{-1t}\star w_2\,Y_1\epsilon^2\sigma X_1+e^{-1t}\star w_2\,\epsilon^2\sigma\big(-2X_1^2+2X_1\big)+1/2Y_1^2\epsilon^2X_1+Y_1\epsilon^2\big(-2X_1^2+2X_1\big)+X_1$

Result normal form SDEs

$\dot Y_1 = -3e^{-1t}\star w_2\,w_2Y_1\epsilon^2\sigma^2+4e^{-1t}\star w_2\,w_1Y_1\epsilon^3\rho\sigma^2+w_2Y_1\epsilon^2\sigma\big(4X_1-6\big)+Y_1\epsilon^2\big(-X_1^2+4X_1-3\big)-Y_1$

$\dot X_1 = -e^{-1t}\star w_2\,w_2\epsilon^2\sigma^2X_1+e^{-1t}\star w_2\,w_1\epsilon^3\rho\sigma^2\big(4X_1-2\big)+w_2\epsilon^2\sigma\big(2X_1^2-2X_1\big)+w_1\epsilon\rho\sigma+\epsilon^2\big(-X_1^3+2X_1^2-X_1+\mu\big)$

Deterministically, this model has multiple equilibria for small μ, but only one equilibria for $\mu > 4/27$, at finite amplitude. The noise $\dot W_1$ causes transitions between such multiple equilibria, and the multiplicative noise $\dot W_2$ contributes as well. But the same order of smallness is the first term in the X_1 SDE above which is a quadratic noise that has a mean drift effect that should enhance the stability of the small x equilibrium.

2.12 Majda's triad models

Majda et al. (2002) investigated averaging in two 3D SDE systems. I also looked at these in 2003.[3] Let's look at the stochastic normal form.

2.12.1 Multiplicative triad model

The multiplicative triad model of Majda et al. (2002) consists of three modes, v_1, v_2 and v_3. These evolve in time according to

$$\frac{dv_1}{dt} = b_1v_2v_3\,,\quad \frac{dv_2}{dt} = b_2v_1v_3\,,\quad \frac{dv_3}{dt} = -v_3+b_3v_1v_2+\sigma\dot W\,, \qquad (19)$$

[3] *Centre manifold analysis of stochastic multiplicative triad model*, technical report.

where b_i and σ are some constants and the noise forces the third mode. Here I have already scaled the equations so that the rate of decay of the third mode is one. Thus on long time scales we expect the third mode to be essentially negligible and the system to be modelled by the relatively slow evolution of the first two modes.

```
if thecase=sdeMajda3m then begin
operator b; defindex b(down);
xrhs := {b(1)*x(2)*y(1),b(2)*x(1)*y(1)}$
yrhs := { -y(1)+b(3)*x(1)*x(2)+w(3) }$
zrhs := {}$
factor small,sigma,yy,y,w,ou;
toosmall:=5;
end;
```

The stochastic coordinate transform

$$y_1 = e^{-1t}\star e^{-1t}\star w_3\ \sigma\big(-b_3b_2X_1^2-b_3b_1X_2^2\big)-4e^{-1t}\star w_3\,Y_1\sigma b_3b_2b_1X_2X_1+e^{-1t}\star w_3\,\sigma\big(-b_3b_2X_1^2-b_3b_1X_2^2+1\big)-2Y_1^2b_3b_2b_1X_2X_1+Y_1-b_3^2b_2X_2X_1^3-b_3^2b_1X_2^3X_1+b_3X_2X_1$$

$$x_1 = e^{-1t}\star e^{-1t}\star w_3\ \sigma\big(b_3b_2b_1X_2X_1^2+b_3b_1^2X_2^3\big)-1/2e^{-1t}\star w_3\,Y_1^2\sigma b_2b_1^2X_2+e^{-1t}\star w_3\,Y_1\sigma b \ e^{-1t}\star w_3\,\sigma\big(2b_3b_2b_1X_2X_1^2+2b_3b_1^2X_2^3-b_1X_2\big)-1/6Y_1^3b_2b_1^2X_2+1/2Y_1^2b_2b_1X_1+Y_1\left(b_3b_2b_1X_2X_1^2+b_3b_1^2X_2^3-b_1X_2\right)+X_1$$

$$x_2 = e^{-1t}\star e^{-1t}\star w_3\ \sigma\big(b_3b_2^2X_1^3+b_3b_2b_1X_2^2X_1\big)-1/2e^{-1t}\star w_3\,Y_1^2\sigma b_2^2b_1X_1+e^{-1t}\star w_3\,Y_1\sigma b \ e^{-1t}\star w_3\,\sigma\big(2b_3b_2^2X_1^3+2b_3b_2b_1X_2^2X_1-b_2X_1\big)-1/6Y_1^3b_2^2b_1X_1+1/2Y_1^2b_2b_1X_2+Y_1\left(b_3b_2^2X_1^3+b_3b_2b_1X_2^2X_1-b_2X_1\right)+X_2$$

Result normal form SDEs

$$\dot Y_1 = e^{-1t}\star w_3\,w_3Y_1\sigma^2\big(2b_3b_2^2b_1X_1^2+2b_3b_2b_1^2X_2^2\big)+4w_3Y_1\sigma b_3b_2b_1X_2X_1+Y_1\big(b_3^2b_2^2X_1^4+2b_3^2b_2b_1X_2^2X_1^2+b_3^2b_1^2X_2^4-b_3b_2X_1^2-b_3b_1X_2^2-1\big)$$

$$\dot X_1 = -2e^{-1t}\star w_3\,w_3\sigma^2b_3b_2b_1^2X_2^2X_1+w_3\sigma\big(-2b_3b_2b_1X_2X_1^2-2b_3b_1^2X_2^3+b_1X_2\big)-b_3^2b_2b_1X_2^2X_1^3-b_3^2b_1^2X_2^4X_1+b_3b_1X_2^2X_1$$

$\dot{X}_2 = -2e^{-1t}\star w_3\, w_3\sigma^2 b_3 b_2^2 b_1 X_2 X_1^2 + w_3\sigma\big(-2b_3 b_2^2 X_1^3 - 2b_3 b_2 b_1 X_2^2 X_1 + b_2 X_1\big) - b_3^2 b_2^2 X_2 X_1^4 - b_3^2 b_2 b_1 X_2^3 X_1^2 + b_3 b_2 X_2 X_1^2$

Majda et al. (2002) predicts, their equation (52), the two leading order terms in the deterministic part and the linear noise part. I suspect their first term in each equation is an Ito version of my Stratonovich modelling. All the higher order terms here are missed by their averaging.

2.12.2 Additive triad model

The additive triad model of Majda et al. (2002) consists of three modes, ν_1, ν_2 and ν_3, as before. However, these now evolve in time according to

$$\begin{aligned}
\frac{d\nu_1}{dt} &= b_1\nu_2\nu_3\,, \\
\frac{d\nu_2}{dt} &= -\nu_2 + b_2\nu_1\nu_3 + \sigma_2\dot{W}_2\,, \\
\frac{d\nu_3}{dt} &= -\nu_3 + b_3\nu_1\nu_2 + \sigma_3\dot{W}_3\,,
\end{aligned} \tag{20}$$

where b_i and σ_i are some constants, and there is independent stochastic forcing of the second and third modes. Here I have already scaled the equations so that the rate of decay of *both* the second and third mode is one.[4] Thus on long time scales we expect the second and third modes to be essentially negligible and the system to be modelled by the relatively slow evolution of the first mode. This section constructs the stochastic normal form of its centre manifold model as the basis for a model over long time scales with new noise processes.

```
if thecase=sdeMajda3a then begin
operator b; defindex b(down);
xrhs := {b(1)*y(2)*y(1)}$
```

[4] In contrast, Majda et al. (2002) set the two modes to have different decay rates. I do not expect much difference in using the same decay rate, it is just more convenient that the memory convolutions are then identical for the two modes rather than being different. Having the decay rates the same is also closer to my expected application to spatial problems.

```
yrhs := { -y(1)+b(2)*x(1)*y(2)+b(21)*w(2)
          ,-y(2)+b(3)*x(1)*y(1)+b(31)*w(3) }$
zrhs := {}$
factor small,sigma,yy,y,xx,x;
toosmall:=3;
end;
```

The stochastic coordinate transform

$$y_1 = X_1\sigma b_{31} b_2 e^{-1t}\star e^{-1t}\star w_3 \ + Y_1 + \sigma b_{21} e^{-1t}\star w_2$$

$$y_2 = X_1\sigma b_{21} b_3 e^{-1t}\star e^{-1t}\star w_2 \ + Y_2 + \sigma b_{31} e^{-1t}\star w_3$$

$$x_1 = X_1 - 1/2 Y_2 Y_1 b_1 + Y_2\sigma\big(- 1/2 b_{21} b_1 e^{t}\star w_2 - 1/2 b_{21} b_1 e^{-1t}\star w_2 \big) + Y_1\sigma\big(- 1/2 b_{31} b_1 e^{t}\star w_3 - 1/2 b_{31} b_1 e^{-1t}\star w_3 \big)$$

Result normal form SDEs

$$\dot Y_1 = X_1 Y_2 b_2 + Y_2\sigma^2\big(- 1/2 b_{31} b_{21} b_2 b_1 e^{-1t}\star e^{-1t}\star w_3\ w_2 - 1/2 b_{31} b_{21} b_2 b_1 e^{-1t}\star w_3\, w_2 - 1/2 b_{31} b_{21} b_2 b_1 e^{-1t}\star w_2\, w_3\big) + Y_1\sigma^2\big(- 1/2 b_{31}^2 b_2 b_1 e^{-1t}\star e^{-1t}\star w_3\ w_3 - 1/2 b_{31}^2 b_2 b_1 e^{-1t}\star w_3\, w_3\big) - Y_1$$

$$\dot Y_2 = X_1 Y_1 b_3 + Y_2\sigma^2\big(- 1/2 b_{21}^2 b_3 b_1 e^{-1t}\star e^{-1t}\star w_2\ w_2 - 1/2 b_{21}^2 b_3 b_1 e^{-1t}\star w_2\, w_2\big) - Y_2 + Y_1\sigma^2\big(- 1/2 b_{31} b_{21} b_3 b_1 e^{-1t}\star e^{-1t}\star w_2\ w_3 - 1/2 b_{31} b_{21} b_3 b_1 e^{-1t}\star w_3\, w_2 - 1/2 b_{31} b_{21} b_3 b_1 e^{-1t}\star w_2\, w_3\big)$$

$$\dot X_1 = X_1\sigma^2\big(1/2 b_{31}^2 b_2 b_1 e^{-1t}\star e^{-1t}\star w_3\ w_3 + 1/2 b_{31}^2 b_2 b_1 e^{-1t}\star w_3\, w_3 + 1/2 b_{21}^2 b_3 b_1 e^{-1t}\star e^{-1t}\star w_2\ w_2 + 1/2 b_{21}^2 b_3 b_1 e^{-1t}\star w_2\, w_2\big) + \sigma^2\big(1/2 b_{31} b_{21} b_1 e^{-1t}\star w_3\, w_2 + 1/2 b_{31} b_{21} b_1 e^{-1t}\star w_2\, w_3\big)$$

The only terms in the model are the quadratic noise-noise interaction terms. Majda et al. (2002) recognise the last, σ^2 term, but not the first, $X_1\sigma^2$ term. They represent it as a mean drift and independent noise (presumably the mean drift comes from the Ito representation of my Stratonovich noise).

2.13 Potzsche and Rasmussen deterministic non-autonomous examples

Potzsche & Rasmussen (2006) establish Taylor approximations of various integral manifolds of non-autonomous systems. They give two examples.

2.13.1 Lorenz near the pitchfork bifurcation

Example 5.1 of Potzsche & Rasmussen (2006) is

$$\begin{aligned}\dot{x}_1 &= \sigma_\epsilon(x_2 - x_1),\\ \dot{x}_2 &= \rho_\epsilon x_1 - x_2 - x_1x_3\,,\\ \dot{x}_3 &= -\beta_\epsilon x_3 + x_1x_2\,.\end{aligned}$$

where parameters are $\sigma_\epsilon = \sigma_0 + \epsilon\sigma(t)$, $\rho_\epsilon = 1 + \rho_0 + \epsilon\rho(t)$ and $\beta_\epsilon = \beta_0 + \epsilon\beta(t)$. When there is no parametric fluctuations, $\epsilon = 0$, there is a pitchfork bifurcation as ρ_0 crosses zero.

To analyses dynamics at this pitchfork bifurcation in the presence of fluctuations, Potzsche & Rasmussen (2006) take a linear transform of the system to variables $\vec y$ and set $\rho_0 = 0$. In the following coding I use `x(1)` $= y_3$, `y(1)` $= y_1$ and `y(2)` $= y_2$; there are no unstable modes. Also the fluctuations $\epsilon\rho(t)$ are represented in the input by `w(rho)` whereas in the output it is represented by σw_ρ, and similarly for the other fluctuating quantities. Note that the algorithm automatically multiplies time varying quantities by the 'small' parameter σ (distinct from the σ in the Lorenz system!) corresponding to their ϵ.

```
if thecase=sdePRLorenz then begin
sig0:=1; bet0:=1;
sig1:=sig0/(sig0+1);
xrhs := {sig0*sig1*y(1)*y(2)-sig1*x(1)*y(2)
    +sig1*x(1)*w(rho)
    +(w(sigma)-w(rho)/(sig0+1))*y(1)}$
yrhs := { -(sig0+1)*y(1)+sig1*y(1)*y(2)-x(1)*y(2)/(sig0+1)
    +w(rho)/(sig0+1)*x(1)
```

```
      -(w(sigma)+w(rho)/(sig0+1))*y(1)
      ,-bet0*y(2)-sig0*y(1)^2+(1-sig0)*x(1)*y(1)+x(1)^2
      -w(beta)*y(2)
      }$
zrhs := {}$
factor small,sigma,yy,y,xx,x;
toosmall:=3;
end;
```

For the general computer algebra I set σ_0 and β_0 to some definite values, here $\sigma_0 = \beta_0 = 1$.

The specified dynamical system

$$\dot{x}_1 = -1/2x_1y_2\varepsilon + 1/2x_1\sigma w_\rho + 1/2y_2y_1\varepsilon + y_1\sigma\big(-1/2w_\rho + w_\sigma\big)$$

$$\dot{y}_1 = -1/2x_1y_2\varepsilon + 1/2x_1\sigma w_\rho + 1/2y_2y_1\varepsilon + y_1\sigma\big(-1/2w_\rho - w_\sigma\big) - 2y_1$$

$$\dot{y}_2 = x_1^2\varepsilon - y_2\sigma w_\beta - y_2 - y_1^2\varepsilon$$

The stochastic coordinate transform

$$y_1 = X_1Y_2\sigma\big(-1/2e^{-1t}\star w_\beta + e^{-1t}\star w_\rho - 1/4e^{-2t}\star w_\rho + 1/2e^{-1t}\star w_\sigma\big) - 1/2X_1Y_2 + 1/2X_1\sigma e^{-2t}\star w_\rho + Y_2Y_1\sigma\big(1/2e^{1t}\star w_\beta - 1/4e^{2t}\star w_\rho + 1/3e^{t}\star w_\rho + 1/2e^{2t}\star w_\sigma - 1/2e^{t}\star w_\sigma\big) - 1/2Y_2Y_1 + Y_1$$

$$y_2 = X_1^2\sigma\big(-e^{-1t}\star w_\beta - e^{-1t}\star w_\rho\big) + X_1^2 + X_1Y_1\sigma\big(e^{2t}\star w_\rho - 2/3e^{t}\star w_\rho + 1/3e^{-2t}\star w_\rho - 2e^{2t}\star w_\sigma + 2e^{1t}\star w_\sigma\big) + Y_2 + Y_1^2\sigma\big(1/3e^{3t}\star w_\beta - 1/3e^{3t}\star w_\rho - 2/3e^{3t}\star w_\sigma\big) + 1/3Y_1^2$$

$$x_1 = X_1Y_2\sigma\big(-1/2e^{t}\star w_\beta - 1/3e^{t}\star w_\rho - 1/12e^{-2t}\star w_\rho + 1/2e^{t}\star w_\sigma\big) + 1/2X_1Y_2 + X_1 + Y_2Y_1\sigma\big(1/6e^{3t}\star w_\beta - 1/3e^{3t}\star w_\rho + 1/4e^{2t}\star w_\rho + 7/6e^{3t}\star w_\sigma - 1/2e^{2t}\star w_\sigma\big) - 1/6Y_2Y_1 + Y_1\sigma\big(1/2e^{2t}\star w_\rho - e^{2t}\star w_\sigma\big)$$

Result normal form SDEs

$$\dot Y_1 = X_1^2Y_1\sigma^2\big(1/2e^{2t}\star e^{2t}\star w_\rho\, w_\rho + 1/4e^{-2t}\star e^{-2t}\star w_\rho\, w_\rho - 1/2e^{-2t}\star e^{-2t}\star w_\rho\, w_\sigma - e^{2t}\star e^{2t}\star w_\sigma\, w_\rho + 1/12e^{2t}\star w_\beta\, w_\rho + 1/2e^{-1t}\star w_\beta\, w_\beta + 1/3e^{-1t}\star w_\beta\, w_\rho - 1/2e^{-1t}\star w_\beta\, w_\sigma \cdots 3/4e^{2t}\star w_\rho\, w_\beta + 13/48e^{2t}\star w_\rho\, w_\rho + 3/8e^{2t}\star w_\rho\, w_\sigma + 1/3e^{t}\star w_\rho\, w_\beta - 2/3e^{t}\star w_\rho\, w_\rho - 1/3e^{t}\star w_\rho\, w_\sigma + 1/2e^{-1t}\star w_\rho\, w_\beta + 1/3e^{-1t}\star w_\rho\, w_\rho - 1/2e^{-1t}\star w_\rho\, w_\sigma - 1/6e^{-2t}\star w_\rho\, w_\beta + 7/12e^{-2t}\star w_\rho\, w_\rho - 7/12e^{-2t}\star w_\rho\, w_\sigma + 3/2e^{2t}\star w_\sigma\, w_\beta - 3/8e^{2t}\star w_\sigma\, w_\rho - 3/4e^{2t}\star w_\sigma\, w_\sigma - e^{t}\star w_\sigma\, w_\beta + 2e^{t}\star w_\sigma\, w_\rho + e^{t}\star w_\sigma\, w_\sigma\big) + X_1^2Y_1\sigma\big(-1/2w_\beta - 3/4w_\rho - 1/4w_\sigma\big) + 1/2X_1^2Y_1 + X_1Y_2^2\sigma^2\big(-1/8e^{2t}\star e^{2t}\star w_\rho\, w_\rho + 1/4e^{2t}\star e^{2t}\star w_\sigma\, w_\rho - 1/12e^{2t}\star w_\beta\, w_\rho + 1/4e^{t}\star w_\beta\, w_\beta - 1/2e^{1t}\star w_\beta\, w_\rho - 1/4e^{t}\star w_\beta\, w_\sigma + 1/4e^{-1t}\star w_\beta\, w_\beta + 1/6e^{-1t}\star w_\beta\, w_\rho - 1/4e^{-1t}\star w_\beta\, w_\sigma + 1/96e^{2t}\star w_\rho\, w_\rho + 1/6e^{t}\star w_\rho\, w_\beta - 1/3e^{t}\star w_\rho\, w_\rho - 1/6e^{t}\star w_\rho\, w_\sigma - 1/2e^{-1t}\star w_\rho\, w_\beta - 1/3e^{-1t}\star w_\rho\, w_\rho + 1/2e^{-1t}\star w_\rho\, w_\sigma - 1/12e^{-2t}\star w_\rho\, w_\beta + 1/96e^{-2t}\star w_\rho \cdots 1/6e^{-2t}\star w_\rho\, w_\sigma - 1/6e^{2t}\star w_\sigma\, w_\rho - 1/4e^{t}\star w_\sigma\, w_\beta + 1/2e^{t}\star w_\sigma\, w_\rho + 1/4e^{t}\star w_\sigma\, w_\sigma - 1/4e^{-1t}\star w_\sigma\, w_\beta - 1/6e^{-1t}\star w_\sigma\, w_\rho + 1/4e^{-1t}\star w_\sigma\, w_\sigma\big) + 3/4X_1Y_2^2\sigma w_\rho - 1/2X_1Y_2^2 + Y_1\sigma^2\big(1/4e^{2t}\star w_\rho\, w_\rho - 1/2e^{2t}\star w_\sigma\, w_\rho\big) + Y_1\sigma\big(-1/2w_\rho - w_\sigma\big) - 2Y_1$$

$$\dot Y_2 = X_1^2Y_2\sigma^2\big(e^{t}\star w_\beta\, w_\beta + e^{t}\star w_\beta\, w_\rho - 1/3e^{-1t}\star w_\beta\, w_\rho + e^{-1t}\star w_\beta\, w_\sigma + 2/3e^{t}\star w_\rho\, w_\beta + 2/3e^{t}\star w_\rho\, w_\rho + 2/3e^{-1t}\star w_\rho\, w_\rho - 2e^{-1t}\star w_\rho\, w_\sigma + 1/3e^{-2t}\star w_\rho\, w_\beta - 13/24e^{-2t}\star w_\rho\, w_\rho + 5/6e^{-2t}\star w_\rho\, w_\sigma - e^{t}\star w_\sigma\, w_\beta - e^{t}\star w_\sigma\, w_\rho + 1/3e^{-1t}\star w_\sigma\, w_\rho - e^{-1t}\star w_\sigma\, w_\sigma\big) + X_1^2Y_2\sigma\big(-w_\beta - 1/2w_\rho + w_\sigma\big) + X_1^2Y_2 - Y_2\sigma w_\beta - Y_2$$

$$\dot X_1 = X_1^3\sigma^2\big(-1/4e^{-2t}\star e^{-2t}\star w_\rho\, w_\rho + 1/2e^{-2t}\star e^{-2t}\star w_\rho\, w_\sigma - 1/2e^{-1t}\star w_\beta\, w_\beta - 1/3e^{-1t}\star w_\beta\, w_\rho + 1/2e^{-1t}\star w_\beta\, w_\sigma + 1/4e^{-2t}\star w_\beta\, w_\rho - 1/2e^{-2t}\star w_\beta\, w_\sigma - 1/2e^{-1t}\star w_\rho\, w_\beta \cdots 1/3e^{-1t}\star w_\rho\, w_\rho + 1/2e^{-1t}\star w_\rho\, w_\sigma - 1/12e^{-2t}\star w_\rho\, w_\beta - 13/48e^{-2t}\star w_\rho\, w_\rho + 3/8e^{-2t}\star w_\rho\, w_\sigma - 1/8e^{-2t}\star w_\sigma\, w_\rho + 1/4e^{-2t}\star w_\sigma\, w_\sigma\big) + X_1^3\sigma\big(1/2w_\beta + 3/4w_\rho - 1/4w_\sigma\big) - 1/2X_1^3 + X_1\sigma^2\big(-1/4e^{-2t}\star w_\rho\, w_\rho + 1/2e^{-2t}\star w_\rho\, w_\sigma\big) + 1/2X_1\sigma w_\rho$$

In their analysis Potzsche & Rasmussen (2006) explicitly report the last and third-to-last terms above, for these choices of σ_0 and β_0, to deduce their model (5.3) which here is

$$\dot X \approx \tfrac12\sigma w_\rho X - \tfrac12 X^3.$$

Nice agreement.

2.13.2 Fluctuating kdV example

Potzsche & Rasmussen (2006) [Example 5.4] seek travelling wave solutions, $u(x-ct)$ with wave speed c, of a modified KdV equation. This leads to the following system

$$\dot x_1 = x_2\,, \quad \dot x_2 = x_3\,, \quad \dot x_3 = c^2 x_2 - a(t)x_1^2x_2\,.$$

My analysis is for wave speed $c^2 = 1$. A transform to diagonalise the linear part into slow variable x, stable y and unstable z is then that $x_1 = x+y+z$, $x_2 = z-y$ and $x_3 = z+y$.

```
if thecase=sdePRKdV then begin
afn:=w(a)*(x(1)+y(1)+z(1))^2*(z(1)-y(1));
xrhs := { afn }$
yrhs := { -y(1)-afn/2 }$
zrhs := { +z(1)-afn/2 }$
factor small,sigma,zz,z,yy,y,xx,x;
toosmall:=3;
end;
```

Using `w(a)` to denote the variable coefficient $a(t)$, it is represented in this output by σw_a.

The specified dynamical system

$\dot x_1 = -x_1^2y_1\sigma w_a + x_1^2z_1\sigma w_a - 2x_1y_1^2\sigma w_a + 2x_1z_1^2\sigma w_a - y_1^3\sigma w_a - y_1^2z_1\sigma w_a + y_1z_1^2\sigma w_a + z_1^3\sigma w_a$

$\dot y_1 = 1/2x_1^2y_1\sigma w_a - 1/2x_1^2z_1\sigma w_a + x_1y_1^2\sigma w_a - x_1z_1^2\sigma w_a + 1/2y_1^3\sigma w_a + 1/2y_1^2z_1\sigma w_a - 1/2y_1z_1^2\sigma w_a - y_1 - 1/2z_1^3\sigma w_a$

$\dot z_1 = 1/2x_1^2y_1\sigma w_a - 1/2x_1^2z_1\sigma w_a + x_1y_1^2\sigma w_a - x_1z_1^2\sigma w_a + 1/2y_1^3\sigma w_a + 1/2y_1^2z_1\sigma w_a - 1/2y_1z_1^2\sigma w_a - 1/2z_1^3\sigma w_a + z_1$

Time dependent coordinate transform

$z_1 = -1/2X_1^2Y_1\sigma e^{2t}\star w_a - X_1Y_1^2\sigma e^{3t}\star w_a - X_1Z_1^2\sigma e^{-1t}\star w_a - 1/2Y_1^3\sigma e^{4t}\star w_a - 1/2Y_1^2Z_1\sigma e^{2t}\star w_a - 1/2Z_1^3\sigma e^{-2t}\star w_a + Z_1$

$y_1 = -1/2X_1^2Z_1\sigma e^{-2t}\star w_a - X_1Y_1^2\sigma e^{t}\star w_a - X_1Z_1^2\sigma e^{-3t}\star w_a - 1/2Y_1^3\sigma e^{2t}\star w_a - 1/2Y_1Z_1^2\sigma e^{-2t}\star w_a + Y_1 - 1/2Z_1^3\sigma e^{-4t}\star w_a$

$x_1 = X_1^2 Y_1 \sigma e^{t}\star w_a + X_1^2 Z_1 \sigma e^{-1t}\star w_a + 2X_1 Y_1^2 \sigma e^{2t}\star w_a + 2X_1 Z_1^2 \sigma e^{-2t}\star w_a + X_1 + Y_1^3 \sigma e^{3t}\star w_a + Y_1^2 Z_1 \sigma e^{t}\star w_a + Y_1 Z_1^2 \sigma e^{-1t}\star w_a + Z_1^3 \sigma e^{-3t}\star w_a$

Putting $Z_1 = 0$ into the coordinate transform gives the centre-stable manifold. Then the expression for z_1 in the above coordinate transform leads to the same convolutions as those of Potzsche & Rasmussen (2006) [pp.453–4]. Conversely, putting $Y_1 = 0$ gives the centre-unstable manifold and the expression for y_1 above leads to the same convolutions as those of Potzsche & Rasmussen (2006). Presumably the distortions of the other variables have a higher order influence on this nice agreement.

2.14 Local analysis of heat exchanger

Roberts (2013) provides novel theoretical support for the method of multiple scales in spatio-temporal systems, and then extends this important method. Perhaps the simplest example is the heat exchanger: the non-autonomous slow manifold analysis that is at the heart of the novel methodology is determined here. Expand advection-exchange in a heat exchanger in powers of $(x-X)^n/n!$. With Taylor Remainder Theorem closing the problem in terms of unknown functions which here are represented by the non-autonomous forcing w_i. Note that `y(j)` $= d_{j-1}$ and `x(j)` $= c_{j-1}$. Also `w(1)` $= d_{4X}\eta_x$ and `w(2)` $= c_{4X}\xi_x$ and evaluate at intensity $\sigma = 5$.

```
if thecase=sdehe then begin
xrhs:={y(2),y(3),y(4),y(5),w(1)};
yrhs:={-y(1)+x(2),-y(2)+x(3),-y(3)+x(4),-y(4)+x(5),-y(5)+w(2)};
zrhs:={ };
toosmall:=6;
factor small,sigma;
end;
```

Specified dynamical system The above writes the ODEs as the following. $\dot x_1 = \varepsilon y_2$, $\dot x_2 = \varepsilon y_3$, $\dot x_3 = \varepsilon y_4$, $\dot x_4 = \varepsilon y_5$, $\dot x_5 = \sigma w_1$, $\dot y_1 = \varepsilon x_2 - y_1$, $\dot y_2 = \varepsilon x_3 - y_2$, $\dot y_3 = \varepsilon x_4 - y_3$, $\dot y_4 = \varepsilon x_5 - y_4$, $\dot y_5 = \sigma w_2 - y_5$.

Time dependent coordinate transform $y_1 = \sigma\varepsilon^4\big(e^{-1t}\star e^{-1t}\star e^{-1t}\star w_2 + 2e^{-1t}\star e^{-1t}\star w_2 + 3e^{-1t}\star w_2\big) - \varepsilon^3 X_4 + \varepsilon X_2 + Y_1$, $y_2 = \sigma\varepsilon^3\big(e^{-1t}\star e^{-1t}\star w_1 + 2e^{-1t}\star w_1\big) - \varepsilon^3 X_5 + \varepsilon X_3 + Y_2$, $y_3 = \sigma\varepsilon^2\big(-e^{-1t}\star e^{-1t}\star w_2 - e^{-1t}\star w_2\big) + \varepsilon X_4 + Y_3$, $y_4 = -\sigma\varepsilon e^{-1t}\star w_1 + \varepsilon X_5 + Y_4$, $y_5 = \sigma e^{-1t}\star w_2 + Y_5$.

And the slow variables $x_1 = \sigma\varepsilon^4\big(-e^{-1t}\star e^{-1t}\star w_1 - 3e^{-1t}\star w_1\big) + \varepsilon^3 Y_4 - \varepsilon Y_2 + X_1$, $x_2 = \sigma\varepsilon^3\big(e^{-1t}\star e^{-1t}\star w_2 + 2e^{-1t}\star w_2\big) + \varepsilon^3 Y_5 - \varepsilon Y_3 + X_2$, $x_3 = \sigma\varepsilon^2 e^{-1t}\star w_1 - \varepsilon Y_4 + X_3$, $x_4 = -\sigma\varepsilon e^{-1t}\star w_2 - \varepsilon Y_5 + X_4$, $x_5 = X_5$.

Result normal form DEs $\dot Y_1 = \varepsilon^4 Y_5 - \varepsilon^2 Y_3 - Y_1$, $\dot Y_2 = -\varepsilon^2 Y_4 - Y_2$, $\dot Y_3 = -\varepsilon^2 Y_5 - Y_3$, $\dot Y_4 = -Y_4$, $\dot Y_5 = -Y_5$.

$\dot X_1 = 3\sigma\varepsilon^4 w_1 - \varepsilon^4 X_5 + \varepsilon^2 X_3$, $\dot X_2 = -2\sigma\varepsilon^3 w_2 + \varepsilon^2 X_4$, $\dot X_3 = -\sigma\varepsilon^2 w_1 + \varepsilon^2 X_5$, $\dot X_4 = \sigma\varepsilon w_2$, $\dot X_5 = \sigma w_1$.

2.14.1 Near the boundary

This is for the case of boundary conditions $c + pd = cd_0(t)$ at $x = 0$ for some parameter p. Computer algebra finds boundary conditions on the fields that reduce the dynamics near the boundary to the following with `x(1)` $= c_1$, `x(2)` $= c_3$, `y(1)` $= d_0$, `y(2)` $= d_2$ and `w(1)` $= d_{3X}\eta_x$ with $\sigma = 4$. Curiously, there is no dependence upon parameter p in these dynamics.

```
if thecase=sdehebc then begin
xrhs:={y(2),w(1)};
yrhs:={-y(1)+x(1),-y(2)+x(2)};
zrhs:={ };
toosmall:=6;
factor small,sigma;
end;
```

Again, I believe the following results are exact.

Specified dynamical system $\dot x_1 = \varepsilon y_2$, $\dot x_2 = \sigma w_1$, $\dot y_1 = \varepsilon x_1 - y_1$, $\dot y_2 = \varepsilon x_2 - y_2$.

Time dependent coordinate transform $y_1 = \sigma\varepsilon^3\big(e^{-1t}\star e^{-1t}\star w_1 + 2e^{-1t}\star w_1\big) - \varepsilon^3 X_2 + \varepsilon X_1 + Y_1$, $y_2 = -\sigma\varepsilon e^{-1t}\star w_1 + \varepsilon X_2 + Y_2$, $x_1 = \sigma\varepsilon^2 e^{-1t}\star w_1 - \varepsilon Y_2 + X_1$, $x_2 = X_2$.

Result normal form DEs $\dot Y_1 = -\varepsilon^2 Y_2 - Y_1$, $\dot Y_2 = -Y_2$, $\dot X_1 = -\sigma\varepsilon^2 w_1 + \varepsilon^2 X_2$, $\dot X_2 = \sigma w_1$.

2.14.2 Heat exchanger with quadratic reaction

Expand advection-reaction-exchange in a heat exchanger in powers of $(x - X)^n/n!$. The reaction is some quadratic that should generate Burgers' equation model. With Taylor Remainder Theorem closing the problem in terms of unknown functions which here are represented by the non-autonomous forcing w_i. Note that `y(j)` $= d_{j-1}$ and `x(j)` $= c_{j-1}$. Also `w(1)` $= 3d_{2x}$ and `w(2)` $= 3c_{2x}$ and evaluate at intensity $\sigma = 1$.

```
if thecase=sdeheqr then begin
xrhs:={y(2)-x(1)*y(1)
      ,y(3)-x(1)*y(2)-x(2)*y(1)
      ,small*w(1)-x(1)*y(3)-2*x(2)*y(2)-x(3)*y(1)
      };
yrhs:={-y(1)+x(2)-(x(1)^2+y(1)^2)/2
      ,-y(2)+x(3)-x(1)*x(2)-y(1)*y(2)
      ,-y(3)+small*w(2)-x(2)^2-x(1)*x(3)-y(2)^2-y(1)*y(3)
      };
zrhs:={ };
toosmall:=4;
factor small,sigma;
end;
```

Alternatively, we could divide the off-diagonal linear terms by `small` (and remove the multiplication of forcing `w`), and the algorithm still converges, albeit in more iterations. The resulting asymptotic expressions then do not assume that x derivatives are successively smaller.

The following uses the default scaling which corresponds to successively smaller x-derivatives provided I also multiply the forcing by `small`.

Specified dynamical system $\dot{x}_1 = \varepsilon(-x_1y_1+y_2)$, $\dot{x}_2 = \varepsilon(-x_2y_1-x_1y_2+y_3)$, $\dot{x}_3 = \sigma\varepsilon w_1 + \varepsilon(-x_3y_1-2x_2y_2-x_1y_3)$, $\dot{y}_1 = \varepsilon(x_2-1/2x_1^2-1/2y_1^2)-y_1$, $\dot{y}_2 = \varepsilon(x_3-x_2x_1-y_2y_1)-y_2$, $\dot{y}_3 = \sigma\varepsilon w_2+\varepsilon(-x_3x_1-x_2^2-y_3y_1-y_2^2)-y_3$

Time dependent coordinate transform

$$y_1 = 1/4\varepsilon^2Y_1^3 + \varepsilon(X_2 - 1/2X_1^2 + 1/2Y_1^2) + Y_1$$

$$y_2 = -\sigma\varepsilon^2e^{-1t}\star w_1 + 3/4\varepsilon^2Y_2Y_1^2 + \varepsilon(X_3 - X_2X_1 + Y_2Y_1) + Y_2$$

$$y_3 = \sigma\varepsilon^2(e^{-1t}\star w_2\, Y_1 + e^{-1t}\star w_1\, X_1) + \sigma\varepsilon e^{-1t}\star w_2 + \varepsilon^2(3/4Y_3Y_1^2 + 3/2Y_2^2Y_1) + \varepsilon(-X_3X_1 - X_2^2 + Y_3Y_1 + Y_2^2) + Y_3$$

$$x_1 = \varepsilon^2(3/4X_1Y_1^2 - Y_2Y_1) + \varepsilon(X_1Y_1 - Y_2) + X_1$$

$$x_2 = -\sigma\varepsilon^2e^{-1t}\star w_2 + \varepsilon^2(3/4X_2Y_1^2 + 3/2X_1Y_2Y_1 - Y_3Y_1 - Y_2^2) + \varepsilon(X_2Y_1 + X_1Y_2 - Y_3) + X_2$$

$$x_3 = \sigma\varepsilon^2(e^{-1t}\star w_2\, X_1 + e^{t}\star w_1\, Y_1) + \varepsilon^2(3/4X_3Y_1^2 + 3X_2Y_2Y_1 + 3/2X_1Y_3Y_1 + 3/2X_1Y_2^2 - 3/2Y_3Y_2) + \varepsilon(X_3Y_1 + 2X_2Y_2 + X_1Y_3) + X_3$$

Result normal form DEs

$$\dot{Y}_1 = \varepsilon^2(-1/2X_1^2Y_1 + 2X_1Y_2 - Y_3) - Y_1$$

$$\dot{Y}_2 = 2\sigma\varepsilon^3w_1Y_1 + \varepsilon^2(-X_2X_1Y_1 + 2X_2Y_2 - 1/2X_1^2Y_2 + 2X_1Y_3) - Y_2$$

$$\dot{Y}_3 = \sigma\varepsilon^3(-2w_1X_1Y_1 + 2w_1Y_2) - \sigma\varepsilon^2w_2Y_1 + \varepsilon^2(-X_3X_1Y_1 - X_3Y_2 - X_2^2Y_1 - 2X_2X_1Y_2 + X_2Y_3 - 1/2X_1^2Y_3) - Y_3$$

$$\dot{X}_1 = -\sigma\varepsilon^3w_1 + \varepsilon^2(X_3 - 2X_2X_1 + 1/2X_1^3)$$

$$\dot{X}_2 = 2\sigma\varepsilon^3w_1X_1 + \sigma\varepsilon^2w_2 + \varepsilon^2(-2X_3X_1 - 2X_2^2 + 3/2X_2X_1^2)$$

$$\dot{X}_3 = \sigma\varepsilon^3(2w_1X_2 - w_1X_1^2) - \sigma\varepsilon^2w_2X_1 + \sigma\varepsilon w_1 + \varepsilon^2(-3X_3X_2 + 3/2X_3X_1^2 + 3X_2^2X_1)$$

Hmmm, looks like this generates the slowly varying model that

$$\frac{\partial C}{\partial t} \approx \frac{\partial^2 C}{\partial x^2} - 2C\frac{\partial C}{\partial x} + \tfrac{1}{2}C^3.$$

Interestingly there is an extra factor of two in the nonlinear advection, and a net cubic reaction.

2.15 Michaelis–Menten–Henri deterministic model

$$\begin{aligned}\dot x &= \epsilon[-x + (x+\kappa-\lambda)y],\\ \dot y &= x - (x+\kappa)y.\end{aligned}$$

A manifold of equilibria occur at $y = x/(x+\kappa)$ and $\epsilon = 0$ (also if $\epsilon \neq 0$ and $\lambda = 0$ but we do not consider this case). Let's explore dynamics based at arbitrary point on this equilibrium manifold: substitute $x(t) = x_0 + x_1(t)$ and $y(t) = x_0/(x_0+\kappa) + y_1(t)$, and derive

$$\begin{aligned}\frac{1}{x_0+\kappa}\dot x_1 &= x_1' = \frac{\epsilon}{x_0+\kappa}\left[-x_0 - x_1 + (x_0+x_1+\kappa-\lambda)(x_0/(x_0+\kappa)+y_1)\right]\\ \frac{1}{x_0+\kappa}\dot y_1 &= y_1' = -y_1 + \frac{1}{x_0+\kappa}\left[-x_1y_1 + x_0 + x_1 - \frac{(x_0+x_1+\kappa)x_0}{x_0+\kappa}\right]\end{aligned}$$

We need the above form in order to get the decay rate of y_1 to be a simple number—but it does mean that we have to be careful interpreting the results. The reason is that the time derivative in the analysis is stretched by the factor $x_0+\kappa$: that is, $(x_0+\kappa)dt = d\tau$ where τ is the time of the analysis. Have only tentatively put in some stochastic effects as 'additive' noise into this system.

```
if thecase=sdemmh then begin
% define rho=1/(x0+kappa)
let rho*x0=>1-rho*kappa;
lam:=lambda;
kappa:=1; lam:=1/2; %temporary for simplicity
xrhs:={ eps*rho*(-x0-x(1)+(x0+x(1)+kappa-lam)*(x0*rho+y(1)))
```

```
    +w(1) };
yrhs:={ -y(1)+rho*(-x(1)*y(1)+x0+x(1)-(x0+x(1)+kappa)*x0*rho)
    +w(2) };
zrhs:={ };
toosmall:=4;
factor eps,sigma;
end;
```

In deterministic results, $Y_1 \to 0$ to form the slow manifold. Then the x-evolution is modelled by simply putting $X_1 = 0$ in the right-hand side of $\dot{X}_1$. This gets rid of a lot of terms.

As a prototypical example, let's investigate the simplest stochastic effects on this MM system of an additive noise. The additive noise will transform to a multiplicative noise on the slow manifold, so it is important to remember that *all* analysis and results are in the *Stratonovich* interpretation.

The analysis here is *strong, pathwise.*

The transformations here only rely on the 'noise' being measurable, so the results also apply to deterministic non-autonomous forcing. The analysis may also apply to to non-Brownian noise provided the appropriate interpretation is used (e.g., the Marcus interpretation). That is, as long as standard rules of integral calculus are valid.

Specified dynamical system Here add 'independent noises', $w_1(t)$ and $w_2(t)$, Stratonovich, to the ODEs of the transformed MM system, noises of strength σ. It is as yet unclear what this addition means in terms of the original MM system.

$$x_1' = \epsilon\varepsilon\big(x_1 y_1 \rho - x_1 \rho^2 - 1/2 y_1 \rho + y_1 + 1/2\rho^2 - 1/2\rho\big) + \sigma w_1$$

$$y_1' = \sigma w_2 + \varepsilon\big(-x_1 y_1 \rho + x_1 \rho^2\big) - y_1$$

Time dependent coordinate transform The algorithm constructs a coordinate transform to variables (X_1, Y_1), including terms quadratic in σ, that to errors $\mathcal{O}(\sigma^2,\varepsilon^3)$, is the following. The coordinate transform depends upon both the past and the future via convolutions $e^{-1t}\star$ and $e^{t}\star$, respectively.

The following expressions are complicated because stochastic effects interact through nonlinearity in a combinatorial explosion of ways. We almost certainly do not need all these terms. I subsequently explain why the blue terms are the ones describing the emergent stochastic slow manifold and the evolution thereon. Further, remember that the dominant terms are towards the end of each expression.

$$y_1 = \epsilon\sigma\varepsilon^2\big(-e^{-1t}\star e^{-1t}\star e^{-1t}\star w_2\ X_1\rho^3 + 1/2e^{-1t}\star e^{-1t}\star e^{-1t}\star w_2\ \rho^3 - 1/2e^{-1t}\star e^{-1t}\star e^{-1t}\star w_2\ \rho^2 - e^{-1t}\star e^{-1t}\star w_2\, X_1\rho^3 + 1/2e^{-1t}\star e^{-1t}\star w_2\, \rho^3 - e^{-1t}\star e^{-1t}\star w_2\, \rho^2 - e^{t}\star e^{t}\star w_1\ Y_1^2\rho^2 - 2e^{-1t}\star w_2\, X_1Y_1\rho^2 - e^{-1t}\star w_2\, X_1\rho^3 + e^{-1t}\star w_2\, Y_1\rho^2 - 2e^{-1t}\star w_2\, Y_1\rho + 1/2e^{-1t}\star w_2\, \rho^3 - e^{-1t}\star w_2\, \rho^2 - e^{1t}\star w_1\ Y_1^2\rho^2 - e^{t}\star w_1\, Y_1\rho^3 - e^{-1t}\star w_1\, \rho^4\big) + \epsilon\varepsilon^2\big(-X_1Y_1^2\rho^2 + X_1\rho^4 + 1/2Y_1^2\rho^2 - Y_1^2\rho - 1/2\rho^4 + 1/2\rho^3\big) + \sigma\varepsilon^2\big(e^{-1t}\star e^{-1t}\star e^{-1t}\star w_2\ X_1^2\rho^2 + e^{-1t}\star e^{-1t}\star w_1\ X_1\rho^3 + 2e^{-1t}\star w_1\, X_1\rho^3\big) + \sigma\varepsilon\big(-e^{-1t}\star e^{-1t}\star w_2\ X_1\rho - e^{-1t}\star w_1\, \rho^2\big) + \sigma e^{-1t}\star w_2 - \varepsilon^2X_1^2\rho^3 + \varepsilon X_1\rho^2 + Y_1$$

$$x_1 = \epsilon^2\sigma\varepsilon^2\big(-e^{t}\star e^{t}\star w_1\ Y_1\rho^3 + e^{-1t}\star w_2\, X_1Y_1\rho^2 - 1/2e^{-1t}\star w_2\, Y_1\rho^2 + e^{-1t}\star w_2\, Y_1\rho - 1/2e^{-1t}\star w_2\, \rho^2 - 1/2e^{2t}\star w_1\ Y_1^2\rho^2 + e^{t}\star w_1\ Y_1^2\rho^2\big) + \epsilon^2\varepsilon^2\big(1/2X_1Y_1^2\rho^2 - 1/4Y_1^2\rho^2 + 1/2Y_1^2\rho - 1/2Y_1\rho^2\big) + \epsilon\sigma\varepsilon^2\big(e^{-1t}\star e^{-1t}\star w_2\ X_1^2\rho^2 - 1/2e^{-1t}\star e^{-1t}\star w_2\ X_1\rho^2 + e^{-1t}\star e^{-1t}\star w_2\ X_1\rho + e^{t}\star e^{t}\star w_1\ X_1Y_1\rho^2 + e^{-1t}\star w_2\, X_1^2\rho^2 - 1/2e^{-1t}\star w_2\, X_1\rho^2 + e^{-1t}\star w_2\, X_1\rho + 2e^{t}\star w_1\, X_1Y_1\rho^2 - 1/2e^{t}\star w_1\, Y_1\rho^2 + e^{t}\star w_1\, Y_1\rho + e^{-1t}\star w_1\, X_1\rho^3 - 1/2e^{-1t}\star w_1\, \rho^3 + e^{-1t}\star w_1\, \rho^2\big) + \epsilon\sigma\varepsilon\big(-e^{-1t}\star w_2\, X_1\rho + 1/2e^{-1t}\star w_2\, \rho - e^{-1t}\star w_2 - e^{t}\star w_1\, Y_1\rho\big) + \epsilon\varepsilon^2\big(X_1^2Y_1\rho^2 - 1/2X_1Y_1\rho^2 + X_1Y_1\rho\big) + \epsilon\varepsilon\big(-X_1Y_1\rho + 1/2Y_1\rho - Y_1\big) + X_1$$

These new coordinates (X_1, Y_1) are non-Markovian in relation to (x_1, y_1), in some sense, but the non-Markovian nature is exponentially decaying away from the current time. The construction of a non-autonomous stochastic slow manifold has to look to the future and the past in order to find out what variations are going to stay bounded for all time.

Result normal form DEs In the (X_1, Y_1) coordinates, the stochastic system satisfies the following Stratonovich system, to errors $\mathcal{O}(\sigma^3, \varepsilon^4)$.

$$Y_1' = \epsilon^2\sigma^2\varepsilon^3\big(e^{-1t}\star e^{-1t}\star e^{-1t}\star w_2\ \ w_1Y_1\rho^4 + e^{-1t}\star e^{-1t}\star w_2\ w_1Y_1\rho^4 + 1/2e^{-1t}\star w_2\,w_2X_1Y_1\rho^3 - 1/4e^{-1t}\star w_2\,w_2Y_1\rho^3 + 1/2e^{-1t}\star w_2\,w_2Y_1\rho^2 - e^{-1t}\star w_2\,w_1Y_1\rho^4\big) + \epsilon^2\sigma\varepsilon^3\big(-w_2X_1Y_1\rho^4 + 1/2w_2Y_1\rho^4 + w_2Y_1\rho^3 - w_1Y_1\rho^5\big) - 1/2\epsilon^2\varepsilon^3Y_1\rho^4 + \epsilon\sigma^2\varepsilon^3\big(-2e^{-1t}\star e^{-1t}\star e^{-1t}\star w_2\ \ w_1X_1Y_1\rho^3 - 4e^{-1t}\star e^{-1t}\star w_2\ w_1X_1Y_1\rho^3 + 1/2e^{-1t}\star e^{-1t}\star w_2\ w_1Y_1\rho^3 - e^{-1t}\star e^{-1t}\star w_2\ w_1Y_1\rho^2 - e^{-1t}\star e^{-1t}\star w_1\ w_1Y_1\rho^4 - 8e^{-1t}\star w_2\,w_1X_1Y_1\rho^3 + 2e^{-1t}\star w_2\,w_1Y_1\rho^3 - 4e^{-1t}\star w_2\,w_1Y_1\rho^2 - 2e^{t}\star w_1\,w_1Y_1\rho^4 - 2e^{-1t}\star w_1\,w_1Y_1\rho^4\big) + \epsilon\sigma^2\varepsilon^2\big(e^{-1t}\star e^{-1t}\star w_2\ w_1Y_1\rho^2 + 2e^{-1t}\star w_2\,w_1Y_1\rho^2\big) + \epsilon\sigma\varepsilon^3\big(-4w_2X_1^2Y_1\rho^3 + 2w_2X_1Y_1\rho^3 - 4w_2X_1Y_1\rho^2 + 2w_1X_1Y_1\rho^4 + 1/2w_1Y_1\rho^4 - w_1Y_1\rho^3\big) + \epsilon\sigma\varepsilon^2\big(2w_2X_1Y_1\rho^2 - w_2Y_1\rho^2 + 2w_2Y_1\rho - w_1Y_1\rho^3\big) + \epsilon\varepsilon^3\big(2X_1^2Y_1\rho^4 - X_1Y_1\rho^4 + 2X_1Y_1\rho^3\big) + \epsilon\varepsilon^2\big(-X_1Y_1\rho^3 + 1/2Y_1\rho^3 - Y_1\rho^2\big) - \varepsilon X_1Y_1\rho - Y_1$$

$$X_1' = -1/4\epsilon^3\sigma^2\varepsilon^3e^{-1t}\star w_2\,w_2\rho^3 + 1/2\epsilon^3\sigma\varepsilon^3w_2\rho^4 + \epsilon^2\sigma^2\varepsilon^3\big(-e^{-1t}\star e^{-1t}\star e^{-1t}\star w_2\ \ w_1X_1\rho^4 + 1/2e^{-1t}\star e^{-1t}\star e^{-1t}\star w_2\ \ w_1\rho^4 - 1/2e^{-1t}\star e^{-1t}\star e^{-1t}\star w_2\ \ w_1\rho^3 - e^{-1t}\star e^{-1t}\star w_2\ w_1X_1\rho^4 + 1/2e^{-1t}\star e^{-1t}\star w_2\ w_1\rho^4 - e^{-1t}\star e^{-1t}\star w_2\ w_1\rho^3 - 2e^{-1t}\star w_2\,w_1X_1\rho^4 + e^{-1t}\star w_2\,w_1\rho^4 - 7/2e^{-1t}\star w_2\,w_1\rho^3 - e^{-1t}\star w_1\,w_1\rho^5\big) + \epsilon^2\sigma\varepsilon^3\big(-w_2X_1^2\rho^4 + w_2X_1\rho^4 - 7/2w_2X_1\rho^3 - 1/4w_2\rho^4 + 5/4w_2\rho^3 - 3/2w_2\rho^2 - w_1X_1\rho^5 + 1/2w_1\rho^5 - 3/2w_1\rho^4\big) + 1/2\epsilon^2\sigma\varepsilon^2w_2\rho^2 + \epsilon^2\varepsilon^3\big(X_1^2\rho^5 - X_1\rho^5 + 3/2X_1\rho^4 + 1/4\rho^5 - 3/4\rho^4 + 1/2\rho^3\big) + \epsilon\sigma^2\varepsilon^3\big(e^{-1t}\star e^{-1t}\star e^{-1t}\star w_2\ \ w_1X_1^2\rho^3 + 2e^{-1t}\star e^{-1t}\star w_2\ w_1X_1^2\rho^3 - 1/2e^{-1t}\star e^{-1t}\star w_2\ w_1X_1\rho^3 + e^{-1t}\star e^{-1t}\star w_2\ w_1X_1\rho^2 + e^{-1t}\star e^{-1t}\star w_1\ w_1X_1\rho^4 + 3e^{-1t}\star w_2\,w_1X_1^2\rho^3 - e^{-1t}\star w_2\,w_1X_1\rho^3 + 2e^{-1t}\star w_2\,w_1X_1\rho^2 + 4e^{-1t}\star w_1\,w_1X_1\rho^4 - 1/2e^{-1t}\star w_1\,w_1\rho^4 + e^{-1t}\star w_1\,w_1\rho^3\big) + \epsilon\sigma^2\varepsilon^2\big(-e^{-1t}\star e^{-1t}\star w_2\ w_1X_1\rho^2 - 2e^{-1t}\star w_2\,w_1X_1\rho^2 + 1/2e^{-1t}\star w_2\,w_1\rho^2 - e^{-1t}\star w_2\,w_1\rho - e^{-1t}\star w_1\,w_1\rho^3\big) + \epsilon\sigma^2\varepsilon e^{-1t}\star w_2\,w_1\rho + \epsilon\sigma\varepsilon^3\big(w_2X_1^3\rho^3 - 1/2w_2X_1^2\rho^3 + w_2X_1^2\rho^2 + 3w_1X_1^2\rho^4 - 3/2w_1X_1\rho^4 + 3w_1X_1\rho^3\big) + \epsilon\sigma\varepsilon^2\big(-w_2X_1^2\rho^2 + 1/2w_2X_1\rho^2 - w_2X_1\rho - w_1X_1\rho^3 + 1/2w_1\rho^3 - w_1\rho^2\big) + \epsilon\sigma\varepsilon\big(w_2X_1\rho - 1/2w_2\rho + w_2\big) + \epsilon\varepsilon^3\big(-X_1^3\rho^4 + 1/2X_1^2\rho^4 - X_1^2\rho^3\big) + \epsilon\varepsilon^2\big(X_1^2\rho^3 - 1/2X_1\rho^3 + X_1\rho^2\big) + \epsilon\varepsilon\big(-X_1\rho^2 + 1/2\rho^2 - 1/2\rho\big) + \sigma w_1$$

Discussion

- In the Y_1' SDE, by construction, every term is $\propto Y_1$, and, further, the leading term gives $Y_1' \approx -Y_1$. Hence, $Y_1 \approx \mathcal{O}(e^{-\tau}) = \mathcal{O}(e^{-\int(x_0+\kappa)dt})$ as time increases. Consequently, by continuity, *in some finite domain* about (x_0, y_0), $Y_1 \to 0$ to form the emergent stochastic slow manifold

$Y_1 = 0$.

- The local shape of the slow manifold is thus given by substituting $Y_1 = 0$ into the expressions for (x_1, y_1). Thus the slow manifold is locally parametrised by X_1, ϵ, σ. Now the variation in X_1 is the Taylor series for the variation in x_0 (as they are both describing the same slow manifold). So all we need is to set $X_1 = 0$ and look at the shape of the slow manifold in terms of x_0, ϵ, σ, that is, the blue terms.
  - Setting $X_1 = Y_1 = 0$ gives $y_1 \approx \epsilon\frac{1}{2}(\rho^3 - \rho^4) - \sigma e^{-1t}\star w_1\,\rho^2 + \sigma e^{-1t}\star w_2$, that is, since $\rho = 1/(x+\kappa) = 1/(x+1)$,
  $$y = y_0 + y_1 \approx \frac{x_0}{x_0+1} + \epsilon\frac{x_0}{2(x_0+1)^4} - \frac{\sigma e^{-1t}\star w_1}{(x+1)^2} + \sigma e^{-1t}\star w_2\,.$$
  Dominantly, the slow manifold jitters up/down in y due to the recent history of noise w_2, but also is affected from the recent history of the noise w_1 in x.
  - Setting $X_1 = Y_1 = 0$ gives $x_1 \approx \epsilon\sigma\varepsilon\big(1/2e^{-1t}\star w_2\,\rho - e^{-1t}\star w_2\big)$, that is,
  $$x = x_0 + x_1 \approx x_0 - \epsilon\sigma\frac{2x+1}{2(x+1)}e^{-1t}\star w_2\,.$$
  The noise w_2 in y generates a history dependent slip between x and the relevant x_0!

    This slip may be seen to be due to the slope of isochrons transversal to the slow manifold—a slope not detected in SPT.

  This stochastic-MM example also shows the general property that although the *existence* of a slow manifold has future dependence, here via $e^{t}\star$ convolutions, the slow manifold itself and the evolution thereon depends only upon the history, here via $e^{-1t}\star$ convolutions.

- Now for the x-evolution on the stochastic slow manifold. Consider $X(t) = x_0 + x_1(t)$, so that $X' = x_1'$. Recall that on the slow manifold, $Y_1 = 0$ and $x \approx x_0 - \epsilon\sigma\frac{2x+1}{2(x+1)}e^{-1t}\star w_2$, $X_1 = 0$, so also putting $X_1 = 0$ and $x_0 = X$ give the evolution for the slow variable X, namely the *global* slow evolution is

$$X' \approx -\epsilon\frac{X}{2(X+1)^2} + \epsilon^2\frac{X(2X+1)}{4(X+1)^5} + \sigma w_1 + \epsilon\sigma\frac{2x+1}{2(x+1)}w_2 \,.$$

The stochastic slow variable X is not quite the same as the physical x. This coordinate transform lacks any convolutions in time. That lack is part of the art of the construction.

If, instead, one wants the slow variable to be precisely x, as many implicitly assume they can, then convolutions must occur in x'). We may see this by constructing a nonlinear coordinate transform that maintains, when $Y_1 = 0$, that $x = x_0$, precisely. It is straightforward to modify the algorithm to do so. The generic consequence is that terms *linear* in the noise appear in the evolution x' that have fast-time history convolutions. That is, the consequence is that undesirable fast-time history integrals occur in the evolution of the supposedly slow variable x.

Noise-noise interactions However, effects which are quadratic in the noise, due to noise-noise interactions, generally involve convolutions that *cannot* be removed from the evolution of the slow variable, as seen in expressions for X_1'. Here, the lowest order example is the term

$$+\epsilon\sigma^2 e^{-1t}\star w_2 w_1/(x+1)$$

which could be included in the retained terms of X_1'. We argued (Chao & Roberts 1996, §4) that such terms 'bring up' *new information* from the fluctuations on the fast-time microscale, and hence cause noise effects in the slow model that are independent of slow-scale sampling of w_1 and w_2. We argued that such terms, when one only samples them on the long-times of the slow manifold, should thus be replaced by a new noise, namely $e^{-\beta t}\star w_2 w_1 \sim \frac{1}{2\sqrt{\beta}} w_3$ when all w_j are formally 'the derivatives' of independent Wiener processes.

The above results are for one example of a stochastic MM system. Almost all other stochastic MM systems would have the same issues.

3 General SDE preliminaries

Deterministic, autonomous, normal forms are constructed simply by omitting any noise term `w()` in the differential equations.

The right-hand sides must be multinomial in variables x_i, y_i, z_i and w_i, but off-hand I do not know an easy way to check for this.

Improve appearance Improve appearance of printed output.

```
on div; off allfac; on revpri;
linelength 70$
```

If for the web, then send text output to file

```
if thecase=webpage then out "sdeo.txt"$
```

3.1 Extract and scale slow equations

The number of slow equations is the number of terms in the list in `xrhs`.

```
write "no. of slow modes ",m:=length(xrhs);
```

Multiply all the right-hand sides by `small` so we can control the truncation of the asymptotic construction through discarding high powers of `small`. Users could use `small` in their equations for appropriate effects.

```
xrhs:=for i:=1:m collect small*part(xrhs,i)$
```

Adjust the noise terms. Remove the `small` multiplication of noise terms, and instead multiply by `sigma` to empower me to independently control the truncation in noise amplitude.

```
xrhs:=(xrhs where w(~j)=>sigma*w(j,1)/small)$
xrhs:=(xrhs where w(~j,1)=>w(j))$
```

Section 5 writes the resulting differential equations for information.

3.2 Extract and scale stable fast equations

The number of stable fast equations is the number of terms in the list in `yrhs`.

```
write "no. of stable fast modes ",ny:=length(yrhs);
```

Extract decay rates Extract the linear decay rates of the fast equations into an array. For each expression in the provided set of right-hand sides:

```
array rate(ny);
for i:=1:ny do begin
```

For the ith right-hand side get the linear dependence upon `y(i)`, then set other dynamic variables to zero to get just the coefficient.

```
  rate(i):=coeffn(part(yrhs,i),y(i),1);
  rate(i):=(rate(i) where {x(~j)=>0,y(~j)=>0,z(~j)=>0,w(~j)=>0})
```

However, the coefficient may depend upon parameters, so if it is not simply a number, but is a sum, then trawl through the sum looking for a simple number to use as the decay rate.

```
  if not numberp(rate(i)) then
  if part(rate(i),0)=plus then begin
    rr:=0;
    for j:=1:arglength(rate(i)) do
      if numberp(part(rate(i),j))
      then rr:=part(rate(i),j);
    rate(i):=rr;
  end;
```

Change sign to make `rate` into positive decay rates, rather than negative growth rates.

```
  rate(i):=-rate(i);
```

If all the above has not ended up with a simple number, then exit with an error message.

```
  if numberp(rate(i))and rate(i)>0 then
  else begin
    write "***** Error *****
    Linear coeffs of y-decay must be negative numbers";
    if thecase=wbepage then <<
      shut "sdeo.txt"; quit >>;
  end;
```

End the loop over all right-hand sides.

```
end;
```

Flag later warning if the linear part not diagonal.

```
offdiag:=0$
for i:=1:ny do for j:=1:ny do if i neq j then begin
  jac:=coeffn(part(yrhs,i),y(j),1);
  if (jac where {x(~k)=>0,y(~k)=>0,z(~k)=>0,w(~k)=>0}) neq 0
  then offdiag:=1$
end;
```

Multiply all the 'nonlinear' terms right-hand sides by `small` so we control the truncation of the asymptotic construction through discarding high powers of `small`. Leave the identified linear decay terms intact. Users could use `small` in their equations for interesting effects.

```
yrhs:=for i:=1:ny collect
  small*part(yrhs,i)+(1-small)*(-rate(i)*y(i))$
```

Remove the `small` multiplication of noise terms, and instead multiply by `sigma` to empower me to independently control the truncation in noise amplitude.

```
yrhs:=( yrhs where w(~j)=>sigma*w(j,1)/small )$
yrhs:=( yrhs where w(~j,1)=>w(j) )$
```

Section 5 writes the resulting differential equations for information.

3.3 Extract and scale unstable fast equations

The number of unstable fast equations is the number of terms in the list in `zrhs`.

```
write "no. of unstable fast modes ",nz:=length(zrhs);
```

Extract decay rates Extract the linear decay rates of the fast equations into an array. For each expression in the provided set of right-hand sides:

```
array ratf(nz);
for i:=1:nz do begin
```

For the ith right-hand side get the linear dependence upon `z(i)`, then set other dynamic variables to zero to get just the coefficient.

```
  ratf(i):=coeffn(part(zrhs,i),z(i),1);
  ratf(i):=(ratf(i) where {x(~j)=>0,y(~j)=>0,z(~j)=>0,w(~j)=>0})
```

However, the coefficient may depend upon parameters, so if it is not simply a number, but is a sum, then trawl through the sum looking for a simple number to use as the decay rate.

```
  if not numberp(ratf(i)) then
  if part(ratf(i),0)=plus then begin
    rr:=0;
    for j:=1:arglength(ratf(i)) do
      if numberp(part(ratf(i),j))
      then rr:=part(ratf(i),j);
    ratf(i):=rr;
  end;
```

If all the above has not ended up with a simple number, then exit with an error message.

```
  if numberp(ratf(i))and ratf(i)>0 then
  else begin
    write "***** Error *****
```

```
    Linear coeffs of z-growth must be positive numbers";
    if thecase=webpage then <<
      shut "sdeo.txt"; quit >>;
  end;
```

End the loop over all right-hand sides.

```
end;
```

Flag warning if the linear part not diagonal.

```
for i:=1:nz do for j:=1:nz do if i neq j then begin
  jac:=coeffn(part(zrhs,i),z(j),1);
  if (jac where {x(~k)=>0,y(~k)=>0,z(~k)=>0,w(~k)=>0}) neq 0
  then offdiag:=1$
end;
```

Multiply all the 'nonlinear' terms right-hand sides by `small` so we control the truncation of the asymptotic construction through discarding high powers of `small`. Leave the identified linear decay terms intact. Users could use `small` in their equations for interesting effects.

```
zrhs:=for i:=1:nz collect
  small*part(zrhs,i)+(1-small)*(+ratf(i)*z(i))$
```

Remove the `small` multiplication of noise terms, and instead multiply by `sigma` to empower me to independently control the truncation in noise amplitude.

```
zrhs:=((zrhs where w(~j)=>sigma*w(j,1)/small)
  where w(~j,1)=>w(j))$
```

Section 5 writes the resulting differential equations for information.

Turn off output to file while writing LaTeX.

```
if thecase=webpage then shut "sdeo.txt"$
```

4 Setup LaTeX output using rlfi

Now setup the rlfi package to write a LaTeX version of the output. It is all a bit tricky and underhand, so hope it works. We override some stuff from `rlfi.red`.[5]

Override some rlfi things First, change `name` to get Big delimiters, not left-right delimiters, so LaTeX can break lines.

```
deflist('((!( !\!b!i!g!() (!) !\!b!i!g!)) (!P!I !\!p!i! )
         (!p!i !\!p!i! ) (!E !e) (!I !i) (e !e) (i !i)),'name)$
```

Override the procedure that prints annoying messages about multicharacter symbols. It ends the output of one expression. This is just a copy from `rlfi.red` with the appropriate if-statement deleted.

```
symbolic procedure prinlaend;
<<terpri();
  prin2 "\end{";
  prin2 mstyle!*;
  prin2t "}\par";
  if !*verbatim then
      <<prin2t "\begin{verbatim}";
        prin2t "REDUCE Input:">>;
  ncharspr!*:=0;
  if ofl!* then linelength(car linel!*)
    else laline!*:=cdr linel!*;
  nochar!*:=append(nochar!*,nochar1!*);
  nochar1!*:=nil >>$
```

Override the procedure that outputs the LaTeX preamble upon the command `on latex`.

```
symbolic procedure latexon;
<<!*!*a2sfn:='texaeval;
```

[5] Find it in `reduce-algebra/trunk/packages/misc/rlfi.red`

```
!*raise:=nil;
prin2t "\documentclass[11pt,a5paper]{article}";
prin2t "\usepackage[a5paper,margin=13mm]{geometry}";
prin2t "\usepackage{parskip,time} \raggedright";
prin2t "\def\ou\big(#1,#2,#3\big){{e^{\if#31\else#3\fi t}\star
prin2t "\title{Normal form of your dynamical system}";
prin2t "\author{A. J. Roberts, University of Adelaide\\";
prin2t "\texttt{http://www.maths.adelaide.edu.au/anthony.robert
prin2t "\date{\now, \today}";
prin2t "\begin{document}";
prin2t "\maketitle";
prin2t "\input{sdeo1}";
if !*verbatim then
    <<prin2t "\begin{verbatim}";
      prin2t "REDUCE Input:">>;
put('tex,'rtypefn,'(lambda(x) 'tex)) >>$
```

Use inline math environment so that long lines, the norm, get line breaks. The command `\raggedright` in the LaTeX preamble appears the best option for the line breaking, but `\sloppy` would also work reasonably.

```
mathstyle math;
```

Define names for LaTeX formatting Define some names I use, so that rlfi translates them to Greek characters in the LaTeX.

```
%defid sig,name=sigma;
defid eps,name=epsilon;
defid small,name=varepsilon;
```

Should not need these translation definitions but somehow we do in order for users to get the Greek alphabet to appear. I am puzzled??

```
defid alpha,name=alpha;
defid beta,name=beta;
defid gamma,name=gamma;
defid delta,name=delta;
```

```
defid epsilon,name=epsilon;
defid varepsilon,name=varepsilon;
defid zeta,name=zeta;
defid eta,name=eta;
defid theta,name=theta;
defid vartheta,name=vartheta;
defid iota,name=iota;
defid kappa,name=kappa;
defid lambda,name=lambda;
defid mu,name=mu;
defid nu,name=nu;
defid xi,name=xi;
defid pi,name=pi;
defid varpi,name=varpi;
defid rho,name=rho;
defid varrho,name=varrho;
defid sigma,name=sigma;
defid varsigma,name=varsigma;
defid tau,name=tau;
defid upsilon,name=upsilon;
defid phi,name=phi;
defid varphi,name=varphi;
defid chi,name=chi;
defid psi,name=psi;
defid omega,name=omega;
defid Gamma,name=Gamma;
defid Delta,name=Delta;
defid Theta,name=Theta;
defid Lambda,name=Lambda;
defid Xi,name=Xi;
defid Pi,name=Pi;
defid Sigma,name=Sigma;
defid Upsilon,name=Upsilon;
defid Phi,name=Phi;
defid Psi,name=Psi;
```

```
defid Omega,name=Omega;
```

For the variables names I use, as operators, define how they appear in the LaTeX, and also define that their arguments appear as subscripts.

```
defindex w(down);
defindex x(down);
defindex y(down);
defindex z(down);
defid xx,name="X";
defid yy,name="Y";
defid zz,name="Z";
defindex xx(down);
defindex yy(down);
defindex zz(down);
defindex hh(down);
defindex gg(down);
defindex ff(down);
```

First use these for the specified dynamical system, later use them for the normal form equations.

```
defid hh,name="\dot z";
defid gg,name="\dot y";
defid ff,name="\dot x";
```

The Ornstein–Uhlenbeck operator is to translate into a LaTeX command, see the preamble, that typesets the convolution in a reasonable manner. The definition of the LaTeX command is a bit dodgy as convolutions of convolutions are not printed in the correct order; however, convolutions commute so it does not matter.

```
defid ou,name="\ou";
defindex ou(arg,arg,arg);
```

Write the LaTeX dynamical system Because of the way rfli works, to get good quality output to the LaTeX document, I need to write the algebraic expressions to a file, then read them back in again. While being read back in, I send the output to the LaTeX file. In this convoluted way I avoid extraneous output lines polluting the LaTeX.

Temporarily use these arrays for the right-hand sides of the dynamical system.

```
array ff(m),gg(ny),hh(nz);
```

Write expressions to the file `sdeo.red` for later reading. Prepend the expressions with an instruction to write a heading, and surround the heading with anti-math mode to cancel the math environment that rlfi puts in.

```
out "sdeo.red"$
write "write ""\end{math}
\paragraph{Specified dynamical system}
\begin{math}""$";
for i:=1:m  do write "ff(",i,"):=1*part(xrhs,",i,");";
for i:=1:ny do write "gg(",i,"):=1*part(yrhs,",i,");";
for i:=1:nz do write "hh(",i,"):=1*part(zrhs,",i,");";
write "end;";
shut "sdeo.red";
```

Then switch on LaTeX output before writing to file as this LaTeX file is to be input from the main LaTeX file and hence does not need a header. The header here gets sent to the 'terminal' instead. Then write to `sdeo1.tex` the expressions we stored in `sdeo.red` as nice LaTeX.

```
on latex$
out "sdeo1.tex"$
in "sdeo.red"$
shut "sdeo1.tex"$
off latex$
```

5 Delayed write of text info

Because it is messy to interleave LaTeX and plain output, I delay writing anything much in plain text until here. Here start writing to the text output file `sdeo.txt`; finish writing to file upon success, or otherwise, of the iteration.

```
if thecase=webpage then out "sdeo.txt"$
```

Write the delayed warning message about off-diagonal terms.

```
if offdiag then write "
***** Warning ****
Off diagonal linear terms in y- or z- equations assumed
small.  Answers are rubbish if not asymptotically
appropriate. "$
```

Write the plain text versions of the dynamical system.

```
write "no. of slow modes ",m:=length(xrhs);
for i:=1:m do write "dx(",i,")/dt = ",1*part(xrhs,i);
write "no. of stable fast modes ",ny:=length(yrhs);
for i:=1:ny do write "dy(",i,")/dt = ",1*part(yrhs,i);
write "no. of unstable fast modes ",nz:=length(zrhs);
for i:=1:nz do write "dz(",i,")/dt = ",1*part(zrhs,i);
```

6 Represent the noise

The white noises `w` depend upon time. But we find it useful to discriminate upon the notionally fast time fluctuations of the noise processes, and the notionally ordinary time variations of the dynamic variables x_i, y_i and z_i. Thus introduce a notionally fast time variable `tt`, which depends upon the ordinary time `t`. Equivalently, view `tt`, a sort of 'partial `t`', as representing variations in time independent of those in the variables x_i, y_i and z_i.

```
depend w,tt;
depend tt,t,ttyz;
```

In the construction, convolutions of the noise arise, both backwards over history and forwards to anticipate the noise. For any non-zero parameter μ, define the Ornstein–Uhlenbeck convolution

$$e^{\mu t}\star\phi = \begin{cases} \int_{-\infty}^{t} \exp[\mu(t-\tau)]\phi(\tau)\,d\tau\,, & \mu<0\,,\\ \int_{t}^{+\infty} \exp[\mu(t-\tau)]\phi(\tau)\,d\tau\,, & \mu>0\,,\end{cases} \tag{21}$$

so that the convolution is always with a bounded exponential. Five useful properties of this convolution are

$$e^{\mu t}\star 1 = \frac{1}{|\mu|}\,, \tag{22}$$

$$\frac{d}{dt}e^{\mu t}\star\phi = -\operatorname{sgn}\mu\,\phi + \mu e^{\mu t}\star\phi\,, \tag{23}$$

$$E[e^{\mu t}\star\phi] = e^{\mu t}\star E[\phi]\,, \tag{24}$$

$$E[(e^{\mu t}\star\phi)^2] = \frac{1}{2|\mu|}\,, \tag{25}$$

$$e^{\mu t}\star e^{\nu t}\star = \begin{cases} \frac{1}{|\mu-\nu|}\left[e^{\mu t}\star + e^{\nu t}\star\right], & \mu\nu<0\,,\\ \frac{-\operatorname{sgn}\mu}{\mu-\nu}\left[e^{\mu t}\star - e^{\nu t}\star\right], & \mu\nu>0\ \&\ \mu\neq\nu\,.\end{cases} \tag{26}$$

Also remember that although with $\mu<0$ the convolution $e^{\mu t}\star$ integrates over the past, with $\mu>0$ the convolution $e^{\mu t}\star$ integrates into the future over a time scale of order $1/\mu$.

The operator `ou(f,tt,mu)` represents the convolution $e^{\mu t}\star f$ as defined by (21): called `ou` because it is an Ornstein–Uhlenbeck process. The operator `ou` is 'linear' over fast time `tt` as the convolution only arises from solving PDEs in the operator $\partial_t-\mu$. Code its derivative (23) and its action upon deterministic terms (22):

```
operator ou; linear ou;
let { df(ou(~f,tt,~mu),t)=>-sign(mu)*f+mu*ou(f,tt,mu)
    , ou(1,tt,~mu)=>1/abs(mu)
```

Also code the transform (26) that successive convolutions at different rates may be transformed into several single convolutions.

```
    , ou(ou(~r,tt,~nu),tt,~mu) =>
      (ou(r,tt,mu)+ou(r,tt,nu))/abs(mu-nu) when (mu*nu<0)
```

```
      , ou(ou(~r,tt,~nu),tt,~mu) =>
        -sign(mu)*(ou(r,tt,mu)-ou(r,tt,nu))/(mu-nu)
        when (mu*nu>0)and(mu neq nu)
      };
```

The above properties are *critical*: they must be correct for the results to be correct.

Second, identify the resonant parts, some of which must go into the evolution `gg(i)`, and some into the transform. It depends upon the exponent of `yz` compared to the decay rate of this mode, here `r`.

```
operator reso; linear reso;
let { reso(~a,yz,~r)=>1 when df(a,yz)*yz=r*a
    , reso(~a,yz,~r)=>0 when df(a,yz)*yz neq r*a
    };
```

Lastly, the remaining terms get convolved at the appropriate rate to solve their respective homological equation by the operator `zres`.

```
depend yz,ttyz;
operator zres; linear zres;
let zres(~a,ttyz,~r)=>ou(sign(df(a,yz)*yz/a-r)
    *sub(yz=1,a),tt,df(a,yz)*yz/a-r);
```

7 Solve homological equation with noise

When solving homological equations of the form $F + \xi_t = \text{Res}$ (the resonant case $\mu = 0$), we separate the terms in the right-hand side Res into those that are integrable in fast time, and hence modify the coordinate transform by changing ξ, and those that are not, and hence must remain in the evolution by changing F. the operator `zint` extracts those parts of a term that we know are integrable; the operator `znon` extracts those parts which are not. Note: with more research, more types of terms may be found to be integrable; hence what is extracted by `zint` and what is left by `zint` may change with more research. These transforms are not critical: changing the transforms may change intermediate computations, but as long as the iteration converges, the computer algebra results will be algebraically correct.

```
operator zint; linear zint;
operator znon; linear znon;
```

First, avoid obvious secularity.

```
let { zint(w(~i),tt)=>0, znon(w(~i),tt)=>w(i)
, zint(1,tt)=>0, znon(1,tt)=>1
, zint(w(~i)*~r,tt)=>0, znon(w(~i)*~r,tt)=>w(i)*r
```

Second, by (23) a convolution may be split into an integrable part, and a part in its argument which in turn may be integrable or not.

```
, zint(ou(~r,tt,~mu),tt)=>ou(r,tt,mu)/mu+zint(r,tt)/abs(mu)
, znon(ou(~r,tt,~mu),tt)=>znon(r,tt)/abs(mu)
```

Third, squares of convolutions may be integrated by parts to an integrable term and a part that may have integrable or non-integrable parts.

```
, zint(ou(~r,tt,~mu)^2,tt)=>ou(~r,tt,~mu)^2/(2*mu)
                           +zint(r*ou(r,tt,mu),tt)/abs(mu)
, znon(ou(~r,tt,~mu)^2,tt)=>znon(r*ou(r,tt,mu),tt)/abs(mu)
```

Fourth, different products of convolutions may be similarly separated using integration by parts.

```
, zint(ou(~r,tt,~mu)*ou(~s,tt,~nu),tt)
  =>ou(r,tt,mu)*ou(s,tt,nu)/(mu+nu)
  +zint(sign(mu)*r*ou(s,tt,nu)+sign(nu)*s*ou(r,tt,mu),tt)
  /(mu+nu) when mu+nu neq 0
, znon(ou(~r,tt,~mu)*ou(~s,tt,~nu),tt)=>
  +znon(sign(mu)*r*ou(s,tt,nu)+sign(nu)*s*ou(r,tt,mu),tt)
  /(mu+nu) when mu+nu neq 0
```

However, a zero divisor arises when $\mu + \nu = 0$ in the above. Here code rules to cater for such terms by increasing the depth of convolutions over past history.

```
, zint(ou(~r,tt,~mu)*ou(~s,tt,~nu),tt)=>
  ou(ou(r,tt,-nu),tt,-nu)*ou(s,tt,nu)
  +zint(ou(ou(r,tt,-nu),tt,-nu)*s,tt) when (mu+nu=0)and(nu>0)
, znon(ou(~r,tt,~mu)*ou(~s,tt,~nu),tt)=>
  znon(ou(ou(r,tt,-nu),tt,-nu)*s,tt) when (mu+nu=0)and(nu>0)
```

The above handles quadratic products of convolutions. Presumably, if we seek cubic noise effects then we may need cubic products of convolutions. However, I do not proceed so far and hence terminate the separation rules.

```
};
```

8 Initialise approximate transform

Truncate asymptotic approximation of the coordinate transform depending upon the parameter `toosmall`, up to a maximum of six. Use the 'instant evaluation' property of a loop index to define the truncation so that Reduce omits small terms on the fly.

```
for j:=toosmall:toosmall do let small^j=>0;
```

Variables `x`, `y` and `z` were operators in the specification of the equations. We now want them to store the approximation to the coordinate transform, so clear and reallocate as storage for the normal form expressions.

```
clear x,y,z;
array x(m),y(ny),z(nz);
```

Express the normal form in terms of new evolving variables X_i,Y_i and Z_i, denoted by operators `xx(i)`, `yy(i)` and `zz(i)`, which are nonlinear modifications to x_i, y_i and z_i. The expressions for the normal form SDEs are stored in `ff`, `gg` and `hh`.

```
operator xx; operator yy; operator zz;
depend xx,t; depend yy,t; depend zz,t;
let { df(xx(~i),t)=>ff(i)
    , df(yy(~i),t)=>gg(i)
```

```
    , df(zz(~i),t)=>hh(i) };
```

The first linear approximation is then $x_i \approx X_i$, $y_i \approx Y_i$ and $z_i = Z_i$, such that $\dot X_i \approx 0$, in `ff(i)`, $\dot Y_i \approx -r_i Y_i$, in `gg(i)`, and $\dot Z_i \approx +r_i Z_i$, in `hh(i)`.

```
for i:=1:m  do begin x(i):=xx(i); ff(i):=0; end;
for i:=1:ny do begin y(i):=yy(i); gg(i):=-rate(i)*yy(i); end;
for i:=1:nz do begin z(i):=zz(i); hh(i):=+ratf(i)*zz(i); end;
```

Update the Y_i evolution `gg(i)` and the y_i transform. The residual is of the form of a sum of terms $\prod_j Y_j^{q_j} Z_k^{r_k} \in \text{Res}$. So updates involve dividing by, or convolving with, $\beta_i - \sum_j \beta_j q_j + \sum_k \gamma_k r_k$. First, form the substitutions needed to introduce `yz` to count the number of variables Y_i and Z_i in any given term, weighted according to their rate coefficient in the homological equation.

```
y4y:=for i:=1:ny collect yy(i)=yy(i)*yz^rate(i)$
z4z:=for i:=1:nz collect zz(i)=zz(i)/yz^ratf(i)$
y4y:=append(y4y,z4z)$
```

9 Iterative updates

We iterate to a solution of the governing SDEs to residuals of some order of error. For the moment, iterate for a maximum of nineteen iterations and to the pre-specified errors.

```
for it:=1:maxiter_ do begin
  ok:=1;
```

9.1 Fast stable modes

Compute the residual of each of the y_i SDEs, updating `ok` to track whether all SDEs are satisfied.

```
  for i:=1:ny do begin
    res:=-df(y(i),t)+part(yrhs,i);
```

```
    ok:=if res=0 then ok else 0;
```

Trace print the length of the residuals to check how the iteration is progressing.

```
    write lengthresy:=length(res);
```

Within the loop: first insert the weighted count of Y and Z variables; then split the residual into two parts of possibly resonant, `res0` and the rest, `res1`; then allocate to the evolution or the transform.

```
    res:=sub(y4y,res);
    res0:=reso(res,yz,+rate(i));
    res1:=res-res0*yz^rate(i);
    gg(i):=gg(i)+znon(res0,tt);
    y(i):=y(i) +zint(res0,tt) -zres(res1,ttyz,rate(i));
  end;
```

9.2 Fast unstable modes

Compute the residual of each of the z_i SDEs, updating `ok` to track whether all SDEs are satisfied.

```
  for i:=1:nz do begin
    res:=-df(z(i),t)+part(zrhs,i);
    ok:=if res=0 then ok else 0;
```

Trace print the length of the residuals to check how the iteration is progressing.

```
    write lengthresz:=length(res);
```

Update the Z_i evolution `hh(i)` and the z_i transform. Within the loop: first insert the weighted count of Y and Z variables; then split the residual into two parts of possibly resonant, `res0`, and the rest, `res1`; then allocate to the evolution or the transform.

```
    res:=sub(y4y,res);
    res0:=reso(res,yz,-ratf(i));
    res1:=res-res0/yz^ratf(i);
    hh(i):=hh(i)+znon(res0,tt);
```

```
    z(i):=z(i) +zint(res0,tt) -zres(res1,ttyz,-ratf(i));
  end;
```

9.3 Slow modes

Compute the residual of each of the x SDEs, updating `ok` to track whether all SDEs are satisfied.

```
  for i:=1:m do begin
    res:=-df(x(i),t) +part(xrhs,i);
    ok:=if res=0 then ok else 0;
```

Trace print the length of this residual.

```
    write lengthresx:=length(res);
```

Update the X_i evolution `ff(i)` and the x_i transform. Use the same process as for the fast variables; the difference is that here the mode rate is zero.

```
    res:=sub(y4y,res);
    res0:=reso(res,yz,0);
    res1:=res-res0;
    ff(i):=ff(i)+znon(res0,tt);
    x(i):=x(i) +zint(res0,tt) -zres(res1,ttyz,0);
  end;

  showtime;
  if ok then write "Number of iterations ",
    it:=1000000+it;
end;
```

10 Post-processing

Terminate if the iteration has not converged.

```
if ok=0 then begin
    write "*****Error *****
```

```
    Failed to converge in maximum allowed iterations";
    if thecase=webpage then <<
      shut "sdeo.txt"; quit >>;
end;
```

If converged, then print results.

```
write "***** Success *****";
```

10.1 Plain text output

Print the resultant coordinate transform: but only print to one lower power in `small` and `sigma` in order to keep output relatively small.

```
write "The stochastic/non-autonomous coordinate transform";
for i:=1:nz do begin z(i):=sigma*small*z(i);
                     write z(i):=z(i)/small/sigma; end;
for i:=1:ny do begin y(i):=sigma*small*y(i);
                     write y(i):=y(i)/small/sigma; end;
for i:=1:m  do begin x(i):=sigma*small*x(i);
                     write x(i):=x(i)/small/sigma; end;
```

Lastly print the normal form SDEs: first the fast, second the slow.

```
write "The normal form S/ODEs";
for i:=1:nz do write "dzz(",i,")/dt = ",hh(i);
for i:=1:ny do write "dyy(",i,")/dt = ",gg(i);
for i:=1:m  do write "dxx(",i,")/dt = ",ff(i);
```

Close the output file and no longer quit but move on to LaTeX output.

```
if thecase=webpage then shut "sdeo.txt";
```

10.2 LaTeX output

As before, we have to write expressions to file for later reading so they get printed without extraneous dross in the LaTeX source. First open up the temporary file `sdeo.red` again.

```
out "sdeo.red";
```

Write the stochastic coordinate transform to file, with a heading, and with an anti-math environment to cancel the auto-math of rlfi. For some reason we have to keep these writes short as otherwise it generates a spurious fatal blank line in the LaTeX.

```
write "write ""\end{math}
\paragraph{Time dependent coordinate transform}
\begin{math}"";";
for i:=1:nz do write "z(",i,"):=z(",i,");";
for i:=1:ny do write "y(",i,"):=y(",i,");";
for i:=1:m  do write "x(",i,"):=x(",i,");";
```

Write the resultant stochastic normal form to file, with a heading, and with an anti-math environment to cancel the auto-math of rlfi.

```
write "write ""\end{math}
\paragraph{Result normal form DEs}
\begin{math}"";";
for i:=1:nz do write "hh(",i,"):=hh(",i,");";
for i:=1:ny do write "gg(",i,"):=gg(",i,");";
for i:=1:m  do write "ff(",i,"):=ff(",i,");";
write "end;";
```

Shut the temporary output file.

```
shut "sdeo.red";
```

Get expressions from file and write the main LaTeX file. But first redefine how these names get printed, namely as the normal form time derivatives.

```
defid hh,name="\dot Z";
defid gg,name="\dot Y";
defid ff,name="\dot X";
```

Finally write to the main LaTeX file so switch on latex after starting to write to the file. Then write expressions in `sdeo.red` to `sdeo.tex` as nice LaTeX.

Switch off latex, to get the end of the document, and finish writing.

```
out "sdeo.tex"$
on latex$
in "sdeo.red"$
off latex$
shut "sdeo.tex"$
```

Everything done, so say so and quit.

```
write "***** Finished *****";
if thecase=webpage then quit;

end;
```

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