



# ODE Problems

✓ `SciMLBase.ODEProblem` — Type

Defines an ordinary differential equation (ODE) problem. Documentation Page:  
[https://docs.sciml.ai/DiffEqDocs/stable/types/ode\\_types/](https://docs.sciml.ai/DiffEqDocs/stable/types/ode_types/)

## Mathematical Specification of an ODE Problem

To define an ODE Problem, you simply need to give the function  $f$  and the initial condition  $u_0$  which define an ODE:

$$M \frac{du}{dt} = f(u, p, t)$$

There are two different ways of specifying  $f$ :

$f(du, u, p, t)$ : in-place. Memory-efficient when avoiding allocations. Best option for most cases unless mutation is not allowed.

$f(u, p, t)$ : returning  $du$ . Less memory-efficient way, particularly suitable when mutation is not allowed (e.g. with certain automatic differentiation packages such as Zygote).

$u_0$  should be an `AbstractArray` (or number) whose geometry matches the desired geometry of  $u$ . Note that we are not limited to numbers or vectors for  $u_0$ ; one is allowed to provide  $u_0$  as arbitrary matrices / higher dimension tensors as well.

For the mass matrix  $M$ , see the documentation of `ODEFunction`.

## Problem Type

### Constructors

`ODEProblem` can be constructed by first building an `ODEFunction` or by simply passing the ODE right-hand side to the constructor. The constructors are:

```
ODEProblem(f::ODEFunction,u0,tspan,p=NULLParameters();kwargs...)
ODEProblem{isinplace,specialize}(f,u0,tspan,p=NULLParameters();kwargs...) :
```

Defines the ODE with the specified functions. `isinplace` optionally sets whether the function is inplace or not. This is determined automatically, but not inferred. `specialize` optionally controls the specialization level. See the [specialization levels section of the SciMLBase documentation](#) for more details. The default is `AutoSpecialize`.

For more details on the in-place and specialization controls, see the `ODEFunction` documentation.

Parameters are optional, and if not given, then a `NULLParameters()` singleton will be used which will throw nice errors if you try to index non-existent parameters. Any extra keyword arguments are passed on to the solvers. For example, if you set a `callback` in the problem, then that `callback` will be added in every solve call.

For specifying Jacobians and mass matrices, see the `ODEFunction` documentation.

## Fields

`f`: The function in the ODE.  
`u0`: The initial condition.  
`tspan`: The timespan for the problem.  
`p`: The parameters.  
`kwargs`: The keyword arguments passed onto the solves.

## Example Problem

```
using SciMLBase
function lorenz!(du,u,p,t)
  du[1] = 10.0(u[2]-u[1])
  du[2] = u[1]*(28.0-u[3]) - u[2]
  du[3] = u[1]*u[2] - (8/3)*u[3]
end
u0 = [1.0;0.0;0.0]
tspan = (0.0,100.0)
prob = ODEProblem(lorenz!,u0,tspan)

# Test that it worked
using OrdinaryDiffEq
sol = solve(prob,Tsit5())
using Plots; plot(sol,vars=(1,2,3))
```

## More Example Problems

Example problems can be found in [DiffEqProblemLibrary.jl](#).

To use a sample problem, such as `prob_ode_linear`, you can do something like:

```
#] add ODEProblemLibrary
using ODEProblemLibrary
prob = ODEProblemLibrary.prob_ode_linear
sol = solve(prob)
```

[source](#)

## ✓ [SciMLBase.ODEFunction](#) — Type

```
ODEFunction{iiip,F,TMM,Ta,Tt,TJ,JVP,VJP,JP,SP,TW,TWt,TPJ,S,S2,S3,O,TCV} <: Abstract{iiip,F,TMM,Ta,Tt,TJ,JVP,VJP,JP,SP,TW,TWt,TPJ,S,S2,S3,O,TCV}
```

A representation of an ODE function  $f$ , defined by:

$$M \frac{du}{dt} = f(u, p, t)$$

and all of its related functions, such as the Jacobian of  $f$ , its gradient with respect to time, and more. For all cases,  $u_0$  is the initial condition,  $p$  are the parameters, and  $t$  is the independent variable.

## Constructor

```
ODEFunction{iiip,specialize}(f;
    mass_matrix = __has_mass_matrix(f) ? f.mass_matrix : I,
    analytic = __has_analytic(f) ? f.analytic : nothing,
    tgrad= __has_tgrad(f) ? f.tgrad : nothing,
    jac = __has_jac(f) ? f.jac : nothing,
    jvp = __has_jvp(f) ? f.jvp : nothing,
    vjp = __has_vjp(f) ? f.vjp : nothing,
    jac_prototype = __has_jac_prototype(f) ? f.jac_prototype : nothing,
    sparsity = __has_sparsity(f) ? f.sparsity : jac_prototype,
    paramjac = __has_paramjac(f) ? f.paramjac : nothing,
    syms = __has_syms(f) ? f.syms : nothing,
    indepsym= __has_indepsym(f) ? f.indepsym : nothing,
    paramsyms = __has_paramsyms(f) ? f.paramsyms : nothing,
    colorvec = __has_colorvec(f) ? f.colorvec : nothing,
```

```
sys = __has_sys(f) ? f.sys : nothing)
```

Note that only the function `f` itself is required. This function should be given as `f!(du,u,p,t)` or `du = f(u,p,t)`. See the section on `iip` for more details on in-place vs out-of-place handling.

All of the remaining functions are optional for improving or accelerating the usage of `f`. These include:

`mass_matrix`: the mass matrix  $M$  represented in the ODE function. Can be used to determine that the equation is actually a differential-algebraic equation (DAE) if  $M$  is singular. Note that in this case special solvers are required, see the DAE solver page for more details:

[https://docs.sciml.ai/DiffEqDocs/stable/solvers/dae\\_solve/](https://docs.sciml.ai/DiffEqDocs/stable/solvers/dae_solve/). Must be an `AbstractArray` or an `AbstractSciMLOperator`.

`analytic(u0,p,t)`: used to pass an analytical solution function for the analytical solution of the ODE. Generally only used for testing and development of the solvers.

`tgrad(dT,u,p,t)` or `dT=tgrad(u,p,t)`: returns  $\frac{\partial f(u,p,t)}{\partial t}$

`jac(J,u,p,t)` or `J=jac(u,p,t)`: returns  $\frac{df}{du}$

`jvp(Jv,v,u,p,t)` or `Jv=jvp(v,u,p,t)`: returns the directional derivative  $\frac{df}{du} v$

`vjp(Jv,v,u,p,t)` or `Jv=vjp(v,u,p,t)`: returns the adjoint derivative  $\frac{df}{du}^* v$

`jac_prototype`: a prototype matrix matching the type that matches the Jacobian. For example, if the Jacobian is tridiagonal, then an appropriately sized `Tridiagonal` matrix can be used as the prototype and integrators will specialize on this structure where possible. Non-structured sparsity patterns should use a `SparseMatrixCSC` with a correct sparsity pattern for the Jacobian. The default is `nothing`, which means a dense Jacobian.

`paramjac(pJ,u,p,t)`: returns the parameter Jacobian  $\frac{df}{dp}$ .

`syms`: the symbol names for the elements of the equation. This should match `u0` in size. For example, if `u0 = [0.0,1.0]` and `syms = [:x, :y]`, this will apply a canonical naming to the values, allowing `sol[:x]` in the solution and automatically naming values in plots.

`indepsym`: the canonical naming for the independent variable. Defaults to `nothing`, which internally uses `t` as the representation in any plots.

`paramsyms`: the symbol names for the parameters of the equation. This should match `p` in size. For example, if `p = [0.0, 1.0]` and `paramsyms = [:a, :b]`, this will apply a canonical naming to the values, allowing `sol[:a]` in the solution.

`colorvec`: a color vector according to the `SparseDiffTools.jl` definition for the sparsity pattern of the `jac_prototype`. This specializes the Jacobian construction when using finite differences and automatic differentiation to be computed in an accelerated manner based on

the sparsity pattern. Defaults to nothing, which means a color vector will be internally computed on demand when required. The cost of this operation is highly dependent on the sparsity pattern.

## iip: In-Place vs Out-Of-Place

`iip` is the optional boolean for determining whether a given function is written to be used in-place or out-of-place. In-place functions are `f!(du,u,p,t)` where the return is ignored, and the result is expected to be mutated into the value of `du`. Out-of-place functions are `du=f(u,p,t)`.

Normally, this is determined automatically by looking at the method table for `f` and seeing the maximum number of arguments in available dispatches. For this reason, the constructor `ODEFunction(f)` generally works (but is type-unstable). However, for type-stability or to enforce correctness, this option is passed via `ODEFunction{true}(f)`.

## specialize: Controlling Compilation and Specialization

The `specialize` parameter controls the specialization level of the `ODEFunction` on the function `f`. This allows for a trade-off between compile and run time performance. The available specialization levels are:

`SciMLBase.AutoSpecialize`: this form performs a lazy function wrapping on the functions of the ODE in order to stop recompilation of the ODE solver, but allow for the `prob.f` to stay unwrapped for normal usage. This is the default specialization level and strikes a balance in compile time vs runtime performance.

`SciMLBase.FullSpecialize`: this form fully specializes the `ODEFunction` on the constituent functions that make its fields. As such, each `ODEFunction` in this form is uniquely typed, requiring re-specialization and compilation for each new ODE definition. This form has the highest compile-time at the cost of being the most optimal in runtime. This form should be preferred for long-running calculations (such as within optimization loops) and for benchmarking.

`SciMLBase.NoSpecialize`: this form fully unspecializes the function types in the `ODEFunction` definition by using an `Any` type declaration. As a result, it can result in reduced runtime performance, but is the form that induces the least compile-time.

`SciMLBase.FunctionWrapperSpecialize`: this is an eager function wrapping form. It is unsafe with many solvers, and thus is mostly used for development testing.

For more details, see the [specialization levels section of the SciMLBase documentation](#).

## Fields

The fields of the ODEFunction type directly match the names of the inputs.

## More Details on Jacobians

The following example creates an inplace ODEFunction whose Jacobian is a Diagonal:

```
using LinearAlgebra
f = (du,u,p,t) -> du .= t .* u
jac = (J,u,p,t) -> (J[1,1] = t; J[2,2] = t; J)
jp = Diagonal(zeros(2))
fun = ODEFunction(f; jac=jac, jac_prototype=jp)
```

Note that the integrators will always make a deep copy of `fun.jac_prototype`, so there's no worry of aliasing.

In general, the Jacobian prototype can be anything that has `mul!` defined, in particular sparse matrices or custom lazy types that support `mul!`. A special case is when the `jac_prototype` is a `AbstractSciMLOperator`, in which case you do not need to supply `jac` as it is automatically set to `update_coefficients!`. Refer to the `AbstractSciMLOperators` documentation for more information on setting up time/parameter dependent operators.

## Examples

### Declaring Explicit Jacobians for ODEs

The most standard case, declaring a function for a Jacobian is done by overloading the function `f(du,u,p,t)` with an in-place updating function for the Jacobian: `f_jac(J,u,p,t)` where the value type is used for dispatch. For example, take the Lotka-Volterra model:

```
function f(du,u,p,t)
    du[1] = 2.0 * u[1] - 1.2 * u[1]*u[2]
    du[2] = -3 * u[2] + u[1]*u[2]
end
```

To declare the Jacobian, we simply add the dispatch:

```
function f_jac(J,u,p,t)
    J[1,1] = 2.0 - 1.2 * u[2]
    J[1,2] = -1.2 * u[1]
    J[2,1] = 1 * u[2]
    J[2,2] = -3 + u[1]
    nothing
```

```
end
```

Then we can supply the Jacobian with our ODE as:

```
ff = ODEFunction(f;jac=f_jac)
```

and use this in an ODEProblem:

```
prob = ODEProblem(ff,ones(2),(0.0,10.0))
```

## Symbolically Generating the Functions

See the `modelingtoolkitize` function from [ModelingToolkit.jl](#) for automatically symbolically generating the Jacobian and more from the numerically-defined functions.

# Solution Type

✓ [SciMLBase.ODESolution](#) — Type

```
struct ODESolution{T, N, uType, uType2, DType, tType, rateType, P, A, IType, S, A}
```

Representation of the solution to an ordinary differential equation defined by an ODEProblem.

## DESolution Interface

For more information on interacting with DESolution types, check out the Solution Handling page of the DifferentialEquations.jl documentation.

<https://docs.sciml.ai/DiffEqDocs/stable/basics/solution/>

## Fields

`u`: the representation of the ODE solution. Given as an array of solutions, where `u[i]` corresponds to the solution at time `t[i]`. It is recommended in most cases one does not access `sol.u` directly and instead use the array interface described in the Solution Handling page of the DifferentialEquations.jl documentation.

`t`: the time points corresponding to the saved values of the ODE solution.

`prob`: the original ODEProblem that was solved.

alg: the algorithm type used by the solver.

stats: statistics of the solver, such as the number of function evaluations required, number of Jacobians computed, and more.

retcode: the return code from the solver. Used to determine whether the solver solved successfully, whether it terminated early due to a user-defined callback, or whether it exited due to an error. For more details, see [the return code documentation](#).

## Example Problems

Example problems can be found in [DiffEqProblemLibrary.jl](#).

To use a sample problem, such as `prob_ode_linear`, you can do something like:

```
#] add DiffEqProblemLibrary
using DiffEqProblemLibrary.ODEProblemLibrary
# load problems
ODEProblemLibrary.importodeproblems()
prob = ODEProblemLibrary.prob_ode_linear
sol = solve(prob)
```

### ✓ [ODEProblemLibrary.prob\\_ode\\_linear](#) — Constant

Linear ODE

$$\frac{du}{dt} = \alpha u$$

with initial condition  $u_0 = \frac{1}{2}$ ,  $\alpha = 1.01$ , and solution

$$u(t) = u_0 e^{\alpha t}$$

with `Float64`s. The parameter is  $\alpha$

### ✓ [ODEProblemLibrary.prob\\_ode\\_2Dlinear](#) — Constant

4x2 version of the Linear ODE



$$\frac{du}{dt} = \alpha u$$

with initial condition  $u_0$  as all uniformly distributed random numbers,  $\alpha = 1.01$ , and solution

$$u(t) = u_0 e^{\alpha t}$$

with Float64s

✓ [ODEProblemLibrary.prob\\_ode\\_bigfloatlinear](#) – Constant

Linear ODE

$$\frac{du}{dt} = \alpha u$$

with initial condition  $u_0 = \frac{1}{2}$ ,  $\alpha = 1.01$ , and solution

$$u(t) = u_0 e^{\alpha t}$$

with BigFloats

✓ [ODEProblemLibrary.prob\\_ode\\_bigfloat2Dlinear](#) – Constant

4x2 version of the Linear ODE

$$\frac{du}{dt} = \alpha u$$

with initial condition  $u_0$  as all uniformly distributed random numbers,  $\alpha = 1.01$ , and solution

$$u(t) = u_0 e^{\alpha t}$$

with BigFloats

✓ [ODEProblemLibrary.prob\\_ode\\_large2Dlinear](#) – Constant

100x100 version of the Linear ODE

$$\frac{du}{dt} = \alpha u$$

with initial condition  $u_0$  as all uniformly distributed random numbers,  $\alpha = 1.01$ , and solution

$$u(t) = u_0 e^{\alpha t}$$

with Float64s

✓ `ODEProblemLibrary.prob_ode_2Dlinear_notinplace` – Constant

4x2 version of the Linear ODE

$$\frac{du}{dt} = \alpha u$$

with initial condition  $u_0$  as all uniformly distributed random numbers,  $\alpha = 1.01$ , and solution

$$u(t) = u_0 e^{\alpha t}$$

on Float64. Purposefully not in-place as a test.

✓ `ODEProblemLibrary.prob_ode_lotkavolterra` – Constant

Lotka-Volterra Equations (Non-stiff)

$$\frac{dx}{dt} = ax - bxy$$

$$\frac{dy}{dt} = -cy + dxy$$

with initial condition  $x = y = 1$

Fitzhugh-Nagumo (Non-stiff)

$$\frac{dv}{dt} = v - \frac{v^3}{3} - w + I_{est}$$

$$\tau \frac{dw}{dt} = v + a - bw$$

with initial condition  $v = w = 1$

The ThreeBody problem as written by Hairer: (Non-stiff)

$$\frac{dy_1}{dt} = y_1 + 2 \frac{dy_2}{dt} - \bar{\mu} \frac{y_1 + \mu}{D_1} - \mu \frac{y_1 - \bar{\mu}}{D_2}$$

$$\frac{dy_2}{dt} = y_2 - 2 \frac{dy_1}{dt} - \bar{\mu} \frac{y_2}{D_1} - \mu \frac{y_2}{D_2}$$

$$D_1 = ((y_1 + \mu)^2 + y_2^2)^{3/2}$$

$$D_2 = ((y_1 - \bar{\mu})^2 + y_2^2)^{3/2}$$

$$\mu = 0.012277471$$

$$\bar{\mu} = 1 - \mu$$

From Hairer Norsett Wanner Solving Ordinary Differential Equations I - Nonstiff Problems Page 129

Usually solved on  $t_0 = 0.0$  and  $T = 17.0652165601579625588917206249$  Periodic with that setup.

## Pleiades Problem (Non-stiff)

$$\frac{d^2 x_i}{dt^2} = \sum_{j \neq i} m_j (x_j - x_i) / r_{ij}$$

$$\frac{d^2 y_i}{dt^2} = \sum_{j \neq i} m_j (y_j - y_i) / r_{ij}$$

where

$$r_{ij} = ((x_i - x_j)^2 + (y_i - y_j)^2)^{3/2}$$

and initial conditions are

$$x_1(0) = 3$$

$$x_2(0) = 3$$

$$x_3(0) = -1$$

$$x_4(0) = -3$$

$$x_5(0) = 2$$

$$x_6(0) = -2$$

$$x_7(0) = 2$$

$$y_1(0) = 3$$

$$y_2(0) = -3$$

$$y_3(0) = 2$$

$$y_4(0) = 0$$

$$y_5(0) = 0$$

$$y_6(0) = -4$$

$$y_7(0) = 4$$

and with  $\frac{dx_i(0)}{dt} = \frac{dy_i(0)}{dt} = 0$  except for

$$\frac{dx_6(0)}{dt} = 1.75$$

$$\frac{dx_7(0)}{dt} = -1.5$$

$$\frac{dy_4(0)}{dt} = -1.25$$

$$\frac{dy_5(0)}{dt} = 1$$

From Hairer Norsett Wanner Solving Ordinary Differential Equations I - Nonstiff Problems Page 244

Usually solved from 0 to 3.

✓ [ODEProblemLibrary.prob\\_ode\\_vanderpol](#) — Constant

Van der Pol Equations

$$\frac{dx}{dt} = y$$

$$\frac{dy}{dt} = \mu((1 - x^2)y - x)$$

with  $\mu = 1.0$  and  $u_0 = [0, \ 3]$

Non-stiff parameters.

✓ [ODEProblemLibrary.prob\\_ode\\_vanderpol\\_stiff](#) – Constant

Van der Pol Equations

$$\frac{dx}{dt} = y$$

$$\frac{dy}{dt} = \mu((1 - x^2)y - x)$$

with  $\mu = 10^6$  and  $u_0 = [0, \ 3]$

Stiff parameters.

✓ [ODEProblemLibrary.prob\\_ode\\_rober](#) – Constant

The Robertson biochemical reactions: (Stiff)

$$\frac{dy_1}{dt} = -k_1y_1 + k_3y_2y_3$$

$$\frac{dy_2}{dt} = k_1y_1 - k_2y_2^2 - k_3y_2y_3$$

$$\frac{dy_3}{dt} = k_2y_2^2$$

where  $k_1 = 0.04$ ,  $k_2 = 3 \times 10^7$ ,  $k_3 = 10^4$ . For details, see:

Hairer Norsett Wanner Solving Ordinary Differential Equations I - Nonstiff Problems Page 129

Usually solved on  $[0, 1e11]$

✓ [ODEProblemLibrary.prob\\_ode\\_rigidbody](#) — Constant

Rigid Body Equations (Non-stiff)

$$\frac{dy_1}{dt} = I_1 y_2 y_3$$

$$\frac{dy_2}{dt} = I_2 y_1 y_3$$

$$\frac{dy_3}{dt} = I_3 y_1 y_2$$

with  $I_1 = -2$ ,  $I_2 = 1.25$ , and  $I_3 = -1/2$ .

The initial condition is  $y = [1.0; 0.0; 0.9]$ .

From Solving Differential Equations in R by Karline Soetaert

or Hairer Norsett Wanner Solving Ordinary Differential Equations I - Nonstiff Problems Page 244

Usually solved from 0 to 20.

✓ [ODEProblemLibrary.prob\\_ode\\_hires](#) — Constant

Hires Problem (Stiff)

It is in the form of

$$\frac{dy}{dt} = f(y)$$

with

$$y(0) = y_0, \quad y \in \mathbb{R}^8, \quad 0 \leq t \leq 321.8122$$

where  $f$  is defined by

$$f(y) = \begin{pmatrix} -1.71y_1 & +0.43y_2 & +8.32y_3 & +0.0007y_4 & & & & \\ 1.71y_1 & -8.75y_2 & & & & & & \\ -10.03y_3 & +0.43y_4 & +0.035y_5 & & & & & \\ 8.32y_2 & +1.71y_3 & -1.12y_4 & & & & & \\ -1.745y_5 & +0.43y_6 & +0.43y_7 & & & & & \\ -280y_6y_8 & +0.69y_4 & +1.71y_5 & -0.43y_6 & +0.69y_7 & & & \\ 280y_6y_8 & -1.81y_7 & & & & & & \\ -280y_6y_8 & +1.81y_7 & & & & & & \end{pmatrix}$$

Reference: [demohires.pdf](#) Notebook: [Hires.ipynb](#)

### ✓ `ODEProblemLibrary.prob_ode_orego` – Constant

Orego Problem (Stiff)

It is in the form of  $\frac{dy}{dt} = f(y)$ ,  $y(0) = y_0$ , with

$$y \in \mathbb{R}^3, \quad 0 \leq t \leq 360$$

where  $f$  is defined by

$$f(y) = \begin{pmatrix} s(y_2 - y_1(1 - qy_1 - y_2)) \\ (y_3 - y_2(1 + y_1))/s \\ w(y_1 - y_3) \end{pmatrix}$$

where  $s = 77.27$ ,  $w = 0.161$  and  $q = 8.375 \cdot 10^{-6}$ .

Reference: [demoorego.pdf](#) Notebook: [Orego.ipynb](#)

### ✓ `ODEProblemLibrary.prob_ode_pollution` – Constant

Pollution Problem (Stiff)

This IVP is a stiff system of 20 non-linear Ordinary Differential Equations. It is in the form of

$$\frac{dy}{dt} = f(y)$$

with



$$y(0) = y_0, \quad y \in \mathbb{R}^2, \quad 0 \leq t \leq 60$$

where  $f$  is defined by

$$f(y) = \begin{pmatrix} -\sum_{j \in \{1,10,14,23,24\}} r_j + \sum_{j \in \{2,3,9,11,12,22,25\}} r_j \\ -r_2 - r_3 - r_9 - r_1^2 + r_1 + r_{21} \\ -r_{15} + r_1 + r_{17} + r_{19} + r_{22} \\ -r_2 - r_{16} - r_{17} - r_{23} + r_{15} \\ -r_3 + 2r_4 + r_6 + r_7 + r_{13} + r_{20} \\ -r_6 - r_8 - r_{14} - r_{20} + r_3 + 2r_{18} \\ -r_4 - r_5 - r_6 + r_{13} \\ r_4 + r_5 + r_6 + r_7 \\ -r_7 - r_8 \\ -r_{12} + r_7 + r_9 \\ -r_9 - r_{10} + r_8 + r_{11} \\ r_9 \\ -r_{11} + r_{10} \\ -r_{13} + r_{12} \\ r_{14} \\ -r_{18} - r_{19} + r_{16} \\ -r_{20} \\ r_{20} \\ -r_{21} - r_{22} - r_{24} + r_{23} + r_{25} \\ -r_{25} + r_{24} \end{pmatrix}$$

with the initial condition of

$$y_0 = (0, 0.2, 0, 0.04, 0, 0, 0.1, 0.3, 0.01, 0, 0, 0, 0, 0, 0, 0.007, 0, 0, 0)^T$$

Analytical Jacobian is included.

Reference: [pollu.pdf](#) Notebook: [Pollution.ipynb](#)

✓ [ODEProblemLibrary.prob\\_ode\\_nonlinchem](#) — Constant

Nonlinear system of reactions with an analytical solution

$$\frac{dy_1}{dt} = -y_1$$

$$\frac{dy_2}{dt} = y_1 - y_2^2$$

$$\frac{dy_3}{dt} = y_2^2$$

with initial condition  $y = [1; 0; 0]$  on a time span of  $t \in (0, 20)$

From

Liu, L. C., Tian, B., Xue, Y. S., Wang, M., & Liu, W. J. (2012). Analytic solution for a nonlinear chemistry system of ordinary differential equations. *Nonlinear Dynamics*, 68(1-2), 17-21.

The analytical solution is implemented, allowing easy testing of ODE solvers.

✓ [ODEProblemLibrary.prob\\_ode\\_brusselator\\_1d](#) – Constant

1D Brusselator

$$\frac{\partial u}{\partial t} = A + u^2 v - (B + 1)u + \alpha \frac{\partial^2 u}{\partial x^2}$$

$$\frac{\partial v}{\partial t} = Bu - u^2 v + \alpha \frac{\partial^2 u}{\partial x^2}$$

and the initial conditions are

$$u(x, 0) = 1 + \sin(2\pi x)$$

$$v(x, 0) = 3$$

with the boundary condition

$$u(0, t) = u(1, t) = 1$$

$$v(0, t) = v(1, t) = 3$$

From Hairer Norsett Wanner Solving Ordinary Differential Equations II - Stiff and Differential-

✓ [ODEProblemLibrary.prob\\_ode\\_brusselator\\_2d](#) — Constant

2D Brusselator

$$\frac{\partial u}{\partial t} = 1 + u^2 v - 4.4u + \alpha \left( \frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} \right) + f(x, y, t)$$

$$\frac{\partial v}{\partial t} = 3.4u - u^2 v + \alpha \left( \frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} \right)$$

where

$$f(x, y, t) = \begin{cases} 5 & \text{if } (x - 0.3)^2 + (y - 0.6)^2 \leq 0.1^2 \text{ and } t \geq 1.1 \\ 0 & \text{else} \end{cases}$$

and the initial conditions are

$$u(x, y, 0) = 22 \cdot y(1 - y)^{3/2}$$

$$v(x, y, 0) = 27 \cdot x(1 - x)^{3/2}$$

with the periodic boundary condition

$$u(x + 1, y, t) = u(x, y, t)$$

$$u(x, y + 1, t) = u(x, y, t)$$

From Hairer Norsett Wanner Solving Ordinary Differential Equations II - Stiff and Differential-Algebraic Problems Page 152

✓ [ODEProblemLibrary.prob\\_ode\\_filament](#) — Constant

Filament PDE Discretization

Notebook: [Filament.ipynb](#)

In this problem is a real-world biological model from a paper entitled Magnetic dipole with a flexible tail as a self-propelling microdevice. It is a system of PDEs representing a Kirchhoff model of an elastic rod, where the equations of motion are given by the Rouse approximation with free boundary conditions.

✓ [ODEProblemLibrary.prob\\_ode\\_thomas](#) — Constant

Thomas' cyclically symmetric attractor equations

$$\frac{dx(t)}{dt} = \sin(y(t)) - bx(t) \quad (2)$$

$$\frac{dy(t)}{dt} = \sin(z(t)) - by(t) \quad (3)$$

$$\frac{dz(t)}{dt} = \sin(x(t)) - bz(t)$$

[Reference](#)

[Wikipedia](#)

✓ [ODEProblemLibrary.prob\\_ode\\_lorenz](#) — Constant

Lorenz equations

$$\frac{dx(t)}{dt} = (-x(t) + y(t))\sigma \quad (5)$$

$$\frac{dy(t)}{dt} = -y(t) + x(t)(-z(t) + \rho) \quad (6)$$

$$\frac{dz(t)}{dt} = x(t)y(t) - z(t)\beta$$

[Reference](#)

[Wikipedia](#)

✓ [ODEProblemLibrary.prob\\_ode\\_aizawa](#) — Constant

## Aizawa equations

$$\frac{dx(t)}{dt} = (-b + z(t))x(t) - dy(t) \quad (8)$$

$$\frac{dy(t)}{dt} = (-b + z(t))y(t) + dx(t) \quad (9)$$

$$\frac{dz(t)}{dt} = c + az(t) - \frac{1}{3}(z(t))^3 + (-1 - ez(t))\left((x(t))^2 + (y(t))^2\right) + (x(t))^3 fz(t)$$

## Reference

### ✓ `ODEProblemLibrary.prob_ode_dadras` – Constant

## Dadras equations

$$\frac{dx(t)}{dt} = y(t) - ax(t) + by(t)z(t) \quad (11)$$

$$\frac{dy(t)}{dt} = z(t) + cy(t) - x(t)z(t) \quad (12)$$

$$\frac{dz(t)}{dt} = -ez(t) + dx(t)y(t)$$

## Reference

### ✓ `ODEProblemLibrary.prob_ode_chen` – Constant

## chen equations

$$\frac{dx(t)}{dt} = a(-x(t) + y(t)) \quad (14)$$

$$\frac{dy(t)}{dt} = (-a + c)x(t) + cy(t) - x(t)z(t) \quad (15)$$

$$\frac{dz(t)}{dt} = -bz(t) + x(t)y(t)$$

## Reference

✓ `ODEProblemLibrary.prob_ode_rossler` – Constant

rossler equations

$$\frac{dx(t)}{dt} = -y(t) - z(t) \quad (17)$$

$$\frac{dy(t)}{dt} = x(t) + ay(t) \quad (18)$$

$$\frac{dz(t)}{dt} = b + (-c + x(t))z(t)$$

[ReferenceWikipedia](#)

✓ `ODEProblemLibrary.prob_ode_rabinovich_fabrikant` – Constant

rabinovich\_fabrikant equations

$$\frac{dx(t)}{dt} = bx(t) + \left(-1 + z(t) + (x(t))^2\right)y(t) \quad (20)$$

$$\frac{dy(t)}{dt} = by(t) + x(t) \left(1 + 3z(t) - (x(t))^2\right) \quad (21)$$

$$\frac{dz(t)}{dt} = -2(a + x(t)y(t))z(t)$$

[Reference](#)

✓ `ODEProblemLibrary.prob_ode_sprott` – Constant

sprott equations

$$\frac{dx(t)}{dt} = y(t) + x(t)z(t) + ax(t)y(t) \quad (23)$$

$$\frac{dy(t)}{dt} = 1 + y(t)z(t) - (x(t))^2b \quad (24)$$

$$\frac{dz(t)}{dt} = x(t) - (x(t))^2 - (y(t))^2$$

hindmarsh\_rose equations

$$\frac{dx(t)}{dt} = i + y(t) - z(t) + (x(t))^2 b - (x(t))^3 a \quad (26)$$

$$\frac{dy(t)}{dt} = c - y(t) - (x(t))^2 d \quad (27)$$

$$\frac{dz(t)}{dt} = r(-z(t) + s(-xr + x(t)))$$