

Lotka-Volterra Work-Precision Diagrams

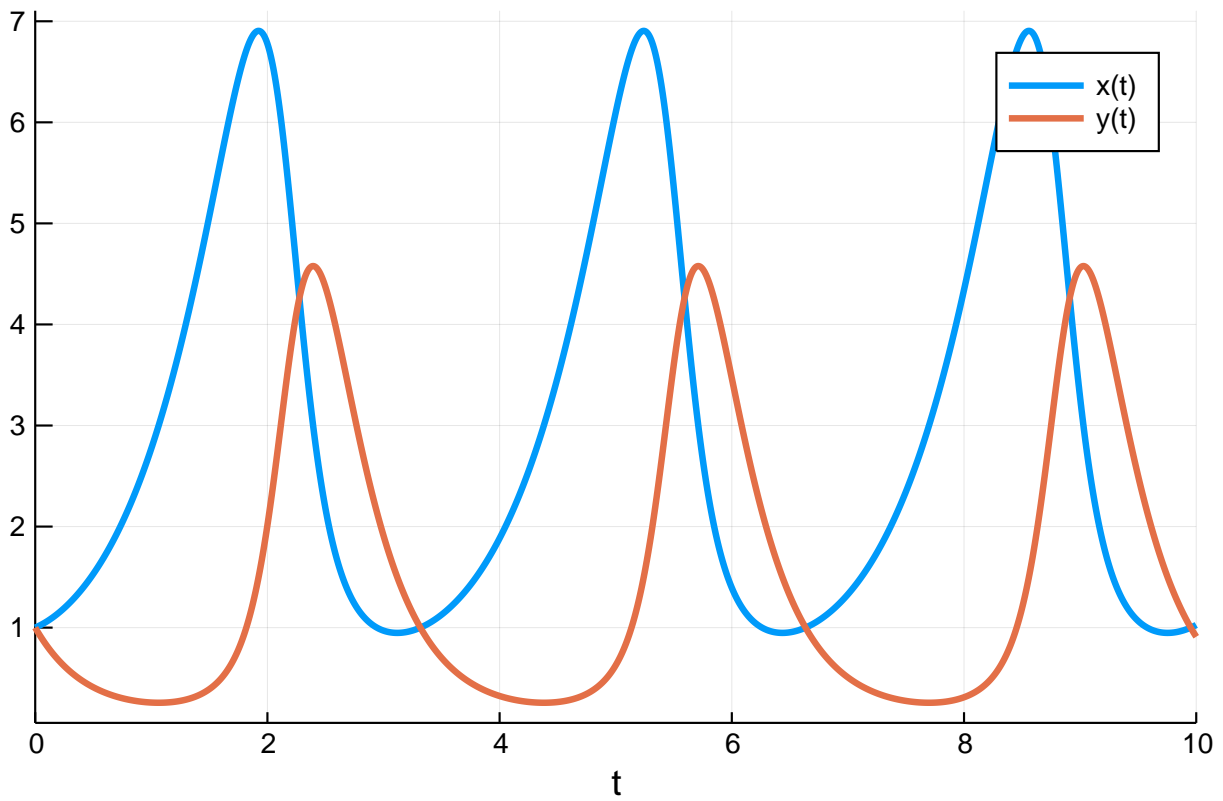
Chris Rackauckas

June 12, 2019

0.1 Lotka-Volterra

The purpose of this problem is to test the performance on easy problems. Since it's periodic, the error is naturally low, and so most of the difference will come down to startup times and, when measuring the interpolations, the algorithm choices.

```
using OrdinaryDiffEq, ParameterizedFunctions, ODE, ODEInterfaceDiffEq, LSODA,  
    Sundials, DiffEqDevTools  
  
f = @ode_def LotkaVolterra begin  
    dx = a*x - b*x*y  
    dy = -c*y + d*x*y  
end a b c d  
  
p = [1.5,1.0,3.0,1.0]  
prob = ODEProblem(f,[1.0;1.0],(0.0,10.0),p)  
  
abstols = 1.0 ./ 10.0 .^ (6:13)  
reltols = 1.0 ./ 10.0 .^ (3:10);  
sol = solve(prob,Vern7(), abstol=1/10^14, reltol=1/10^14)  
test_sol = TestSolution(sol)  
using Plots; gr()  
  
plot(sol)
```

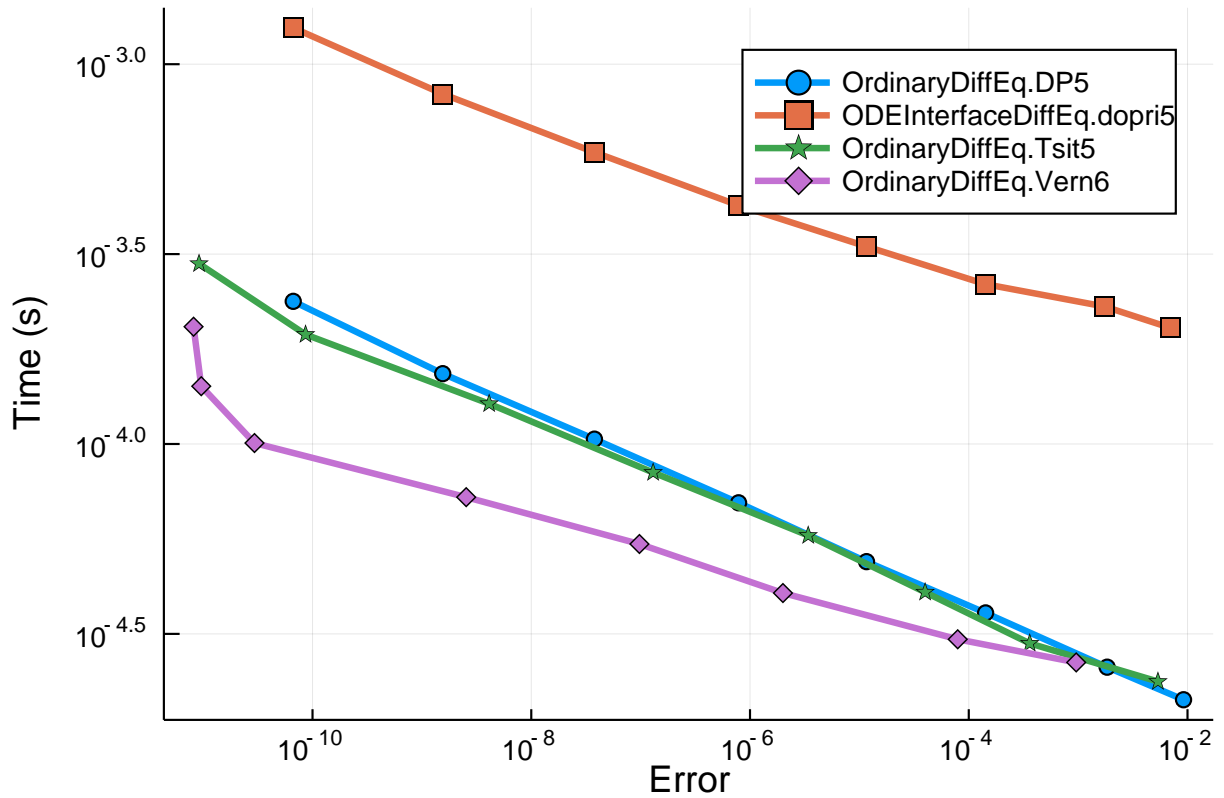


0.1.1 Low Order

```

setups = [Dict(:alg=>DP5())
           #Dict(:alg=>ode45()) # fail
           Dict(:alg=>dopri5())
           Dict(:alg=>Tsit5())
           Dict(:alg=>Vern6())
]
wp =
  WorkPrecisionSet(prob, abstols, reltols, setups; appxsol=test_sol, save_everystep=false, maxiters=10000,
  plot(wp)

```



Here we see the OrdinaryDiffEq.jl algorithms once again far in the lead.

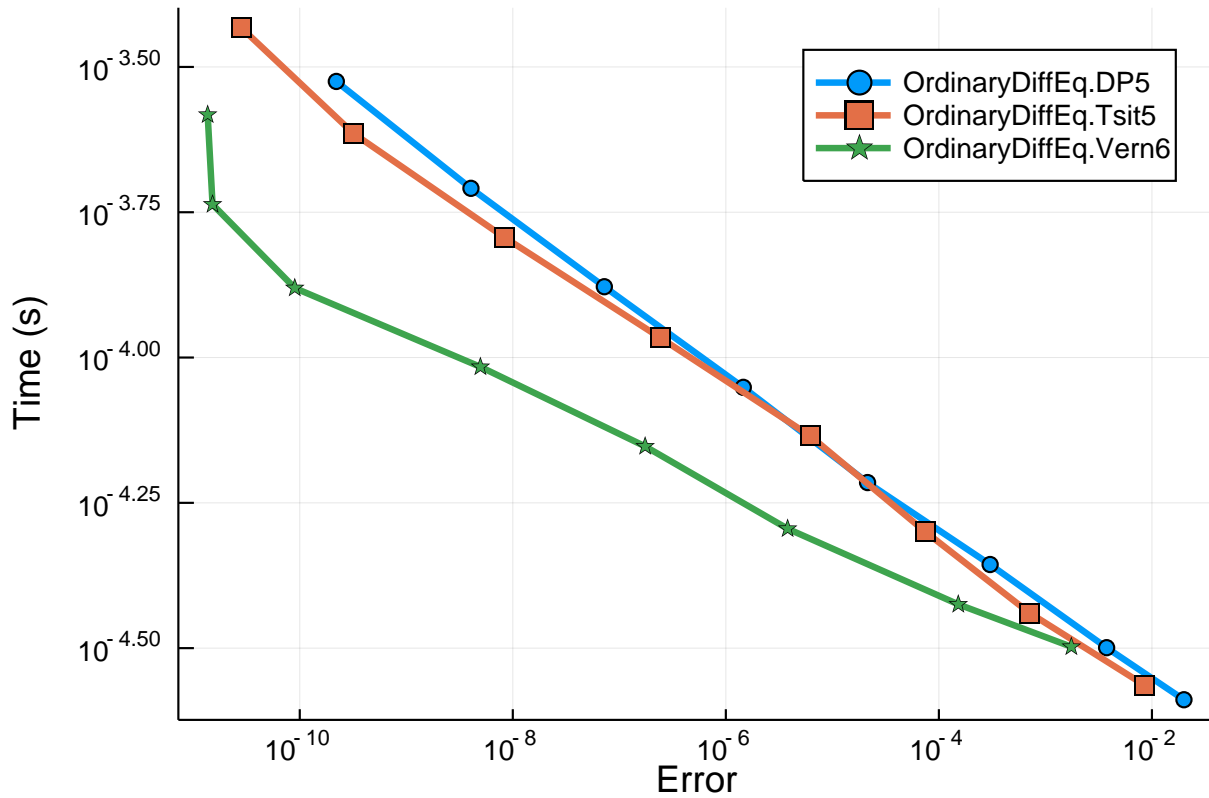
0.1.2 Interpolation Error

Since the problem is periodic, the real measure of error is the error throughout the solution.

```

setups = [Dict(:alg=>DP5())
           #Dict(:alg=>ode45())
           Dict(:alg=>Tsit5())
           Dict(:alg=>Vern6())
]
wp =
  WorkPrecisionSet(prob, abstols, reltols, setups; appxsol=test_sol, maxiters=10000, error_estimate=:L2, de
plot(wp)

```



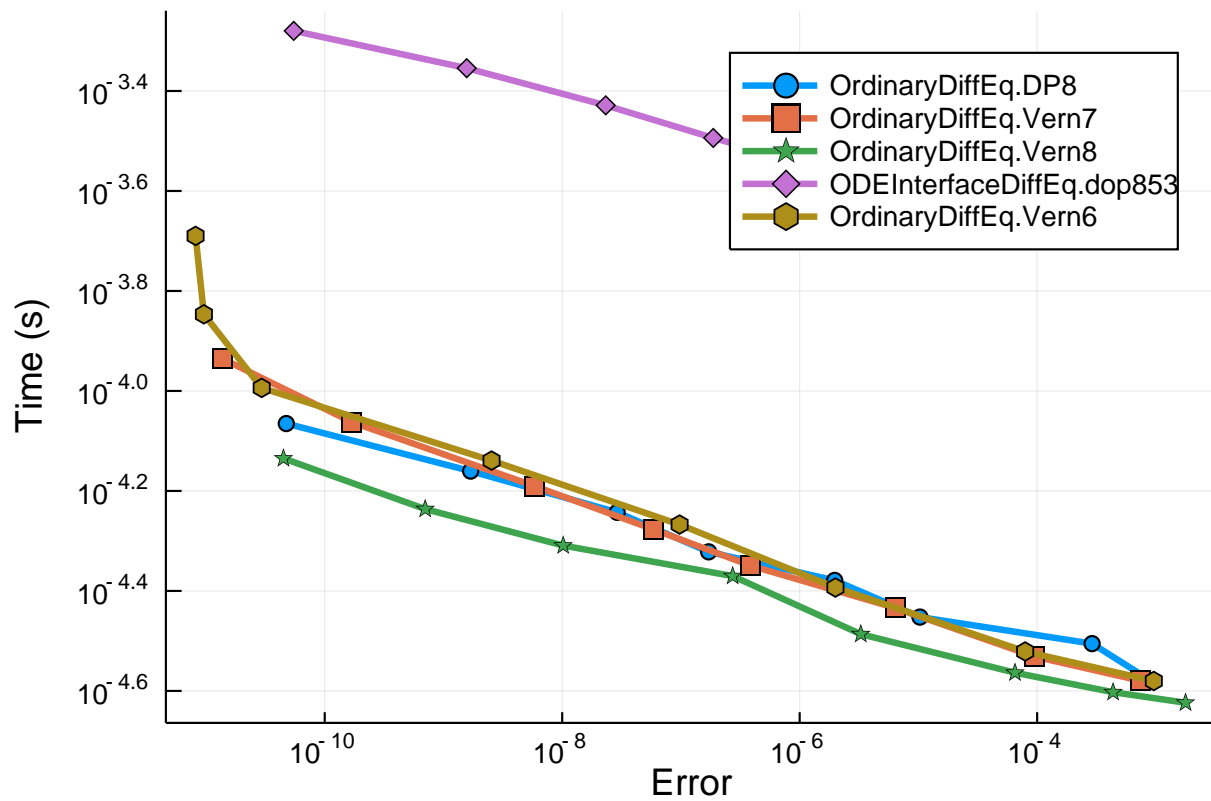
Here we see the power of algorithm specific interpolations. The ODE.jl algorithm is only able to reach 10^{-7} error even at a tolerance of 10^{-13} , while the DifferentialEquations.jl algorithms are below 10^{-10}

0.2 Higher Order

```

setups = [Dict{:alg=>DP8()}
           #Dict{:alg=>ode78()} # fails
           Dict{:alg=>Vern7()}
           Dict{:alg=>Vern8()}
           Dict{:alg=>dop853()}
           Dict{:alg=>Vern6()}
]
wp =
  WorkPrecisionSet(prob, abstols, reltols, setups; appxsol=test_sol, save_everystep=false, maxiters=1000, n
plot(wp)

```

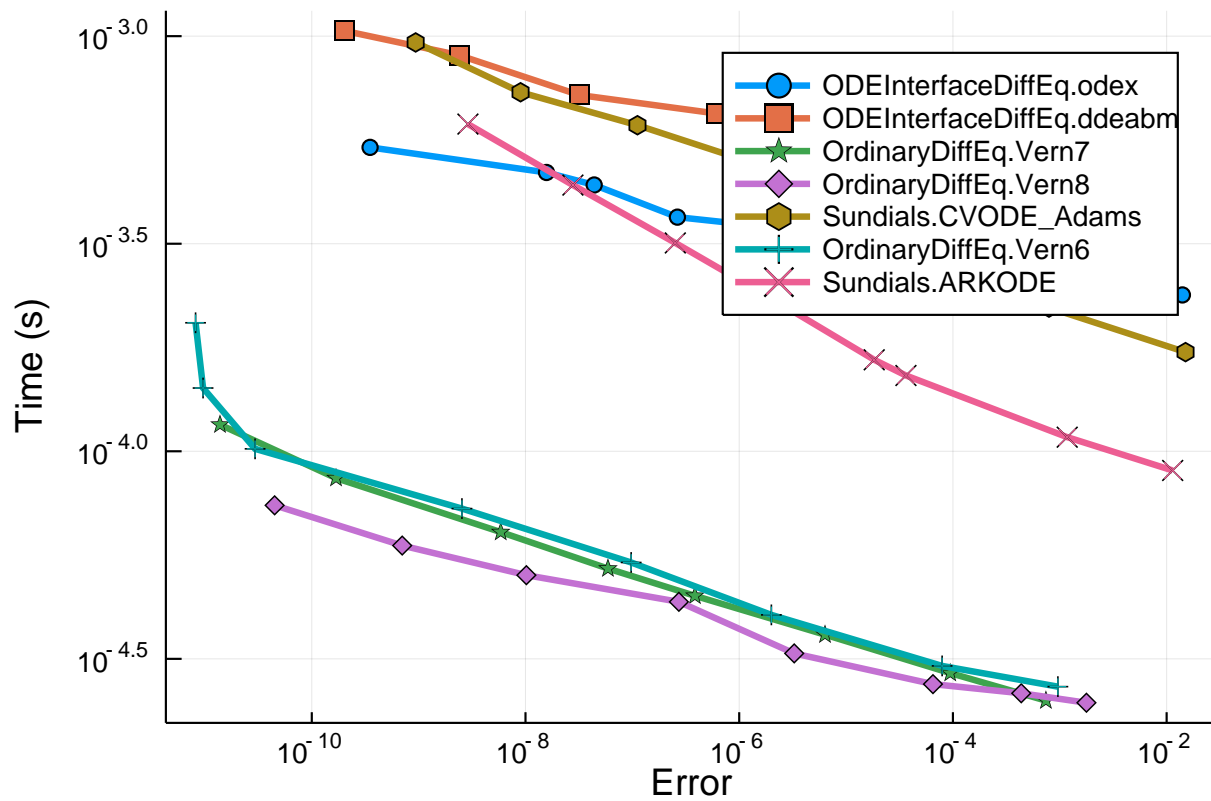


```

setups = [Dict(:alg=>odex())
          Dict(:alg=>ddeabm())
          Dict(:alg=>Vern7())
          Dict(:alg=>Vern8())
          Dict(:alg=>CVODE_Adams())
          #Dict(:alg=>lsoda())
          Dict(:alg=>Vern6())
          Dict(:alg=>ARKODE(Sundials.Explicit(),order=6))
        ]

wp =
  WorkPrecisionSet(prob, abstols, reltols, setups; appxsol=test_sol, save_everystep=false, maxiters=1000, n
plot(wp)

```

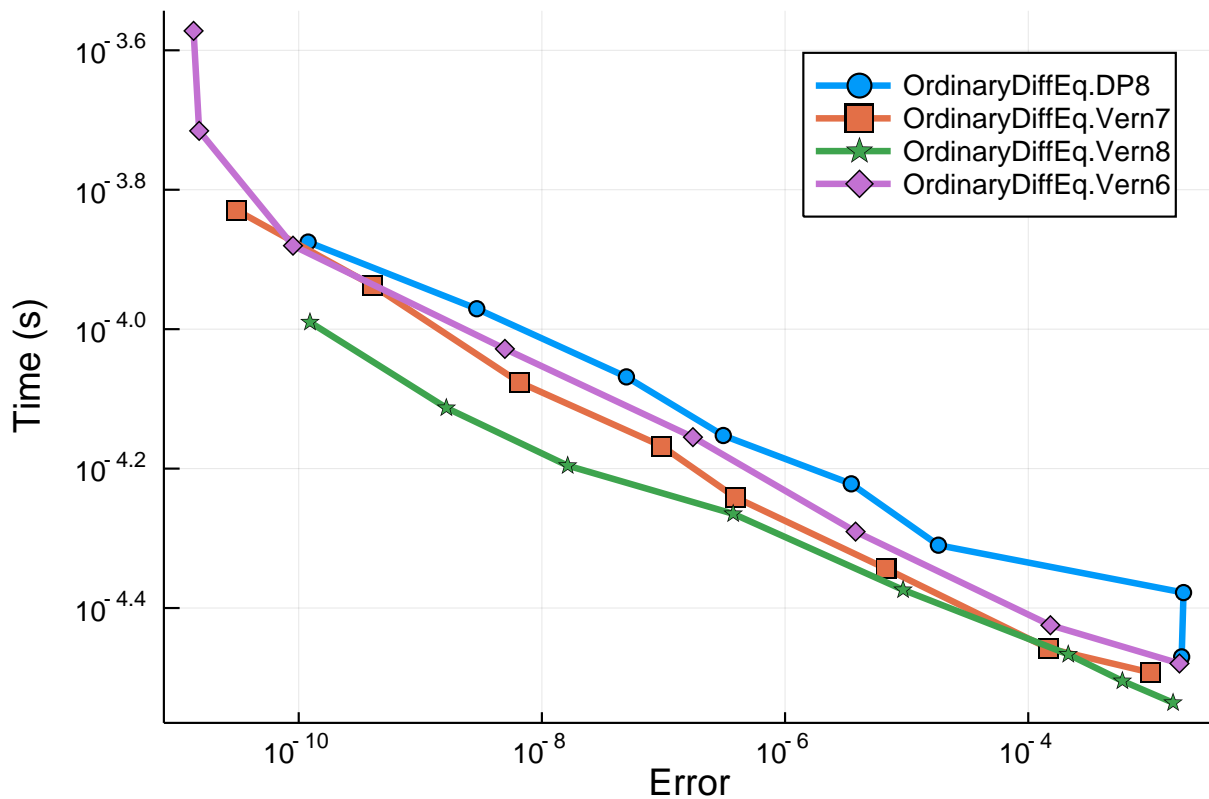


Again we look at interpolations:

```

setups = [Dict(:alg=>DP8())
           #Dict(:alg=>ode78())
           Dict(:alg=>Vern7())
           Dict(:alg=>Vern8())
           Dict(:alg=>Vern6())
]
wp =
  WorkPrecisionSet(prob, abstols, reltols, setups; appxsol=test_sol, dense=true, maxiters=1000, error_estim
plot(wp)

```



Again, the ODE.jl algorithms suffer when measuring the interpolations due to relying on an order 3 Hermite polynomial instead of an algorithm-specific order matching interpolation which uses the timesteps.

0.3 Conclusion

The OrdinaryDiffEq.jl are quicker and still solve to a much higher accuracy, especially when the interpolations are involved. ODE.jl errors a lot.

```
using DiffEqBenchmarks
DiffEqBenchmarks.bench_footer(WEAVE_ARGS[:folder], WEAVE_ARGS[:file])
```

0.4 Appendix

These benchmarks are a part of the DiffEqBenchmarks.jl repository, found at: <https://github.com/JuliaDiffEq/DiffEqBenchmarks.jl>

To locally run this tutorial, do the following commands:

```
using DiffEqBenchmarks
DiffEqBenchmarks.weave_file("NonStiffODE", "LotkaVolterra_wpd.jmd")
```

Computer Information:

```
Julia Version 1.1.0
Commit 80516ca202 (2019-01-21 21:24 UTC)
Platform Info:
  OS: Linux (x86_64-pc-linux-gnu)
```

CPU: Intel(R) Xeon(R) CPU E5-2680 v4 @ 2.40GHz
WORD_SIZE: 64
LIBM: libopenlibm
LLVM: libLLVM-6.0.1 (ORCJIT, haswell)

Package Information:

```
Status: `~/home/crackauckas/.julia/environments/v1.1/Project.toml`  
[c52e3926-4ff0-5f6e-af25-54175e0327b1] Atom 0.8.7  
[bcd4f6db-9728-5f36-b5f7-82caef46ccdb] DelayDiffEq 5.3.0  
[bb2cbb15-79fc-5d1e-9bf1-8ae49c7c1650] DiffEqBenchmarks 0.1.0  
[459566f4-90b8-5000-8ac3-15dfb0a30def] DiffEqCallbacks 2.5.2  
[f3b72e0c-5b89-59e1-b016-84e28bfd966d] DiffEqDevTools 2.8.0  
[aae7a2af-3d4f-5e19-a356-7da93b79d9d0] DiffEqFlux 0.5.0  
[78ddff82-25fc-5f2b-89aa-309469cbf16f] DiffEqMonteCarlo 0.14.0  
[77a26b50-5914-5dd7-bc55-306e6241c503] DiffEqNoiseProcess 3.3.1  
[9fdde737-9c7f-55bf-ade8-46b3f136cc48] DiffEqOperators 3.5.0  
[055956cb-9e8b-5191-98cc-73ae4a59e68a] DiffEqPhysics 3.1.0  
[a077e3f3-b75c-5d7f-a0c6-6bc4c8ec64a9] DiffEqProblemLibrary 4.1.0  
[41bf760c-e81c-5289-8e54-58b1f1f8abe2] DiffEqSensitivity 3.2.2  
[0c46a032-eb83-5123-abaf-570d42b7fbaa] DifferentialEquations 6.4.0  
[b305315f-e792-5b7a-8f41-49f472929428] Elliptic 0.5.0  
[587475ba-b771-5e3f-ad9e-33799f191a9c] Flux 0.8.3  
[e5e0dc1b-0480-54bc-9374-aad01c23163d] Juno 0.7.0  
[7f56f5a3-f504-529b-bc02-0b1fe5e64312] LSODA 0.4.0  
[c030b06c-0b6d-57c2-b091-7029874bd033] ODE 2.4.0  
[54ca160b-1b9f-5127-a996-1867f4bc2a2c] ODEInterface 0.4.5  
[09606e27-ecf5-54fc-bb29-004bd9f985bf] ODEInterfaceDiffEq 3.3.0  
[1dea7af3-3e70-54e6-95c3-0bf5283fa5ed] OrdinaryDiffEq 5.8.1  
[2dcacdae-9679-587a-88bb-8b444fb7085b] ParallelDataTransfer 0.5.0  
[65888b18-ceab-5e60-b2b9-181511a3b968] ParameterizedFunctions 4.1.1  
[91a5bcd-d55d7-5caf-9e0b-520d859cae80] Plots 0.25.1  
[d330b81b-6aea-500a-939a-2ce795aea3ee] PyPlot 2.8.1  
[731186ca-8d62-57ce-b412-fbd966d074cd] RecursiveArrayTools 0.20.0  
[295af30f-e4ad-537b-8983-00126c2a3abe] Revise 2.1.6  
[90137ffa-7385-5640-81b9-e52037218182] StaticArrays 0.11.0  
[789caeaf-c7a9-5a7d-9973-96adeb23e2a0] StochasticDiffEq 6.2.0  
[c3572dad-4567-51f8-b174-8c6c989267f4] Sundials 3.6.0  
[92b13dbe-c966-51a2-8445-caca9f8a7d42] TaylorIntegration 0.5.0  
[44d3d7a6-8a23-5bf8-98c5-b353f8df5ec9] Weave 0.9.0  
[e88e6eb3-aa80-5325-afca-941959d7151f] Zygote 0.3.1
```