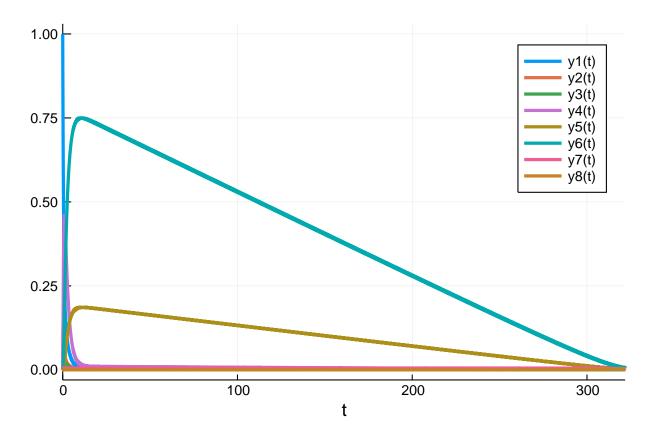
HIRES Work-Precision Diagrams

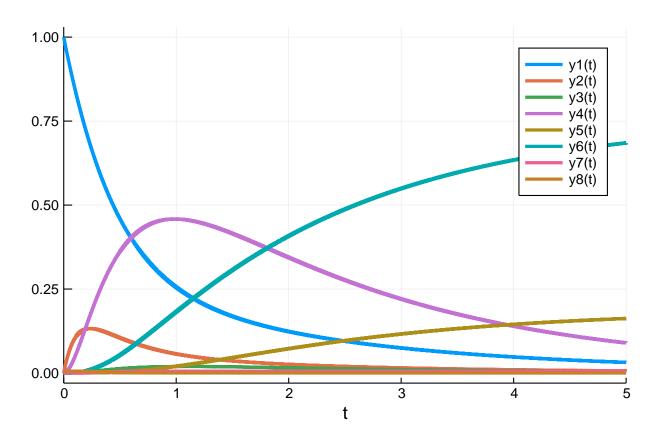
Chris Rackauckas

June 12, 2019

```
using OrdinaryDiffEq, ParameterizedFunctions, Plots, ODE, ODEInterfaceDiffEq, LSODA,
   DiffEqDevTools, Sundials
gr() #gr(fmt=:png)
f = @ode_def Hires begin
  dy1 = -1.71*y1 + 0.43*y2 + 8.32*y3 + 0.0007
  dy2 = 1.71*y1 - 8.75*y2
  dy3 = -10.03*y3 + 0.43*y4 + 0.035*y5
  dy4 = 8.32*y2 + 1.71*y3 - 1.12*y4
  dy5 = -1.745*y5 + 0.43*y6 + 0.43*y7
  dy6 = -280.0*y6*y8 + 0.69*y4 + 1.71*y5 -
           0.43*y6 + 0.69*y7
 dy7 = 280.0*y6*y8 - 1.81*y7
  dy8 = -280.0*y6*y8 + 1.81*y7
u0 = zeros(8)
u0[1] = 1
u0[8] = 0.0057
prob = ODEProblem(f,u0,(0.0,321.8122))
sol = solve(prob,Rodas5(),abstol=1/10^14,reltol=1/10^14)
test_sol = TestSolution(sol)
abstols = 1.0 ./ 10.0 .^{(4:11)}
reltols = 1.0 ./ 10.0 .^ (1:8);
plot(sol)
```



plot(sol,tspan=(0.0,5.0))



0.1 Omissions

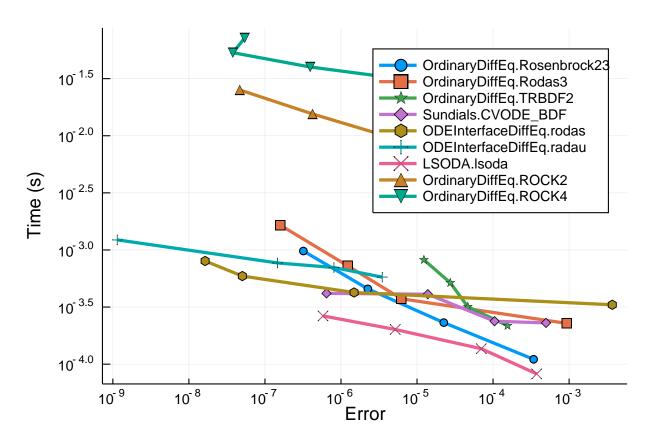
The following were omitted from the tests due to convergence failures. ODE.jl's adaptivity is not able to stabilize its algorithms, while GeometricIntegratorsDiffEq has not upgraded to Julia 1.0. GeometricIntegrators.jl's methods used to be either fail to converge at comparable dts (or on some computers errors due to type conversions).

```
 \#sol = solve(prob,ode23s()); \ println("Total ODE.jl \ steps: \ \$(length(sol))") \\ \#using \ GeometricIntegratorsDiffEq \\ \#try \\ \#sol = solve(prob,GIRadIIA3(),dt=1/10) \\ \#catch \ e \\ \# \ println(e) \\ \#end
```

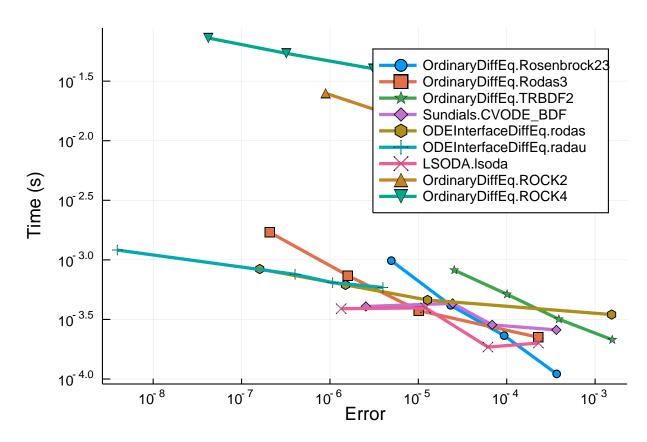
0.2 High Tolerances

This is the speed when you just want the answer.

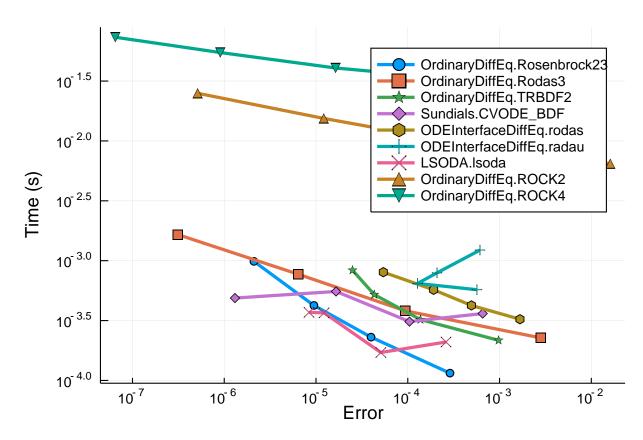
```
solve(prob, ddebdf())
solve(prob, rodas())
solve(prob, radau())
abstols = 1.0 ./ 10.0 .^ (5:8)
reltols = 1.0 ./ 10.0 .^ (1:4);
setups = [Dict(:alg=>Rosenbrock23()),
          Dict(:alg=>Rodas3()),
          Dict(:alg=>TRBDF2()),
          Dict(:alg=>CVODE_BDF()),
          Dict(:alg=>rodas()),
          Dict(:alg=>radau()),
          Dict(:alg=>lsoda()),
          Dict(:alg=>ROCK2()),
          Dict(:alg=>ROCK4())
wp = WorkPrecisionSet(prob,abstols,reltols,setups;
                      save_everystep=false,appxsol=test_sol,maxiters=Int(1e5),numruns=10)
plot(wp)
```

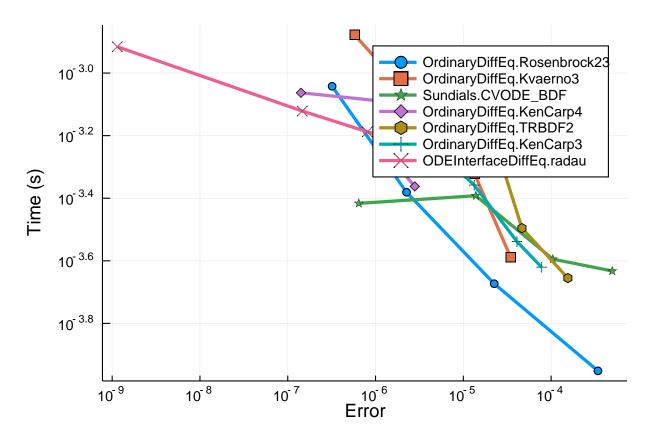


 $\label{eq:wp_recisionSet} $$ wp = WorkPrecisionSet(prob,abstols,reltols,setups;dense = false,verbose=false, appxsol=test_sol,maxiters=Int(1e5),error_estimate=:12) $$ plot(wp) $$$

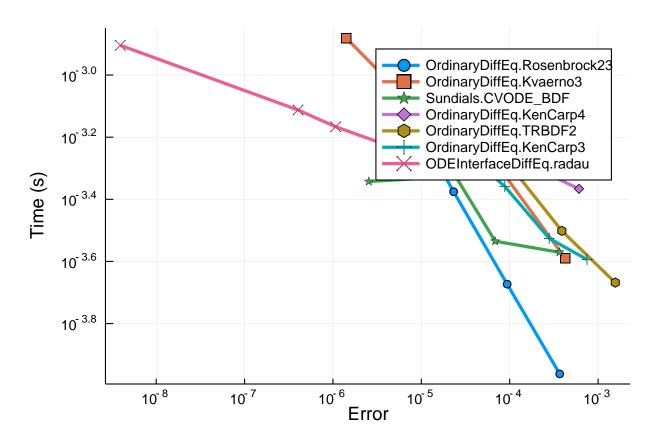


plot(wp)

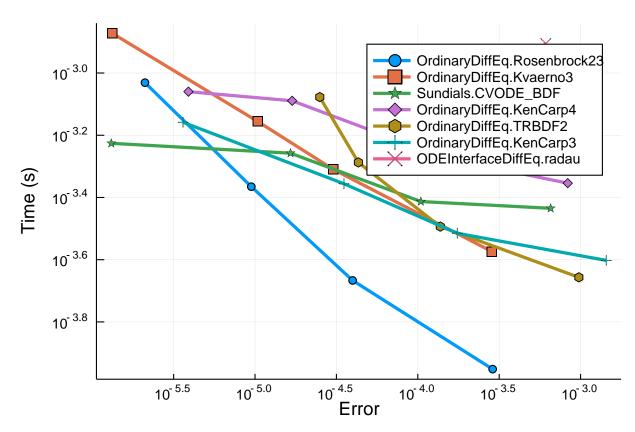


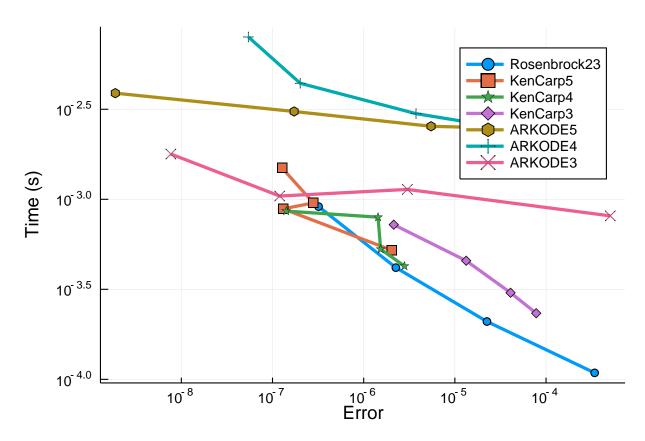


 $\label{eq:wp_recisionSet} $$ wp = WorkPrecisionSet(prob,abstols,reltols,setups;dense = false,verbose=false, appxsol=test_sol,maxiters=Int(1e5),error_estimate=:12) $$ plot(wp) $$$

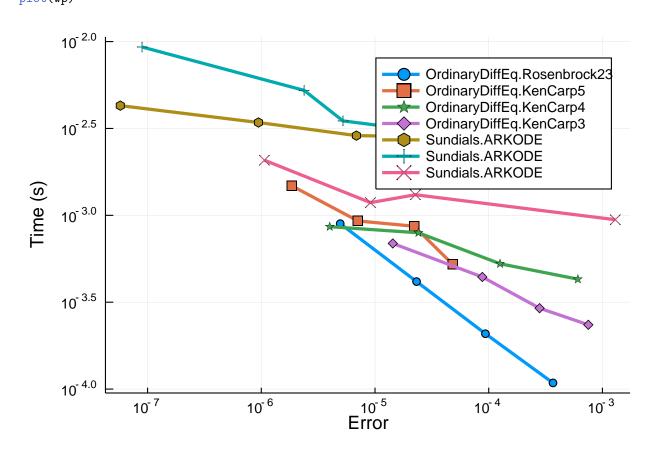


plot(wp)



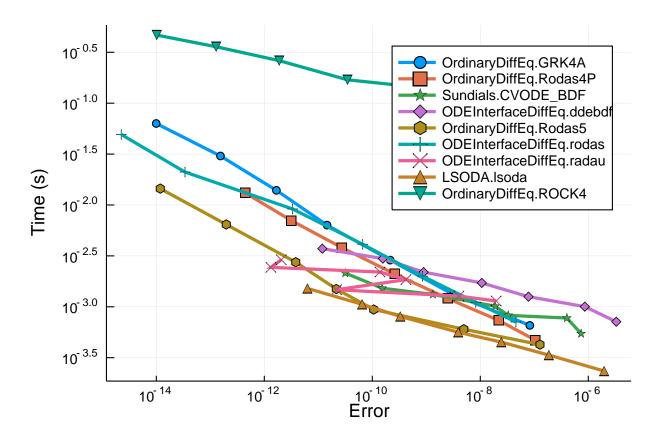


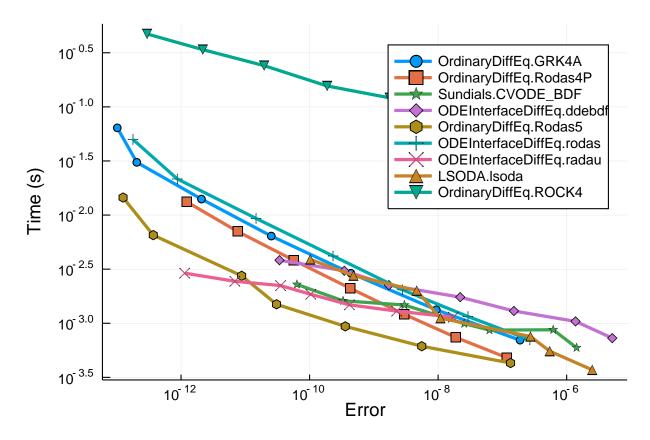
 $\label{eq:wp_recisionSet} $$ wp = WorkPrecisionSet(prob,abstols,reltols,setups;dense = false,verbose=false, appxsol=test_sol,maxiters=Int(1e5),error_estimate=:12) $$ plot(wp) $$$

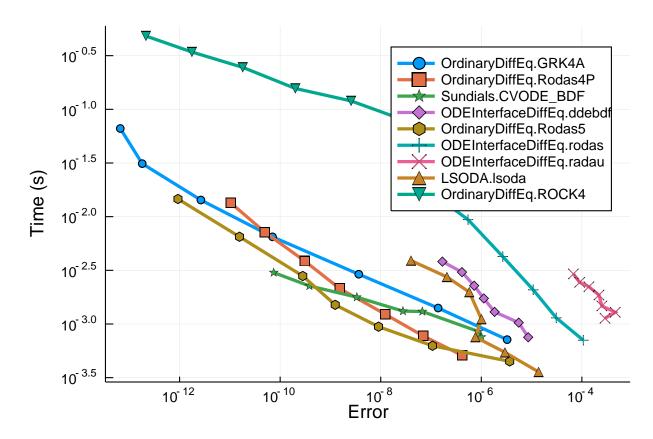


0.2.1 Low Tolerances

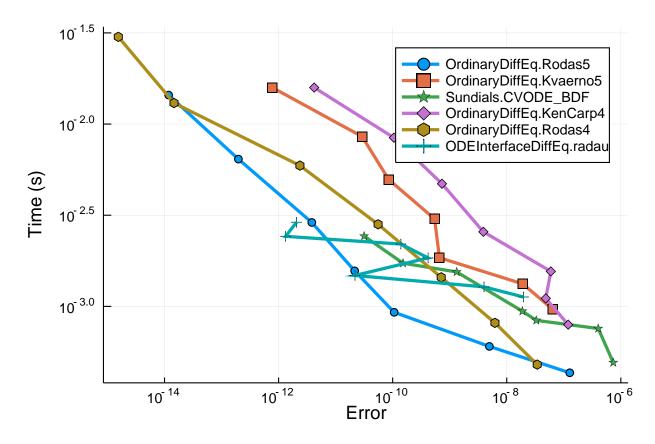
This is the speed at lower tolerances, measuring what's good when accuracy is needed.



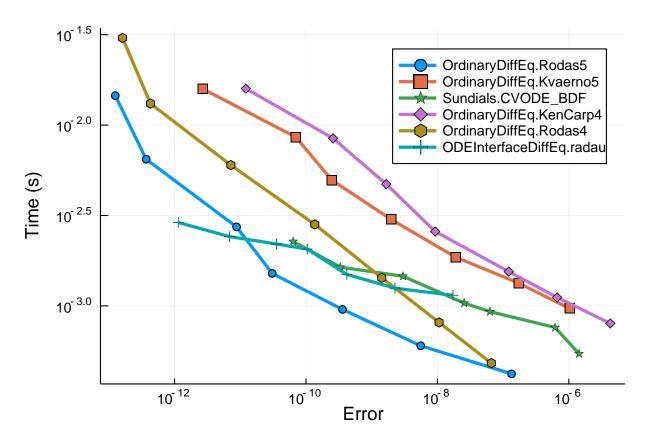


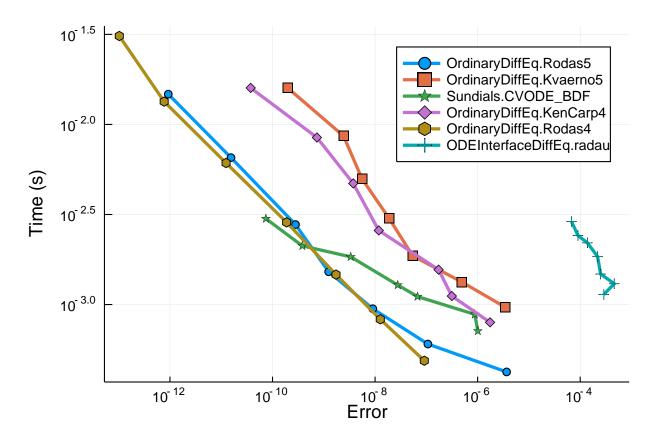


```
setups = [
          Dict(:alg=>Rodas5()),
```

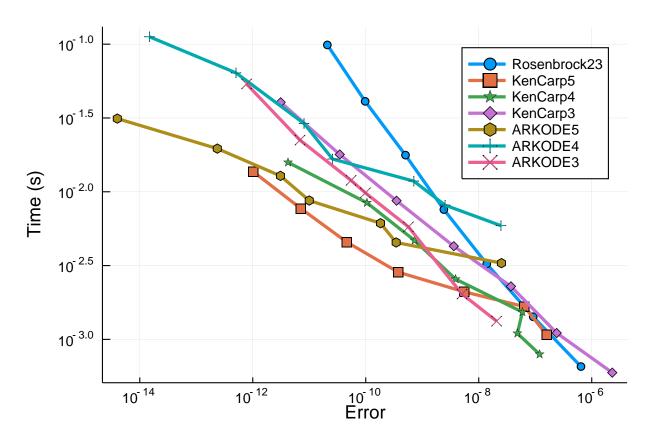


 $\label{eq:wp_recisionSet} $$ wp = WorkPrecisionSet(prob,abstols,reltols,setups;verbose=$false$, $$ dense=$false$,appxsol=test_sol,maxiters=Int(1e5),error_estimate=:12) $$ plot(wp) $$$

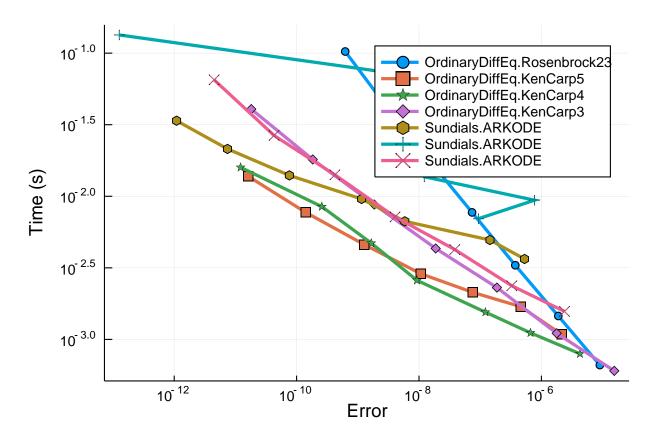




```
Dict(:alg=>KenCarp4()),
    Dict(:alg=>KenCarp3()),
    Dict(:alg=>ARKODE(order=5)),
    Dict(:alg=>ARKODE()),
    Dict(:alg=>ARKODE()),
    Dict(:alg=>ARKODE(order=3))]
names = ["Rosenbrock23" "KenCarp5" "KenCarp4" "KenCarp3" "ARKODE5" "ARKODE4" "ARKODE3"]
wp = WorkPrecisionSet(prob,abstols,reltols,setups;
    names=names,save_everystep=false,appxsol=test_sol,maxiters=Int(1e5))
plot(wp)
```



 $\label{eq:wp} $$ $ wp = WorkPrecisionSet(prob,abstols,reltols,setups;verbose= false,\\ dense= false,appxsol=test_sol,maxiters=Int(1e5),error_estimate=:12) $$ plot(wp) $$$



The following algorithms were removed since they failed.

0.2.2 Conclusion

At high tolerances, Rosenbrock23 and lsoda hitsthe the error estimates and are fast. At lower tolerances and normal user tolerances, Rodas5 is extremely fast. When you get down to reltol=1e-10 radau begins to become as efficient as Rodas4, and it continues to do well below that.

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

0.3 Appendix

These benchmarks are a part of the DiffEqBenchmarks.jl repository, found at: https://github.com/JuliaDenter.gl/github.com/g

```
using DiffEqBenchmarks
DiffEqBenchmarks.weave_file("StiffODE","Hires.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
[Oc46aO32-eb83-5123-abaf-57Od42b7fbaa] 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
[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
[91a5bcdd-55d7-5caf-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