

ROBER Work-Precision Diagrams

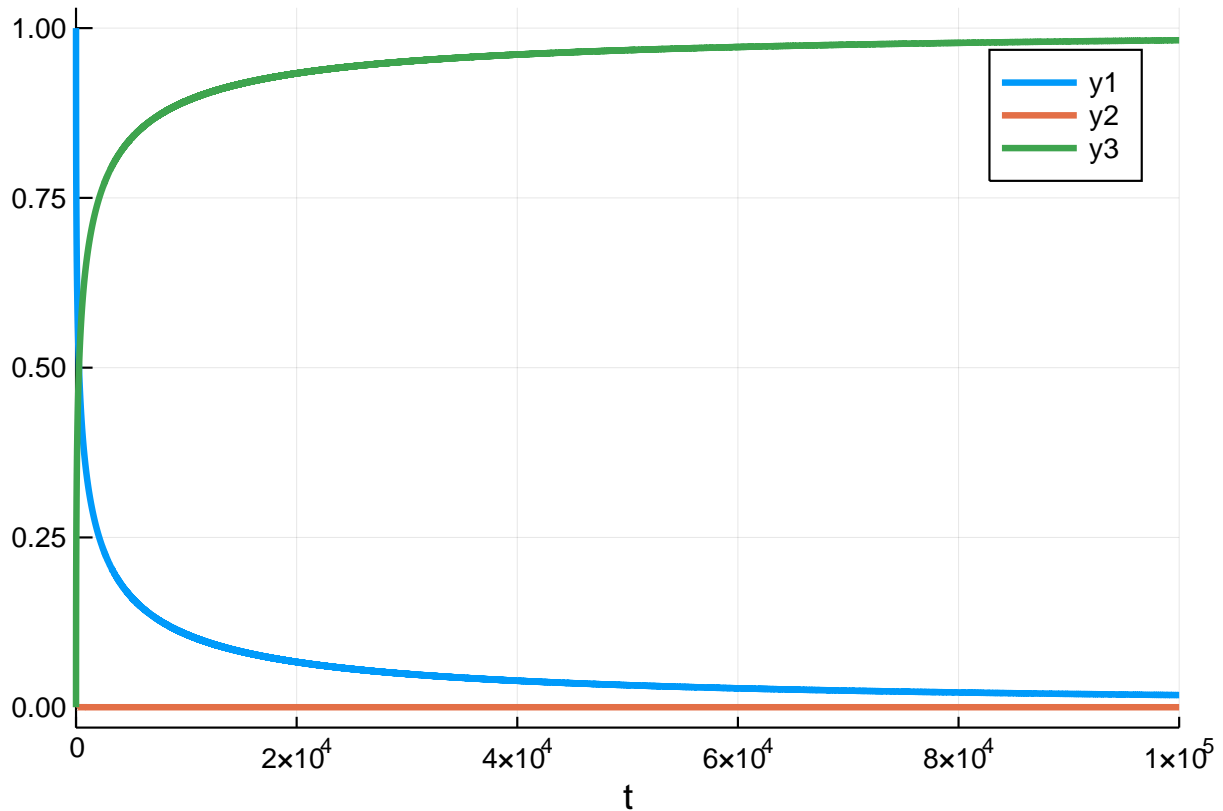
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March 10, 2019

```
using OrdinaryDiffEq, DiffEqDevTools, Sundials, ParameterizedFunctions, Plots, ODE,
    ODEInterfaceDiffEq, LSODA
gr()
rober = @ode_def begin
    dy_1 = -k_1*y_1+k_3*y_2*y_3
    dy_2 = k_1*y_1-k_2*y_2^2-k_3*y_2*y_3
    dy_3 = k_2*y_2^2
end k_1 k_2 k_3
prob = ODEProblem(rober,[1.0,0.0,0.0],(0.0,1e5),(0.04,3e7,1e4))
sol = solve(prob,CVODE_BDF(), abstol=1/10^14, reltol=1/10^14)
test_sol = TestSolution(sol)
abstols = 1.0 ./ 10.0 .^ (4:11)

8-element Array{Float64,1}:
 0.0001
 1.0e-5
 1.0e-6
 1.0e-7
 1.0e-8
 1.0e-9
 1.0e-10
 1.0e-11

plot(sol,labels=["y1","y2","y3"])
```



0.1 Omissions And Tweaking

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
```

ARKODE needs a lower `nonlinear_convergence_coefficient` in order to not diverge.

```
#sol = solve(prob,ARKODE(nonlinear_convergence_coefficient =
    1e-6), abstol=1e-5, reltol=1e-1); # Noisy, output omitted

sol = solve(prob,ARKODE(nonlinear_convergence_coefficient =
    1e-7), abstol=1e-5, reltol=1e-1);
```

Note that `1e-7` matches the value from the Sundials manual which was required for their example to converge on this problem. The default is `1e-1`.

```
#sol = solve(prob,ARKODE(order=3), abstol=1e-4, reltol=1e-1); # Fails to diverge but
    doesn't finish

#sol = solve(prob,ARKODE(order=5), abstol=1e-4, reltol=1e-1); # Noisy, output omitted
```

```
#sol = solve(prob,ARKODE(order=5,nonlinear_convergence_coefficient =
    1e-9), abstol=1e-5, reltol=1e-1); # Noisy, output omitted
```

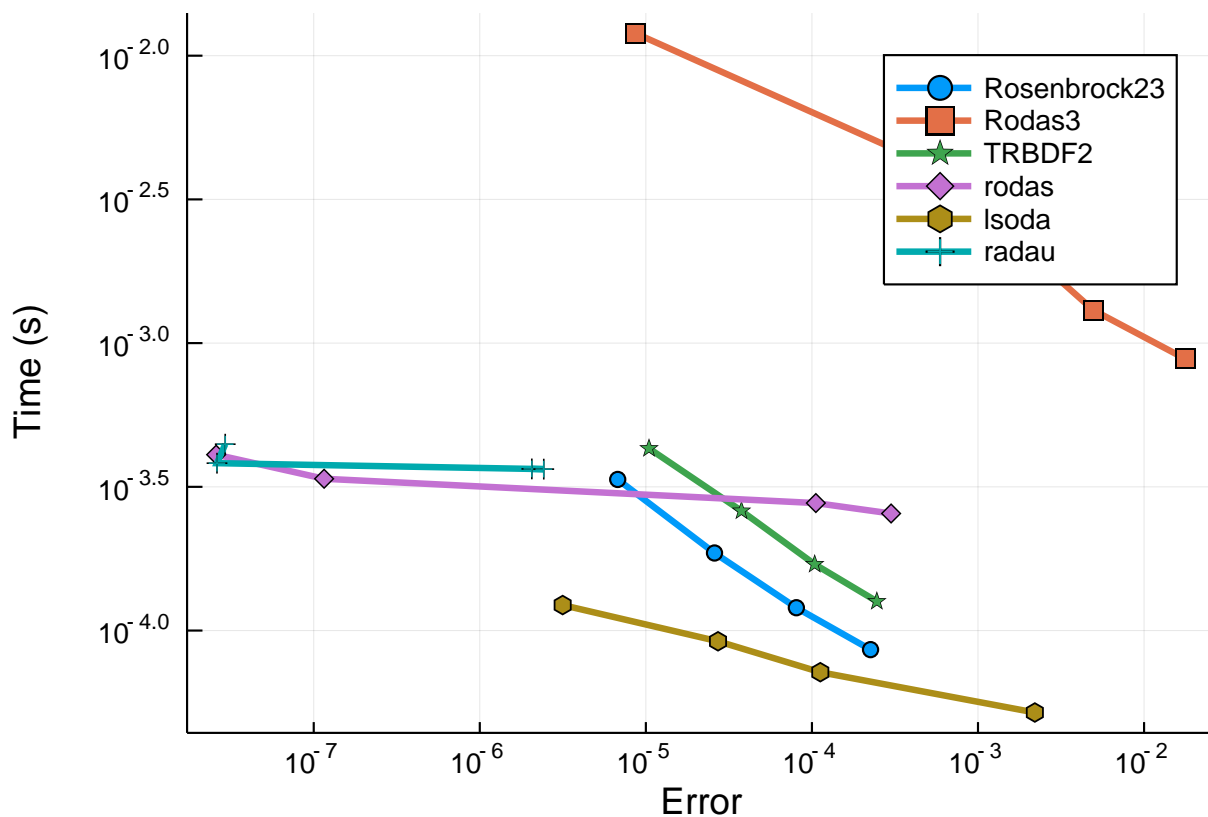
0.2 High Tolerances

This is the speed when you just want the answer. ode23s from ODE.jl was removed since it fails. Note that at high tolerances Sundials' CVODE_BDF fails as well so it's excluded from this test.

```
solve(prob, ddebdm())
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=>rodas()),
    Dict(:alg=>lsoda()),
    Dict(:alg=>radau())]

gr()
names = ["Rosenbrock23" "Rodas3" "TRBDF2" "rodas" "lsoda" "radau"]
wp = WorkPrecisionSet(prob,abstols,reltols,setups;names=names,
    save_everystep=false, appxsol=test_sol,maxiters=Int(1e5),numruns=10)

plot(wp)
```

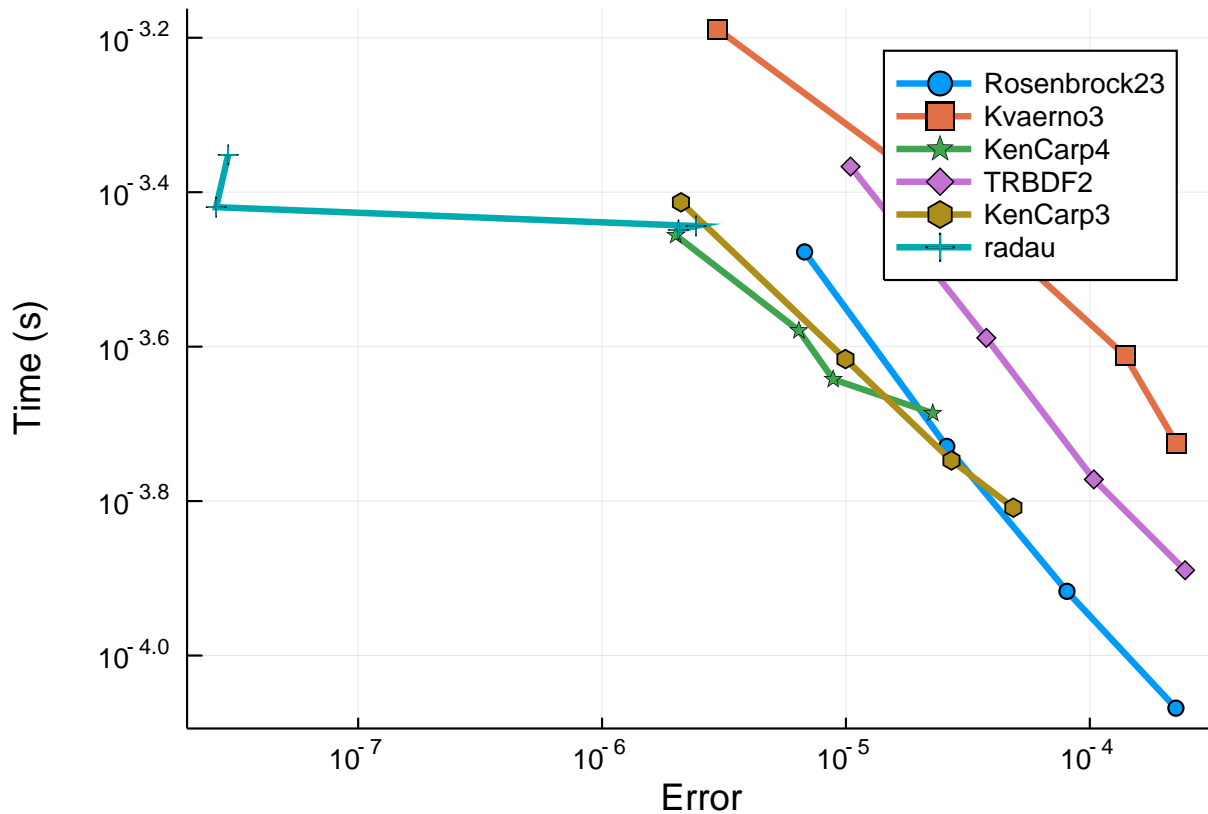


```
setups = [Dict(:alg=>Rosenbrock23()),
    Dict(:alg=>Kvaerno3()),
    Dict(:alg=>KenCarp4()),
    Dict(:alg=>TRBDF2()),
```

```

Dict(:alg=>KenCarp3()),
# Dict(:alg=>SDIRK2()), # Removed because it's bad
Dict(:alg=>radau())]
names = ["Rosenbrock23" "Kvaerno3" "KenCarp4" "TRBDF2" "KenCarp3" "radau"]
wp = WorkPrecisionSet(prob, abstols, reltols, setups; names=names,
    save_everystep=false, appxsol=test_sol, maxiters=Int(1e5), numruns=10)
plot(wp)

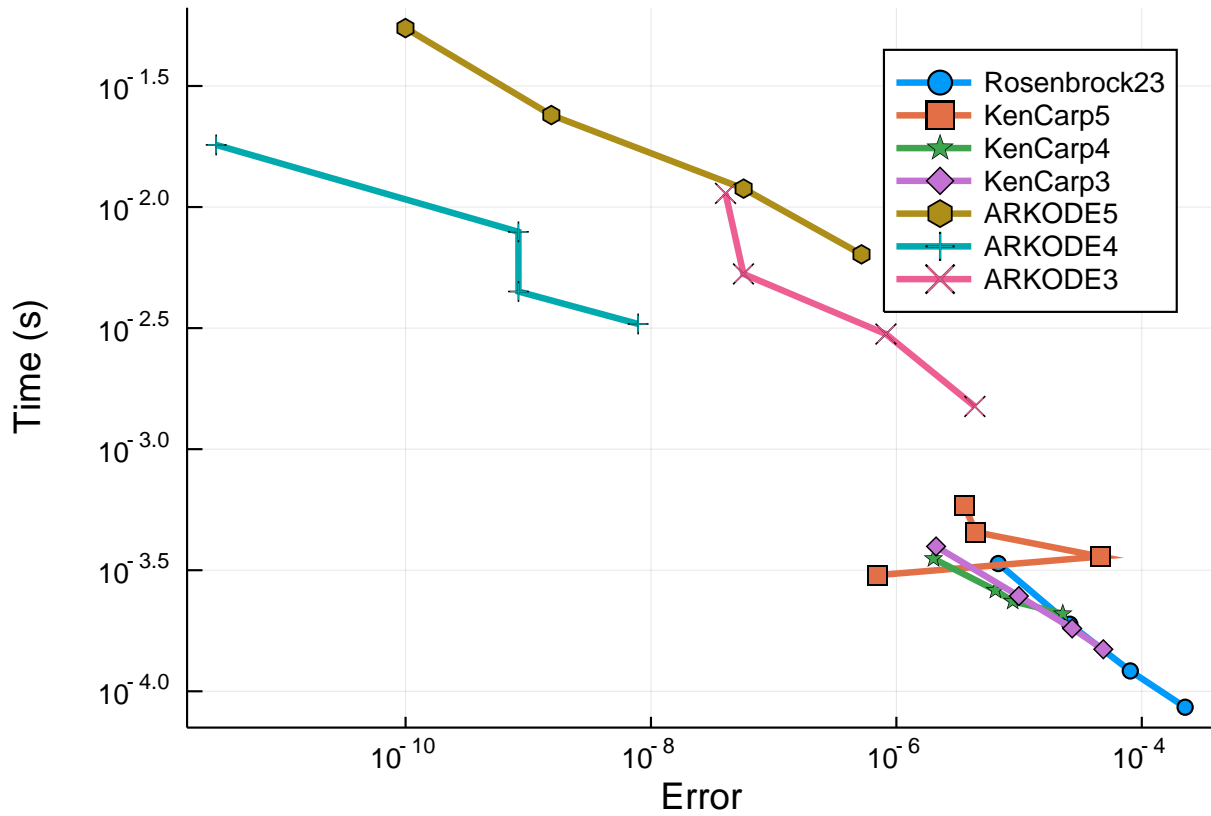
```



```

setups = [Dict(:alg=>Rosenbrock23()),
    Dict(:alg=>KenCarp5()),
    Dict(:alg=>KenCarp4()),
    Dict(:alg=>KenCarp3()),
    Dict(:alg=>ARKODE(nonlinear_convergence_coefficient = 1e-9, order=5)),
    Dict(:alg=>ARKODE(nonlinear_convergence_coefficient = 1e-8)),
    Dict(:alg=>ARKODE(nonlinear_convergence_coefficient = 1e-7, 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), numruns=10)
plot(wp)

```

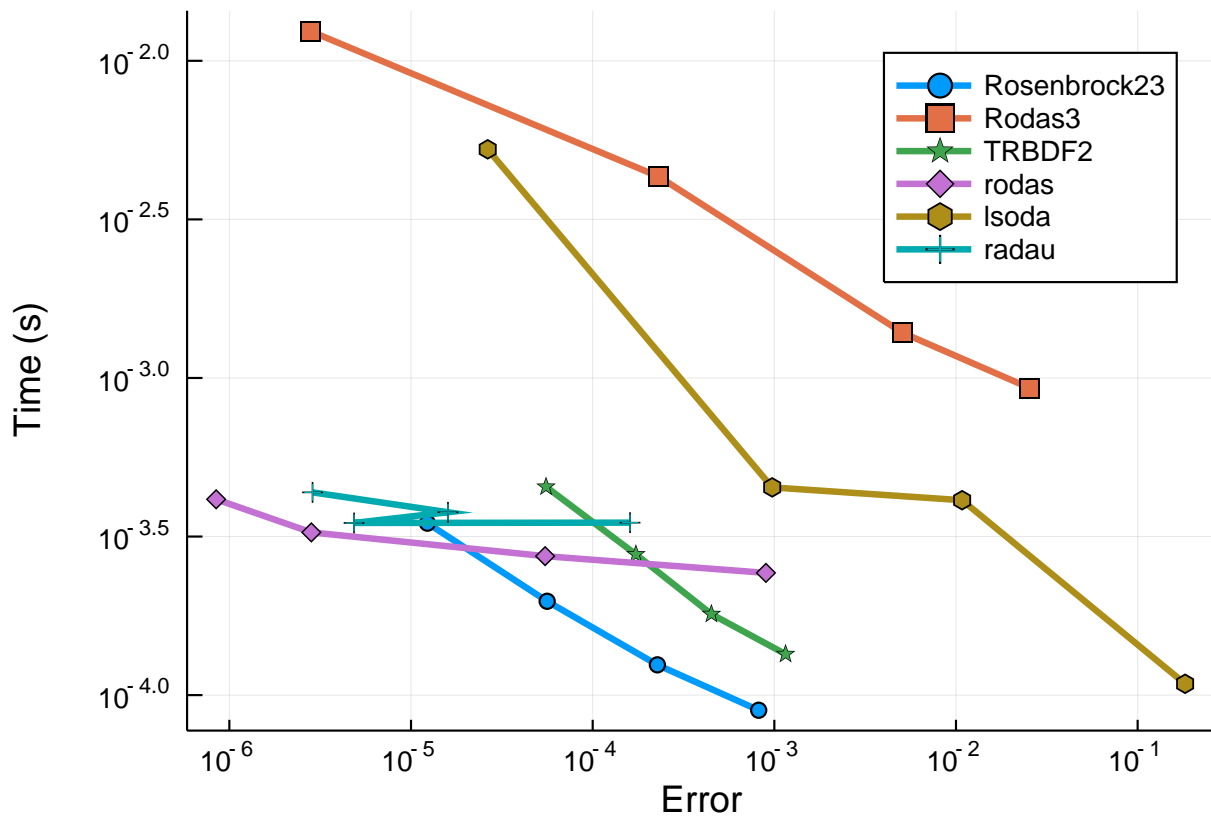


0.2.1 Timeseries Errors

```

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=>rodas()),
          Dict(:alg=>lsoda()),
          Dict(:alg=>radau())]
names = ["Rosenbrock23" "Rodas3" "TRBDF2" "rodas" "lsoda" "radau"]
wp = WorkPrecisionSet(prob,abstols,reltols,setups;names=names,
                      appxsol=test_sol,maxiters=Int(1e5),error_estimate=:l2,numruns=10)
plot(wp)

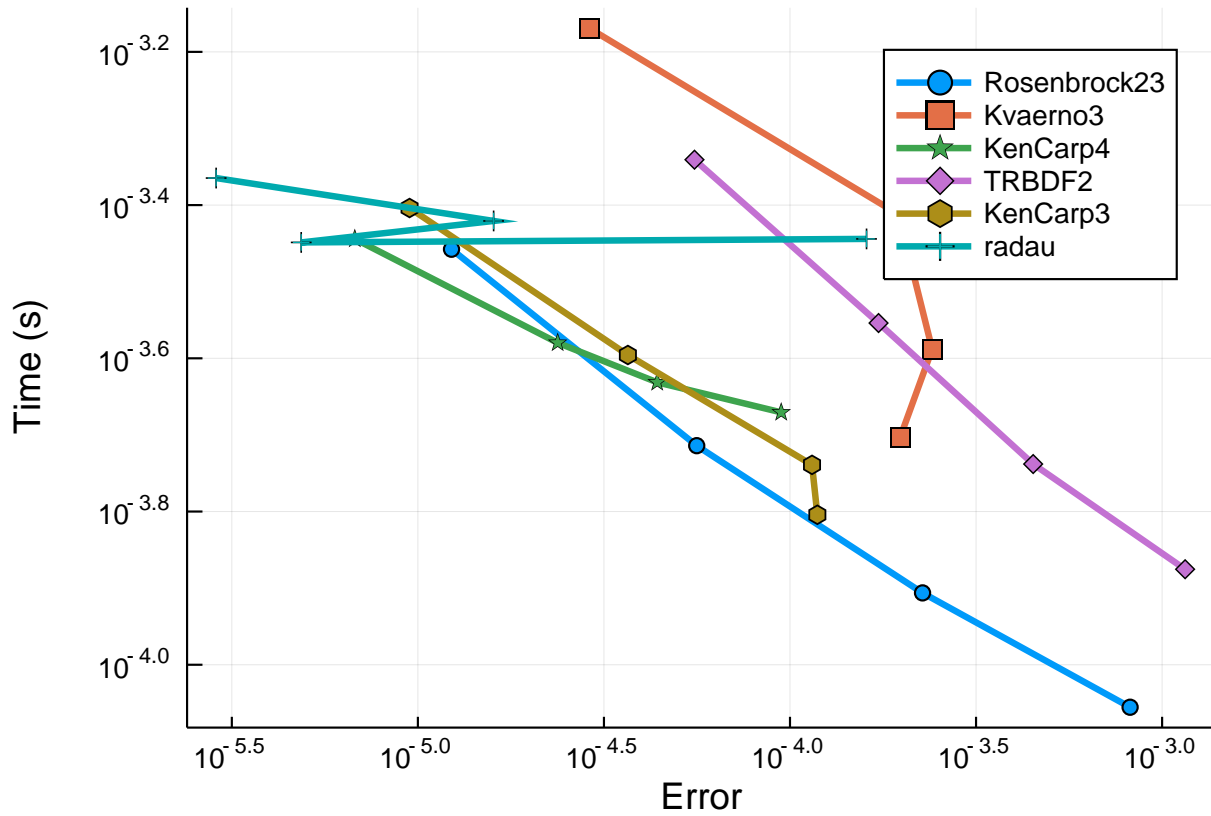
```



```

setups = [Dict(:alg=>Rosenbrock23()),
          Dict(:alg=>Kvaerno3()),
          Dict(:alg=>KenCarp4()),
          Dict(:alg=>TRBDF2()),
          Dict(:alg=>KenCarp3()),
          # Dict(:alg=>SDIRK2()), # Removed because it's bad
          Dict(:alg=>radau())]
names = ["Rosenbrock23" "Kvaerno3" "KenCarp4" "TRBDF2" "KenCarp3" "radau"]
wp = WorkPrecisionSet(prob, abstols, reltols, setups; names=names,
                     appxsol=test_sol, maxiters=Int(1e5), error_estimate=:l2, numruns=10)
plot(wp)

```



0.2.2 Low Tolerances

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

```
abstols = 1.0 ./ 10.0 .^ (7:12)
reltols = 1.0 ./ 10.0 .^ (4:9)
```

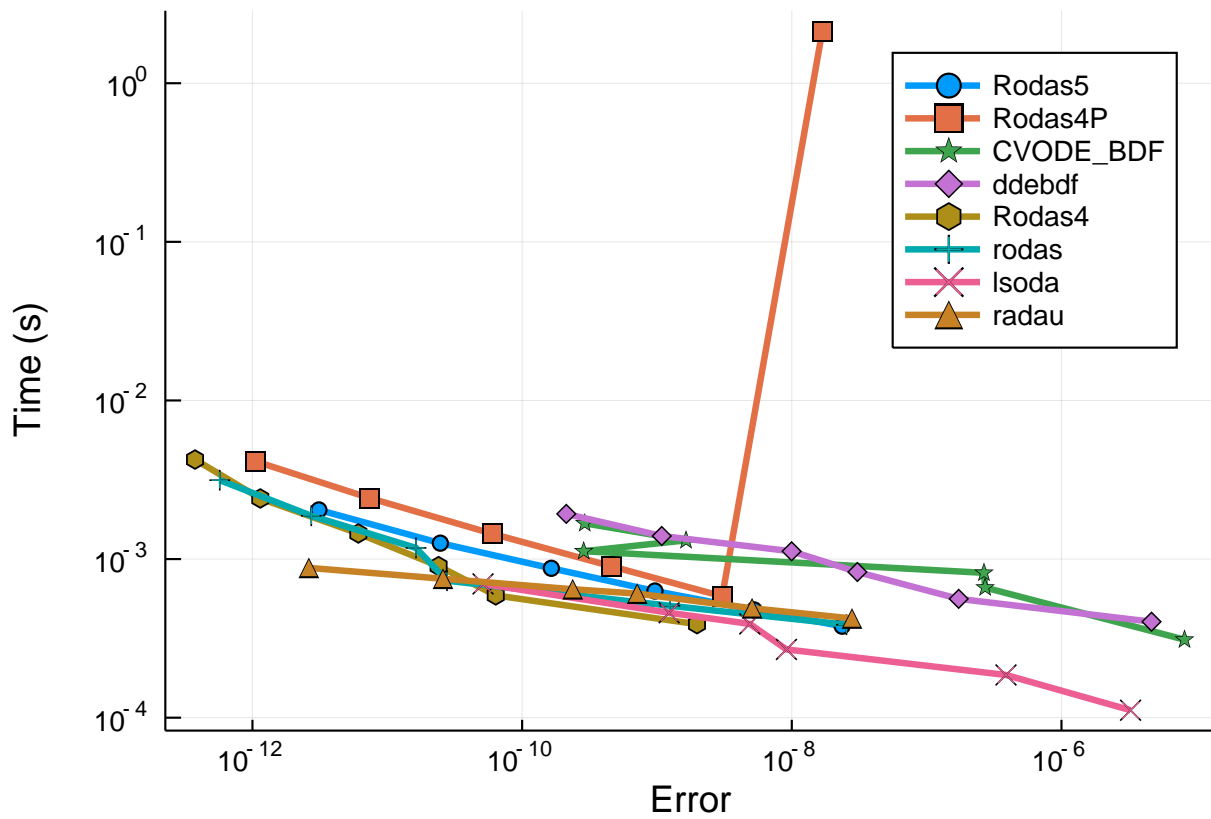
```
setups = [Dict(:alg=>Rodas5()),
           Dict(:alg=>Rodas4P()),
           Dict(:alg=>CVODE_BDF()),
           Dict(:alg=>ddebdf()),
           Dict(:alg=>Rodas4()),
           Dict(:alg=>rodas()),
           Dict(:alg=>lsoda()),
           Dict(:alg=>radau())
```

```
]
```

```
names = ["Rodas5" "Rodas4P" "CVODE_BDF" "ddebdf" "Rodas4" "rodas" "lsoda" "radau"]
```

```
wp = WorkPrecisionSet(prob,abstols,reltols,setups;names=names,
                      save_everystep=false,appxsol=test_sol,maxiters=Int(1e5),numruns=10)
```

```
plot(wp)
```



```

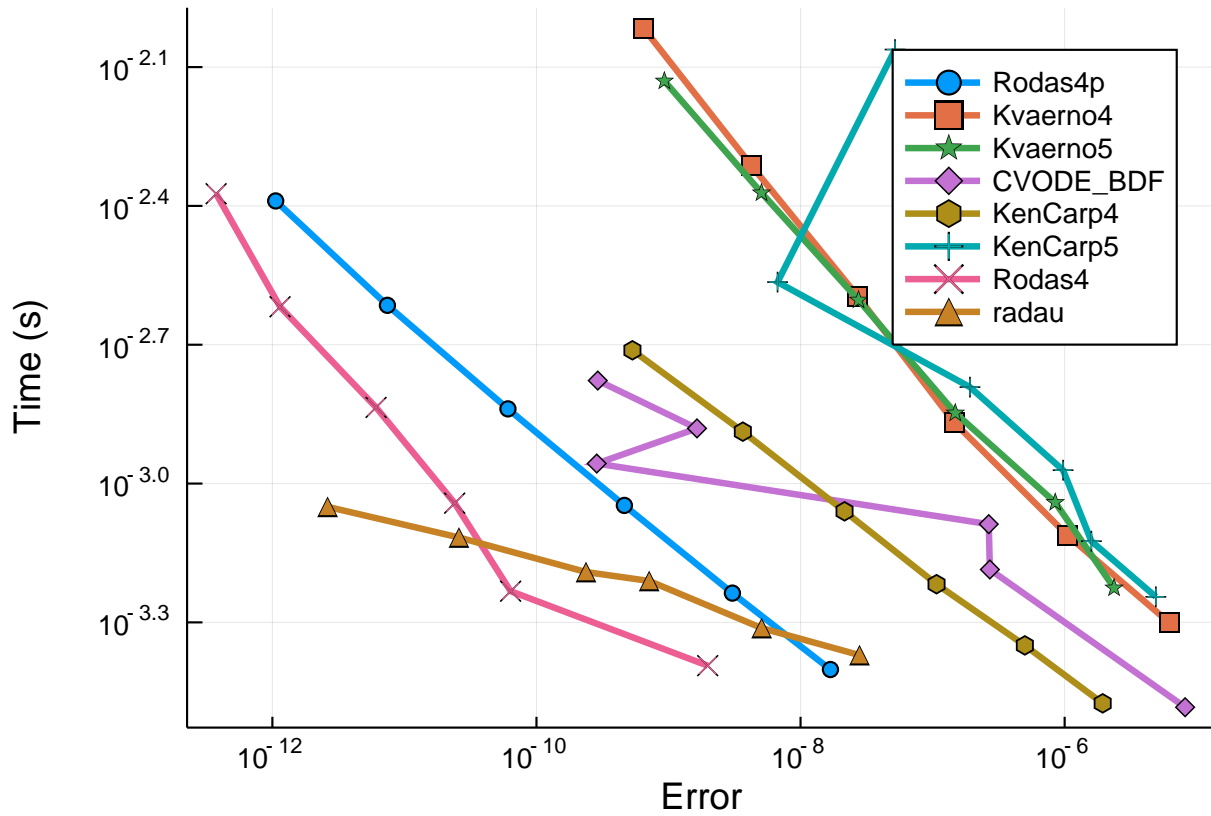
setups = [Dict(:alg=>Rodas4P()),
          Dict(:alg=>Kvaerno4()),
          Dict(:alg=>Kvaerno5()),
          Dict(:alg=>CVODE_BDF()),
          Dict(:alg=>KenCarp4()),
          Dict(:alg=>KenCarp5()),
          Dict(:alg=>Rodas4()),
          Dict(:alg=>radau())]

names = ["Rodas4p" "Kvaerno4" "Kvaerno5" "CVODE_BDF" "KenCarp4" "KenCarp5" "Rodas4"
         "radau"]

wp = WorkPrecisionSet(prob, abstols, reltols, setups; names=names,
                     save_everystep=false, appxsol=test_sol, maxiters=Int(1e5), numruns=10)

plot(wp)

```

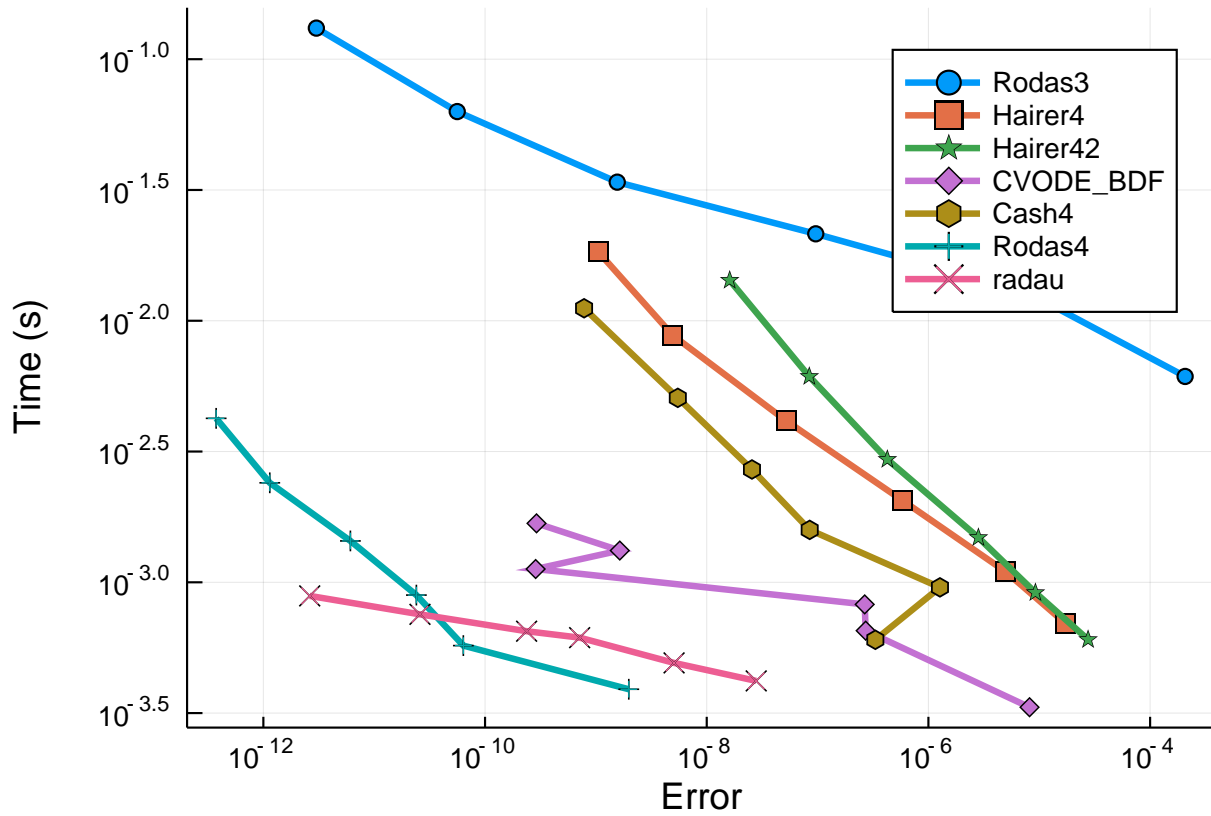
```

setups = [Dict(:alg=>Rodas3()),
           Dict(:alg=>Hairer4()),
           Dict(:alg=>Hairer42()),
           Dict(:alg=>CVODE_BDF()),
           Dict(:alg=>Cash4()),
           Dict(:alg=>Rodas4()),
           Dict(:alg=>radau())]

names = ["Rodas3" "Hairer4" "Hairer42" "CVODE_BDF" "Cash4" "Rodas4" "radau"]
wp = WorkPrecisionSet(prob, abstols, reltols, setups; names=names,
                      save_everystep=false, appxsol=test_sol, maxiters=Int(1e5), numruns=10)

plot(wp)

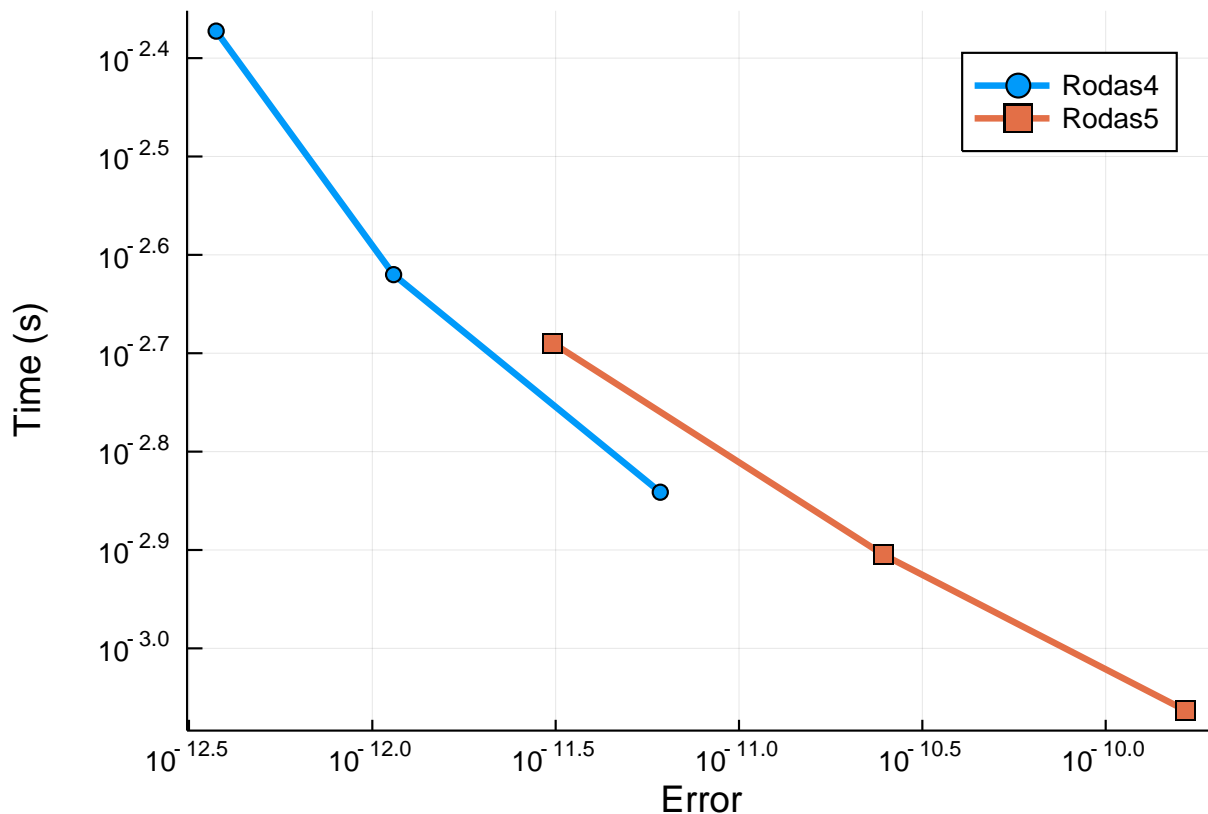
```



Rodas5 requires much lower tolerances to be stable here. Even then, it does not outdo Rodas4.

```
abstols = 1.0 ./ 10.0 .^ (10:12)
reltols = 1.0 ./ 10.0 .^ (7:9)

setups = [Dict(:alg=>Rodas4())
          Dict(:alg=>Rodas5())]
names = ["Rodas4" "Rodas5"]
wp = WorkPrecisionSet(prob,abstols,reltols,setups;names=names,
                      save_everystep=false,appxsol=test_sol,maxiters=Int(1e5),numruns=10)
plot(wp)
```



0.2.3 Conclusion

At high tolerances, `Rosenbrock23` and `lsoda` hit the the error estimates and are fast. At lower tolerances and normal user tolerances, `Rodas4` and `Rodas5` are extremely fast. `lsoda` does quite well across both ends. When you get down to `reltol=1e-9` `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/JuliaDiffEq/DiffEqBenchmarks.jl>

To locally run this tutorial, do the following commands:

```
using DiffEqBenchmarks
DiffEqBenchmarks.weave_file("StiffODE","ROBER.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/yingboma/.julia/dev/DiffEqBenchmarks/Project.toml`
[f3b72e0c-5b89-59e1-b016-84e28bfd966d] DiffEqDevTools 2.7.2
[7073ff75-c697-5162-941a-fcdaad2a7d2a] IJulia 1.17.0
[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 3.1.0
[1dea7af3-3e70-54e6-95c3-0bf5283fa5ed] OrdinaryDiffEq 5.3.0
[65888b18-ceab-5e60-b2b9-181511a3b968] ParameterizedFunctions 4.1.1
[91a5bcdd-55d7-5caf-9e0b-520d859cae80] Plots 0.23.1
[c3572dad-4567-51f8-b174-8c6c989267f4] Sundials 3.2.0
[44d3d7a6-8a23-5bf8-98c5-b353f8df5ec9] Weave 0.8.1
[b77e0a4c-d291-57a0-90e8-8db25a27a240] InteractiveUtils
[d6f4376e-aef5-505a-96c1-9c027394607a] Markdown
[44cfe95a-1eb2-52ea-b672-e2afdf69b78f] Pkg
[9a3f8284-a2c9-5f02-9a11-845980a1fd5c] Random