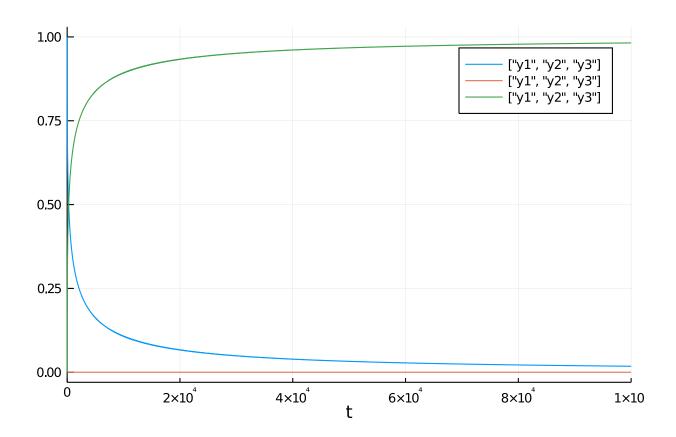
# ROBER Work-Precision Diagrams

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```
using OrdinaryDiffEq, DiffEqDevTools, Sundials, ParameterizedFunctions, Plots, ODE,
ODEInterfaceDiffEq, LSODA
gr()
using LinearAlgebra
LinearAlgebra.BLAS.set_num_threads(1)
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)}
reltols = 1.0 ./ 10.0 .^ (1:8);
8-element Array{Float64,1}:
 0.1
 0.01
 0.001
 0.0001
 1.0e-5
 1.0e-6
 1.0e-7
 1.0e-8
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).

ARKODE needs a lower nonlinear convergence coefficient in order to not diverge.

```
0.00015919299528398823
      0.00019473885760988623
      0.0002302847199357842
      0.00035022443063666944
      0.00038020935831189074
 62450.26310057348
 70536.63677961758
 78623.01045866167
 78957.02592024508
 79291.0413818285
 81054.01621605121
 83798.217146969
 93021.2875410419
 100000.0
u: 385-element Array{Array{Float64,1},1}:
 [1.0, 0.0, 0.0]
 [0.9999997270425205, 2.72952395412688e-7, 5.084119036489719e-12]
  \hbox{\tt [0.9999983223566348, 1.6764639202776004e-6, 1.1794449124939497e-9]} \\
  \hbox{\tt [0.9999969176727255, 3.07502694912627e-6, 7.300325444722892e-9]} \\
 [0.9999952749852724, 4.6988176517206366e-6, 2.619707607178777e-8]
 [0.9999936323005916, 6.303925964595367e-6, 6.377344407902863e-8]
 [0.9999922104763843, 7.673474247665867e-6, 1.160493681292835e-7]
 [0.9999907886544295, 9.02079987816384e-6, 1.905456924629883e-7]
 [0.9999859911271564, 1.3359747132384226e-5, 6.491257113639376e-7]
 [0.9999847917506273, 1.4385869142677054e-5, 8.223802300319677e-7]
 [0.026919791481076052, 1.1080274672128835e-7, 0.9730800977161898]
 [0.024252831120375448, 9.62216107464543e-8, 0.975747072658026]
 [0.022076376153276338, 5.623379155293192e-8, 0.9779235676129406]
 [0.02199498954140826, 9.00387535223304e-8, 0.9780049204198467]
 [0.02191424491866828, 9.23793668985613e-8, 0.9780856627019733]
 [0.021497918504541656, 9.204611150591026e-8, 0.9785019894493556]
 [0.020881079924596074, 8.519003071552104e-8, 0.9791188348853826]
 [0.019048726877150486, 8.042715971555324e-8, 0.9809511926956965]
 [0.017866016778411445, 7.285304497393234e-8, 0.9821339103685527]
```

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
```

Additionally, the ROCK methods do not perform well on this benchmark.

```
setups = [
     #Dict(:alg=>ROCK2()) #Unstable
     #Dict(:alg=>ROCK4()) #needs more iterations
]
```

O-element Array{Any,1}

Some of the bad Rosenbrocks fail:

```
setups = [
  #Dict(:alg=>Hairer4()),
  #Dict(:alg=>Hairer42()),
  #Dict(:alg=>Cash4()),
O-element Array{Any,1}
The EPIRK and exponential methods also fail:
sol = solve(prob, EXPRB53s3(), dt=2.0^(-8));
Error: InexactError: trunc(Int64, Inf)
sol = solve(prob, EPIRK4s3B(), dt=2.0^(-8));
Error: InexactError: trunc(Int64, Inf)
sol = solve(prob, EPIRK5P2(), dt=2.0^(-8));
Error: InexactError: trunc(Int64, Inf)
PDIRK44 also fails
sol = solve(prob, PDIRK44(), dt=2.0^{(-8)};
retcode: ConvergenceFailure
Interpolation: 3rd order Hermite
t: 1-element Array{Float64,1}:
u: 1-element Array{Array{Float64,1},1}:
 [1.0, 0.0, 0.0]
```

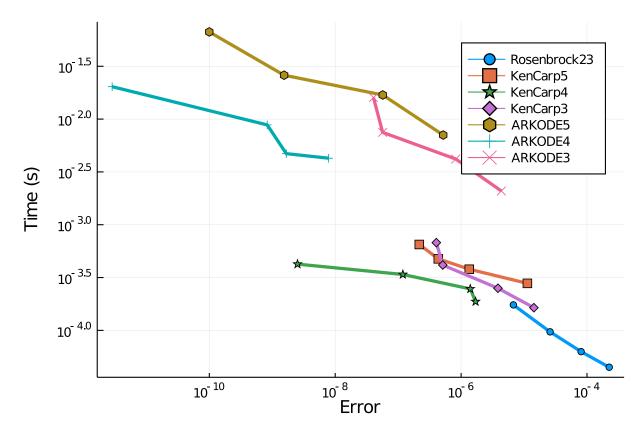
In fact, all non-adaptive methods fail on this problem.

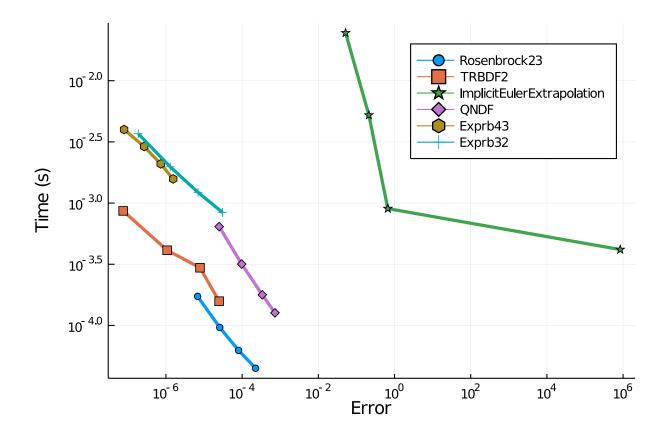
### 0.2 High Tolerances

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.

Error: Cannot find method(s) for rodas! I've tried to loadODESolvers(), but it didn't work. Please check ODEInterface.help\_solversupport() and call lo adODESolvers and check also this output. For further information see also ODEInterface.help\_install.

```
plot(wp)
Error: UndefVarError: wp not defined
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)
Error: Cannot find method(s) for radau! I've tried to loadODESolvers(), but
 it didn't work. Please check ODEInterface.help_solversupport() and call lo
adODESolvers and check also this output. For further information see also {\tt O}
DEInterface.help_install.
plot(wp)
Error: UndefVarError: wp not defined
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)
```

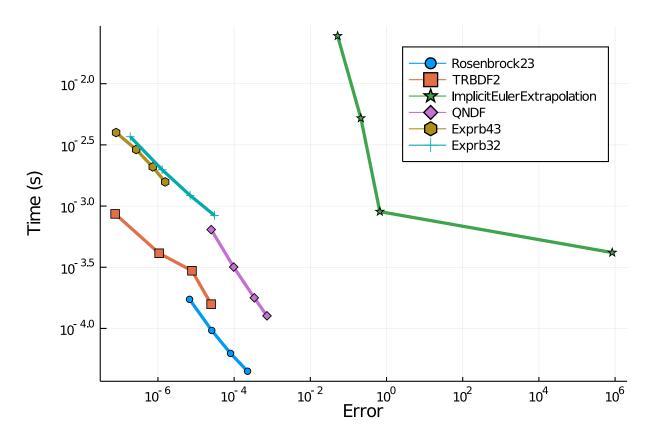




#### 0.2.1 Timeseries Errors

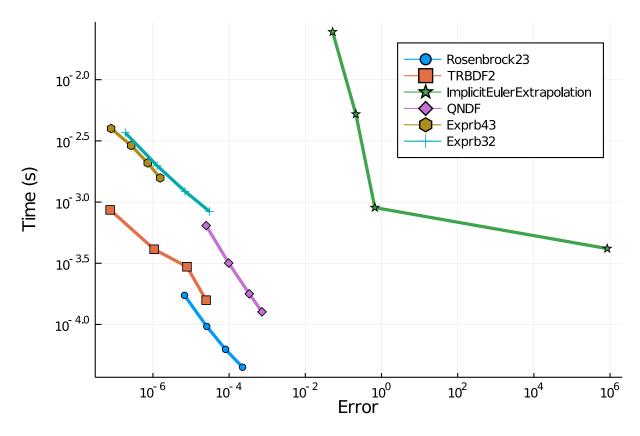
Error: Cannot find method(s) for rodas! I've tried to loadODESolvers(), but it didn't work. Please check ODEInterface.help\_solversupport() and call lo adODESolvers and check also this output. For further information see also O DEInterface.help\_install.

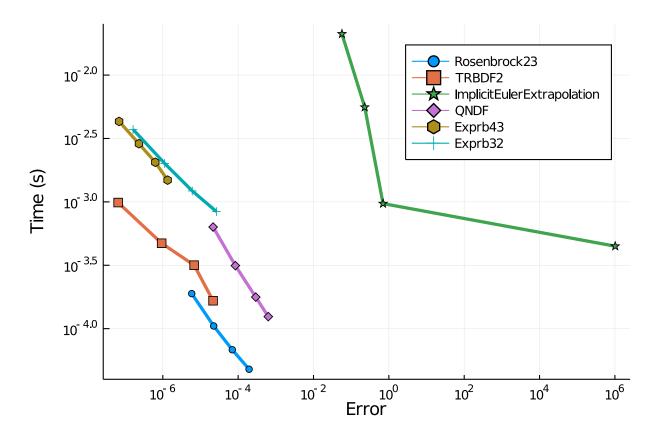
plot(wp)



Error: Cannot find method(s) for radau! I've tried to loadODESolvers(), but it didn't work. Please check ODEInterface.help\_solversupport() and call lo adODESolvers and check also this output. For further information see also O DEInterface.help\_install.

plot(wp)



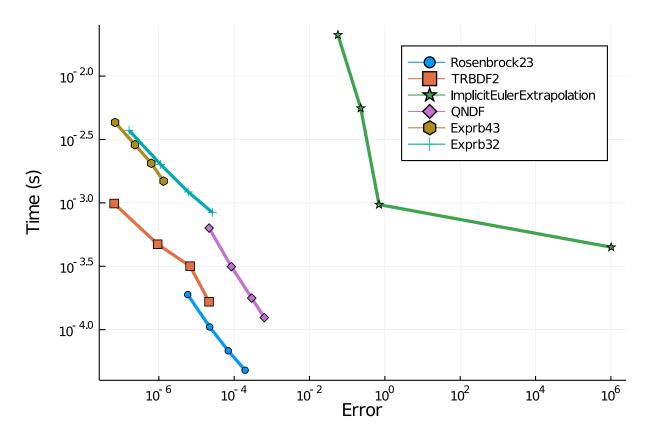


#### 0.2.2 Low Tolerances

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

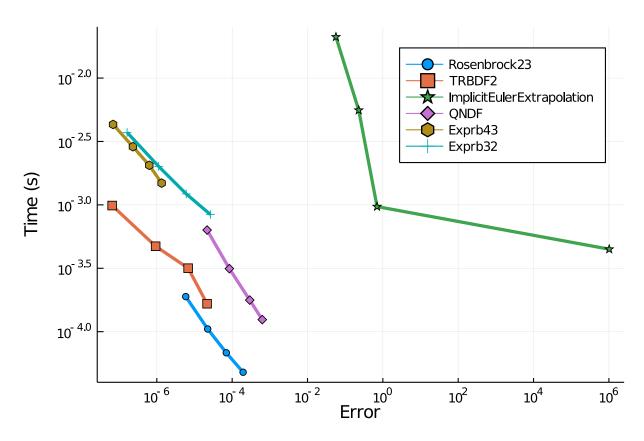
Error: Cannot find method(s) for ddebdf! I've tried to loadODESolvers(), bu t it didn't work. Please check ODEInterface.help\_solversupport() and call 1 oadODESolvers and check also this output. For further information see also ODEInterface.help\_install.

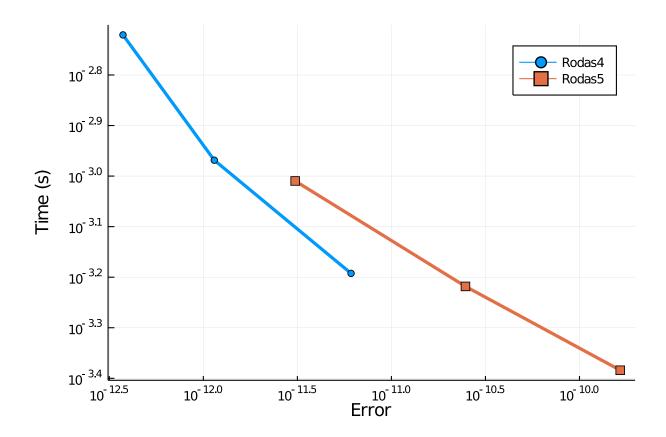
```
plot(wp)
```



Error: Cannot find method(s) for radau! I've tried to loadODESolvers(), but it didn't work. Please check ODEInterface.help\_solversupport() and call lo adODESolvers and check also this output. For further information see also ODEInterface.help\_install.

plot(wp)





#### 0.2.3 Conclusion

At high tolerances, Rosenbrock23 and lsoda hit 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/JuliaDenchmarks.jl repository,

```
using DiffEqBenchmarks
DiffEqBenchmarks.weave file("StiffODE","ROBER.jmd")
```

Computer Information:

```
Julia Version 1.4.2

Commit 44fa15b150* (2020-05-23 18:35 UTC)

Platform Info:

OS: Linux (x86_64-pc-linux-gnu)

CPU: Intel(R) Core(TM) i7-9700K CPU @ 3.60GHz
```

```
WORD_SIZE: 64
LIBM: libopenlibm
LLVM: libLLVM-8.0.1 (ORCJIT, skylake)
Environment:
JULIA_DEPOT_PATH = /builds/JuliaGPU/DiffEqBenchmarks.jl/.julia
JULIA_CUDA_MEMORY_LIMIT = 2147483648
JULIA_PROJECT = @.
JULIA_NUM_THREADS = 4
```

#### Package Information:

```
Status: `/builds/JuliaGPU/DiffEqBenchmarks.jl/benchmarks/StiffODE/Project.toml`
[eb300fae-53e8-50a0-950c-e21f52c2b7e0] DiffEqBiological 4.3.0
[f3b72e0c-5b89-59e1-b016-84e28bfd966d] DiffEqDevTools 2.22.0
[5a33fad7-5ce4-5983-9f5d-5f26ceab5c96] GeometricIntegratorsDiffEq 0.1.0
[7f56f5a3-f504-529b-bc02-0b1fe5e64312] LSODA 0.6.1
[c030b06c-0b6d-57c2-b091-7029874bd033] ODE 2.5.0
[09606e27-ecf5-54fc-bb29-004bd9f985bf] ODEInterfaceDiffEq 3.7.0
[1dea7af3-3e70-54e6-95c3-0bf5283fa5ed] OrdinaryDiffEq 5.41.0
[65888b18-ceab-5e60-b2b9-181511a3b968] ParameterizedFunctions 5.3.0
[91a5bcdd-55d7-5caf-9e0b-520d859cae80] Plots 1.5.3
[b4db0fb7-de2a-5028-82bf-5021f5cfa881] ReactionNetworkImporters 0.1.5
[c3572dad-4567-51f8-b174-8c6c989267f4] Sundials 4.2.5
[a759f4b9-e2f1-59dc-863e-4aeb61b1ea8f] TimerOutputs 0.5.6
[37e2e46d-f89d-539d-b4ee-838fcccc9c8e] LinearAlgebra
```