# Finding Maxima and Minima of DiffEq Solutions

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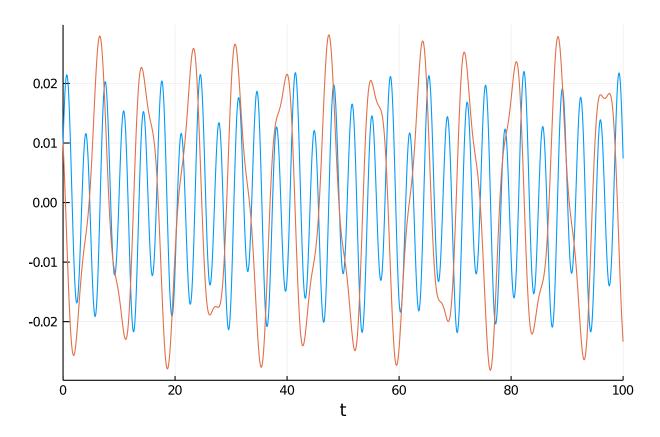
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### 0.0.1 Setup

In this tutorial we will show how to use Optim.jl to find the maxima and minima of solutions. Let's take a look at the double pendulum:

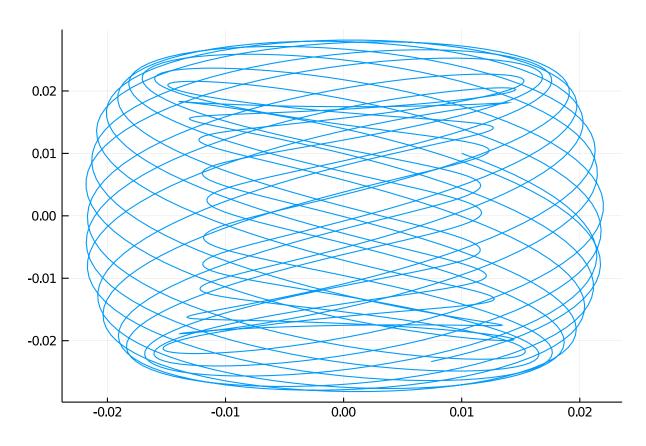
```
#Constants and setup
using OrdinaryDiffEq
initial = [0.01, 0.01, 0.01, 0.01]
tspan = (0.,100.)
#Define the problem
function double_pendulum_hamiltonian(udot,u,p,t)
    \alpha = u[1]
    1\alpha = u[2]
    \beta = u[3]
    1\beta = u[4]
    udot .=
    [2(1\alpha-(1+\cos(\beta))1\beta)/(3-\cos(2\beta)),
    -2\sin(\alpha) - \sin(\alpha+\beta),
    2(-(1+\cos(\beta))1\alpha + (3+2\cos(\beta))1\beta)/(3-\cos(2\beta)),
    -\sin(\alpha+\beta) - 2\sin(\beta)*(((1\alpha-1\beta)1\beta)/(3-\cos(2\beta))) + 2\sin(2\beta)*((1\alpha^2 - 2(1+\cos(\beta))1\alpha*1\beta))
+ (3+2\cos(\beta))1\beta^2/(3-\cos(2\beta))^2
end
#Pass to solvers
poincare = ODEProblem(double_pendulum_hamiltonian, initial, tspan)
ODEProblem with uType Array{Float64,1} and tType Float64. In-place: true
timespan: (0.0, 100.0)
u0: [0.01, 0.01, 0.01, 0.01]
sol = solve(poincare, Tsit5())
retcode: Success
Interpolation: specialized 4th order "free" interpolation
t: 193-element Array{Float64,1}:
   0.0
   0.08332584852065579
   0.24175300587841853
   0.4389533535703127
   0.6797301355043014
   0.9647629621490508
   1.3179425637594349
   1.7031226016307728
   2.0678503967116617
```

```
2.4717899847517866
 95.8457309586563
 96.3577910122243
 96.92913461915474
 97.44679415429573
 97.96248479179103
 98.51183391850897
 99.0608253308051
 99.58284388126884
100.0
u: 193-element Array{Array{Float64,1},1}:
 [0.01, 0.01, 0.01, 0.01]
 [0.00917068738040534, 0.0066690004553842845, 0.012420525490765832, 0.00826
6408515192912]
 [0.007673275265972518, 0.00037461737897660346, 0.016442590227730366, 0.004
636827483318282]
 [0.0061259744192393014, -0.007305450189721184, 0.01996737108423187, -0.000]
33649798308967233]
 [0.0049661106627111465, -0.01630851653373806, 0.021440659476204688, -0.00688]
705037098400459]
 [0.004795568331019467, -0.026238103489235838, 0.018824325208837592, -0.013
913364556753717]
  \begin{bmatrix} 0.006054679825355352, & -0.037124551879080515, & 0.010055702788069582, & -0.0218263636363 \end{bmatrix} 
0381274786473551
 [0.007900784412908595, -0.04667606960847389, -0.002673581831574413, -0.025
183036272033735]
 [0.008276510489473131, -0.05278433365633968, -0.012731546444725274, -0.025]
25804037623959]
 [0.005523496816741225, -0.055252504144926044, -0.01684388188262178, -0.021]
898963191274146]
 9082291418783]
 068009081627
 [0.004124711787696242, 0.056748788205059394, -0.005154187391921515, 0.0175]
9698310394298]
 [0.013079718118471311, 0.04807704307739497, -0.013770661225089886, 0.01828
6648610391705]
 [0.01531604024144831, 0.0316309595575519, -0.008956991644883512, 0.0171184]
0404984504]
 [0.011115490017374378, 0.009929018220630217, 0.0072974814212210725, 0.0103
53371812537737]
 [0.005713878919291268, -0.011787427051187304, 0.02050806401368939, -0.0023]
104589058526802]
  \hbox{\tt [0.0042114397261269225, -0.029911199361470082, 0.01875044642290467, -0.015] }
65071229490751]
 [0.005741239607321662, -0.04165385985159511, 0.007413270184092719, -0.0233]
48978525280305]
In time, the solution looks like:
using Plots; gr()
plot(sol, vars=[(0,3),(0,4)], leg=false, plotdensity=10000)
```



while it has the well-known phase-space plot:

plot(sol, vars=(3,4), leg=false)



#### 0.0.2 Local Optimization

Let's fine out what some of the local maxima and minima are. Optim.jl can be used to minimize functions, and the solution type has a continuous interpolation which can be used. Let's look for the local optima for the 4th variable around t=20. Thus our optimization function is:

```
f = (t) -> sol(t,idxs=4)
#1 (generic function with 1 method)
```

using Optim

first(t) is the same as t[1] which transforms the array of size 1 into a number. idxs=4 is the same as sol(first(t))[4] but does the calculation without a temporary array and thus is faster. To find a local minima, we can simply call Optim on this function. Let's find a local minimum:

```
Error: ArgumentError: Package Optim not found in current path:
- Run `import Pkg; Pkg.add("Optim")` to install the Optim package.
opt = optimize(f, 18.0, 22.0)
Error: UndefVarError: optimize not defined
From this printout we see that the minimum is at t=18.63 and the value is -2.79e-2. We
can get these in code-form via:
println(opt.minimizer)
Error: UndefVarError: opt not defined
println(opt.minimum)
Error: UndefVarError: opt not defined
To get the maximum, we just minimize the negative of the function:
f = (t) \rightarrow -sol(first(t), idxs=4)
opt2 = optimize(f, 0.0, 22.0)
Error: UndefVarError: optimize not defined
Let's add the maxima and minima to the plots:
plot(sol, vars=(0,4), plotdensity=10000)
scatter!([opt.minimizer],[opt.minimum],label="Local Min")
Error: UndefVarError: opt not defined
scatter!([opt2.minimizer],[-opt2.minimum],label="Local Max")
Error: UndefVarError: opt2 not defined
```

Brent's method will locally minimize over the full interval. If we instead want a local maxima nearest to a point, we can use BFGS(). In this case, we need to optimize a vector [t], and thus dereference it to a number using first(t).

```
f = (t) -> -sol(first(t),idxs=4)
opt = optimize(f,[20.0],BFGS())
```

Error: UndefVarError: BFGS not defined

## 0.0.3 Global Optimization

If we instead want to find global maxima and minima, we need to look somewhere else. For this there are many choices. A pure Julia option is BlackBoxOptim.jl, but I will use NLopt.jl. Following the NLopt.jl tutorial but replacing their function with out own:

```
import NLopt, ForwardDiff
Error: ArgumentError: Package NLopt not found in current path:
- Run `import Pkg; Pkg.add("NLopt")` to install the NLopt package.
count = 0 # keep track of # function evaluations
function g(t::Vector, grad::Vector)
  if length(grad) > 0
    #use ForwardDiff for the gradients
    grad[1] = ForwardDiff.derivative((t)->sol(first(t),idxs=4),t)
  sol(first(t),idxs=4)
opt = NLopt.Opt(:GN_ORIG_DIRECT_L, 1)
Error: UndefVarError: NLopt not defined
NLopt.lower_bounds!(opt, [0.0])
Error: UndefVarError: NLopt not defined
NLopt.upper_bounds!(opt, [40.0])
Error: UndefVarError: NLopt not defined
NLopt.xtol_rel!(opt,1e-8)
Error: UndefVarError: NLopt not defined
NLopt.min_objective!(opt, g)
Error: UndefVarError: NLopt not defined
(minf,minx,ret) = NLopt.optimize(opt,[20.0])
Error: UndefVarError: NLopt not defined
println(minf," ",minx," ",ret)
Error: UndefVarError: minf not defined
NLopt.max_objective!(opt, g)
Error: UndefVarError: NLopt not defined
(maxf,maxx,ret) = NLopt.optimize(opt,[20.0])
Error: UndefVarError: NLopt not defined
println(maxf," ",maxx," ",ret)
Error: UndefVarError: maxf not defined
```

```
plot(sol, vars=(0,4), plotdensity=10000)
scatter!([minx],[minf],label="Global Min")
Error: UndefVarError: minx not defined
scatter!([maxx],[maxf],label="Global Max")
Error: UndefVarError: maxx not defined
```

## 0.1 Appendix

This tutorial is part of the DiffEqTutorials.jl repository, found at: https://github.com/JuliaDiffEq/DiffEqTo locally run this tutorial, do the following commands:

```
using DiffEqTutorials
DiffEqTutorials.weave_file("ode_extras","03-ode_minmax.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/DiffEqTutorials.jl/.julia
 JULIA_CUDA_MEMORY_LIMIT = 536870912
  JULIA PROJECT = 0.
 JULIA_NUM_THREADS = 4
```

#### Package Information:

```
Status `/builds/JuliaGPU/DiffEqTutorials.jl/tutorials/ode_extras/Project.toml`
[961ee093-0014-501f-94e3-6117800e7a78] ModelingToolkit 3.11.0
[2774e3e8-f4cf-5e23-947b-6d7e65073b56] NLsolve 4.4.0
[1dea7af3-3e70-54e6-95c3-0bf5283fa5ed] OrdinaryDiffEq 5.41.0
[91a5bcdd-55d7-5caf-9e0b-520d859cae80] Plots 1.5.0
[37e2e46d-f89d-539d-b4ee-838fcccc9c8e] LinearAlgebra
[2f01184e-e22b-5df5-ae63-d93ebab69eaf] SparseArrays
```