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Parallel Homotopy Continuation in Julia

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1 Introduction

Homotopy Continuation is a numerical method for solving systems of polynomial equations. It is based on the idea of "deforming" a given system of equations into a simpler one, whose solutions are known, and then tracking the solutions of the original system as the deformation is undone.

In this project, the method will be implemented in the Julia programming language, making use of parallel computing in order to speed multiple root finding. The method is described in detail in [1], which was the primary source for this report.

2 Homotopy Continuation

We will only consider *square* systems of polynomial equations, i.e. systems of n polynomial equations in n variables, although or over- or under-determined systems can often be solved by reducing them to square systems, by respectively choosing a suitable square subsystem or adding equations. Morever, we will restrict ourselves to systems with isolated solutions, i.e. zero-dimensional varieties.

There are many ways to choose the "simpler" system, from now on called a *start system*, but in general we can observe that, by Bezout's theorem, a system $F = (f_1, \ldots, f_n)$ has at most $D := d_1 \ldots d_n$ solutions, where d_i is the degre of $f_i(x_1, \ldots, x_n)$. So, we could build a start system of the same size and whose polynomials have the same degrees, but whose solutions are easy to find, and thus can be used as starting points for the method.

For instance, the system $G = (g_1, \ldots, g_n)$, where

$$g_i(x_1,\ldots x_n) = x_i^{d_i} - 1,$$

is such a system, since it has exactly the D solutions

$$\left\{ \left(e^{\frac{k_1}{d_1}2\pi i}, \dots, e^{\frac{k_n}{d_n}2\pi i}\right), \text{ for } 0 \le k_i \le d_i - 1 \text{ and } i = 1, \dots, n \right\}.$$

2.1 Choosing the homotopy

The deformation between the original system and the start system is a *homotopy*, for instance the convex combination of F and G

$$H(x,t) = (1-t)F(x) + tG(x),$$
(1)

where $x := (x_1, \ldots, x_n)$ and $t \in [0, 1]$. This is such that the roots of H(x, 0) = G(x) are known, and the roots of H(x, 1) = F(x) are the solutions of the original system. Therefore, we can implicitly define a curve z(t) in \mathbb{C}^n by the equation

$$H(z(t),t) = 0, (2)$$

so that in order to approximate the roots of F it is enough to numerically track z(t).

To do so, we derive the expression (2) with respect to t, and get the Davidenko Differential Equation

$$\frac{\partial H}{\partial z}\frac{\mathrm{d}z}{\mathrm{d}t} + \frac{\partial H}{\partial t} = 0,$$

where $\frac{\partial H}{\partial z}$ is the Jacobian matrix of H with respect to z:

$$\frac{\partial H}{\partial z} = \begin{pmatrix} \frac{\partial H_1}{\partial z_1} & \cdots & \frac{\partial H_1}{\partial z_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial H_n}{\partial z_1} & \cdots & \frac{\partial H_n}{\partial z_n} \end{pmatrix}.$$

This can be rewritten as

$$\dot{z} = -\frac{\partial H}{\partial z}^{-1} \frac{\partial H}{\partial t}.$$
(3)

This is a system of n first-order differential equations, which can be solved numerically for z(t) as an initial value problem, and is called *path tracking*.

2.1.1 Gamma trick

While (1) is a fine choice of a homotopy, it's not what it's called a *good homotopy*: in order to ensure that the solution paths z(t) for different roots

- have no singularities, i.e. never cross each other for t > 0 (at t = 0, F could have singular solutions), and
- don't go to infinity for $t \to 0$ (F could have a solution at infinity),

we can employ the *Gamma trick*: this consists in modifying the linear homotopy (1) by subtituting the parameter $t \in [0, 1]$ with a complex curve q(t) connecting 0 and 1:

$$q(t) = \frac{\gamma t}{\gamma t + (1-t)},$$

where $\gamma \in (0, 1)$ is a random complex parameter. This "probability one" procedure, i.e. for any particular system choosing γ outside of a finite amount of lines through the origin ensures that we get a good homotopy, basically because of the finiteness of the branch locus of the homotopy. After substituting, we get

$$H(x,t) = \frac{(1-t)}{\gamma t + (1-t)}F(x) + \frac{\gamma t}{\gamma t + (1-t)}G(x),$$

and clearing denominators, here's our final homotopy:

$$H(x,t) = (1-t)F(x) + \gamma t G(x).$$
 (4)

2.2 Tracking down the roots

We now want to track down individual roots, following the solution paths from a root z_0 of the start system by solving the initial value problem associated to the Davidenko differential equation (3) with starting value z_0 and t ranging from 1 to 0.

This will be done numerically, using a first-order predictor-corrector tracking method, which consists in first using Euler's method to get an approximation \tilde{z}_i , and then using Newton's method to correct it using equation (2) so that it becomes a good approximation z_i of the next value of the solution path.

2.2.1 Predictor: Euler's method

Recall that Euler's method consists in approximating the solution of the initial value problem associated to a first-order ordinary differential equations

$$\begin{cases} \dot{z} = f(z, t) \\ z(t_0) = z_0 \end{cases}$$

by the sequence of points $(z_i)_{i\in\mathbb{N}}$ defined by the recurrence relation

$$z_{i+1} = z_i + h \cdot f(z_i, t_i),$$

where h is the step size. In our case, we have

$$f(z,t) = -\left(\frac{\partial H}{\partial z}(z,t)\right)^{-1} \frac{\partial H}{\partial t}(z,t)$$

and $t_0 = 1$, since we track from 1 to 0. For the same reason, we set

$$t_{i+1} = t_i - h.$$

2.2.2 Corrector: Newton's method

Since we want to solve

$$H(z,t) = 0,$$

we can use Newton's method to improve the approximation \tilde{z}_i obtained by Euler's method to a solution of such equation. This is done by moving towards the root of the tangent line of H at the current approximation, or in other words through the iteration

$$z_{i+1} = z_i - \left(\frac{\partial H}{\partial z}(z_i, t_{i+1})\right)^{-1} H(z_i, t_{i+1}).$$

where this time $z_0 = \tilde{z}_i$, with \tilde{z}_i and t_{i+1} obtained from the *i*-th Euler step.

Usually, only a few steps of Newton's method are needed; we will use a fixed number of 5 iterations. At this point, we use the final value of the Newton iteration as the starting value for the next Euler step.

2.2.3 Adaptive step size

In order to improve the efficiency of the method, we will use an adaptive step size, which will be based on the norm of the residual of the Newton iteration. If the desired accuracy is not reached, for instance when the norm of $H(z_i, t_i)$ is bigger than 10^{-8} , then we halve the step size; if instead we have 5 "successful" iterations in a row, we double the step size.

3 Parallelization

3.1 Multithreading

When testing the method, we tried to use multithreading to speed up the computation. This was done in Julia by using the Threads.@threads macro, which automatically distributes the work of a for loop among the available threads. However, in the case of looping over multiple roots, this didn't improve the performance, as the overhead of the multithreading was too big compared to the actual computation time, as the systems were too small to benefit from this kind of parallelization, as can be seen by the results in Appendix B.

3.2 MPI

Next, we tried to use MPI to parallelize the tracking of the roots. This was done by using the MPI.jl package, which provides a Julia interface to the MPI library.

4 Appendix A: Implementation

4.1 Julia code

Listing 1: solve.jl

```
# External dependencies
using TypedPolynomials
# Local dependencies
include("start-system.jl")
include("homotopy.jl")
# include("homogenize.jl"
include("euler-newton.jl")
include("adapt-step.jl")
include("plot.jl")
using .StartSystem
using .Homotopy
    using .Homogenize
using .EulerNewton
using .AdaptStep
using .Plot
# Main homotopy continuation loop
# Homotopy contaction toop
function solve(F, (G, roots) = start_system(F), maxsteps = 1000)
# F=homogenize(F)
   H=homotopy(F,G)
   solutions = []
   step_array = []
   Threads.@threads for r in roots
      t = 1.0
      step_size = 0.01
      x0 = r
m = 0
      steps = 0
      while t > 0 && steps < maxsteps</pre>
        x0 = en_step(H, x0, t, step_size)
        (m, step_size) = adapt_step(H, x0, t, step_size, m)
         t -= step_size
        steps += 1
      end
     push!(solutions, x0)
      push!(step_array, steps)
   end
   return (solutions, step_array)
end
# Input polynomial system
# input portionize system

(polyvar x y)

C = [x^3 - y + 5x^2 - 10, 2x^2 - y - 10]

Q = [x^2 + 2y, y - 3x^3]

F = [x*y - 1, x^2 + y^2 - 4]

T = [x*y - 1, x^2 + y^2 - 2]
(sC, stepsC) = solve(C)
(sQ, stepsQ) = solve(Q)
(sF, stepsF) = solve(F)
(sT, stepsT) = solve(T)
println("C: ", stepsC)
```

```
println("Q: ", stepsQ)
println("F: ", stepsF)
println("T: ", stepsT)
sC = filter(u -> imag(u[1]) < 0.1 && imag(u[2]) < 0.1, sC)
sQ = filter(u -> imag(u[1]) < 0.1 && imag(u[2]) < 0.1, sQ)
sF = filter(u -> imag(u[1]) < 0.1 && imag(u[2]) < 0.1, sF)
sT = filter(u -> imag(u[1]) < 0.1 && imag(u[2]) < 0.1, sT)
# Plotting the system and the real solutions
ENV["GKSwstype"]="nul"
plot_real(sC, C, 6, 12, "1")
plot_real(sC, F, 4, 4, "3")
plot_real(sT, T, 4, 4, "4")</pre>
```



```
module StartSystem
using TypedPolynomials
export start_system
# Define start system based on total degree
function start_system(F)
   degrees = [maxdegree(p) for p in F]
   G = [x.i^4d - 1 for (d, x.i) in zip(degrees, variables(F))]
   r = [[exp(2im*pi/d)^k for k=0:d-1] for d in degrees]
   roots = vec([collect(root) for root in collect(Iterators.product(r...))])
   return (G, roots)
end
end
```

Listing 3: homotopy.jl

Listing 4: homogenize.jl

```
module Homogenize
using TypedPolynomials
export homogenize, homogenized_start_system
function homogenize(F)
@polyvar h
return [sum([h^(maxdegree(p)-maxdegree(t))*t for t in p.terms]) for p in F
]
end
```

function homogenized_start_system(F)

```
degrees = [maxdegree(p) for p in F]
@polyvar h
G = [x_i^d - h^d for (d, x_i) in zip(degrees, variables(F))]
r = [[exp(2im*pi/d)^k for k=0:d-1] for d in degrees]
roots = vec([vcat(collect(root), 1) for root in collect(Iterators.product(r
...))])
return (G, roots)
end
end
```

Listing 5: euler-newton.jl

```
module EulerNewton
   using LinearAlgebra
   using TypedPolynomials
   export en_step
   # Euler-Newton predictor-corrector
   function en_step(H, x, t, step_size)
      # Predictor step
     vars = variables(H(t))
     # Jacobian of H evaluated at (x,t)
JH = [jh(vars=>x) for jh in differentiate(H(t), vars)]
     # \partial H/\partial t is the same as VG-F=H(1)-H(0) for our choice of homotopy \Delta x = JH \setminus -[gg(vars=x) \text{ for } gg \text{ in } H(1)-H(0)]
     xh = x + \Delta x * step_size
      # Corrector step
      JHh=differentiate(H(t-step_size), vars)
     for _ in 1:5

JH = [jh(vars=>xh) for jh in JHh]

\Delta x = JH \setminus -[h(vars=>xh) for h in H(t-step_size)]

xh = xh + \Delta x
      end
      return xh
   end
end
```

Listing 6: adapt-step.jl

```
module AdaptStep
  using LinearAlgebra
  using TypedPolynomials
  export adapt_step
  # Adaptive step size
  \begin{array}{l} \Delta = \text{LinearAlgebra.norm}([h(variables(H(t))=>x) \text{ for } h \text{ in } H(t-step)]) \end{array}
    if \Delta > 1e-8
      step = 0.5 * step
       m = 0
     else
       m+=1
       if (m == 5) && (step < 0.05)
        step = 2 * step
         m = 0
       end
     end
     return (m, step)
  end
end
```

Listing 7: plot.jl

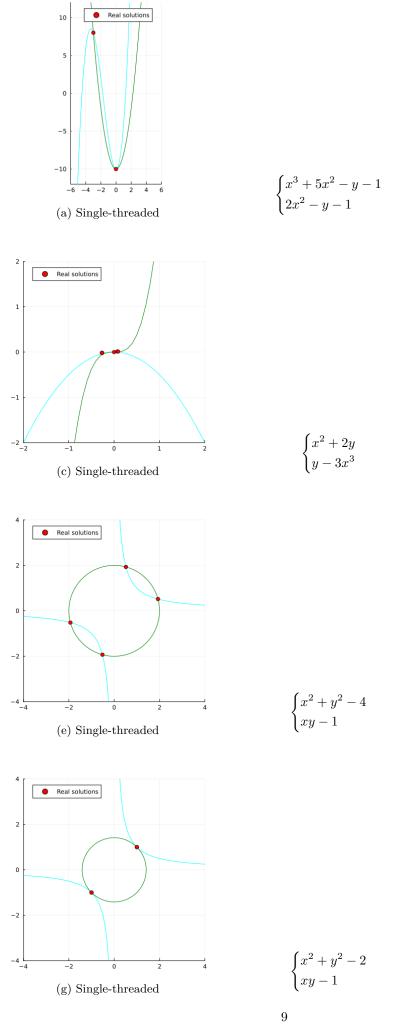
```
module Plot
using Plots, TypedPolynomials
export plot_real
function plot_real(solutions, F, h, v, name)
    plot(xlim = (-h, h), ylim = (-v, v), aspect_ratio = :equal)
    contour!(-h:0.1:h, -v:0.1:v, (x,y)->F[1](variables(F)=>[x,y]), levels=[0],
    cbar=false, color=:cyan)
    contour!(-h:0.1:h, -v:0.1:v, (x,y)->F[2](variables(F)=>[x,y]), levels=[0],
    cbar=false, color=:green)
    scatter!([real(sol[1]) for sol in solutions], [real(sol[2]) for sol in
    solutions], color = "red", label = "Real solutions")
    png(joinpath("plots", "solutions" * name))
end
end
```

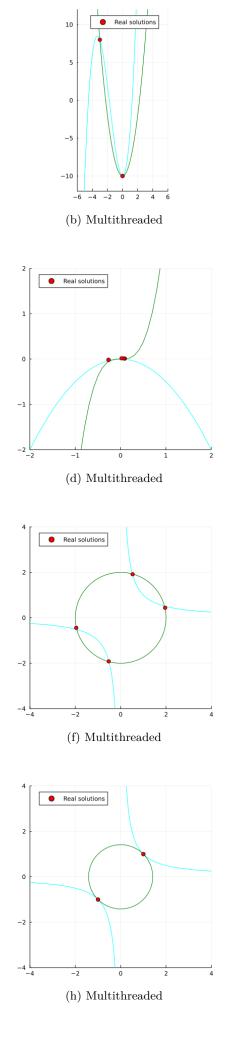
4.2 Hardware

5 Appendix B: Results

5.1 Multithreading

Here are the plots for the solutions of four different 2x2 systems, with the single-threaded version next to the multithreaded one:





References

[1] Bates, Daniel J. Numerically solving polynomial systems with Bertini. SIAM, Society for Industrial Applied Mathematics, 2013.