

newton

June 16, 2020

1 Newton method

```
[1]: using Plots  
plotly()
```

```
[1]: Plots.PlotlyBackend()
```

1.1 Newton-Raphson in optimization

Newton-Raphson iteration

$$x_{k+1} = x_k - \frac{f'(x_k)}{f''(x_k)}$$

```
[2]: function Newton(f::Function, df::Function, d2f::Function,  
    xstart::Float64, verbose::Bool = false, store=false,  
    ::Float64 = 1e-6, nmax::Int64 = 1000)  
    k = 1  
    x = xstart  
    if (store)  
        iter = [ f(x) x ]  
    end  
    if (verbose)  
        fx = f(x)  
        println("$k. x = $x, f(x) = $fx")  
    end  
    dfx = df(x)  
    while (abs(dfx) > 1e-6 && k < nmax)  
        k += 1  
        dfx = df(x)  
        x = x-dfx/d2f(x)  
        if (store)  
            iter = [iter ; f(x) x]  
        end  
        if (verbose)  
            fx = f(x)  
            println("$k. x = $x, f(x) = $fx")  
        end  
    end  
end
```

```
end

if (store)
    return iter
end
end
```

[2]: Newton (generic function with 5 methods)

```
[3]: func(x) = -10x^2 + 4sin(x) + x^4
      dfunc(x) = -20x + 4cos(x) + 4x^3
      d2func(x) = -20 - 4sin(x) + 12x^2
```

[3]: d2func (generic function with 1 method)

```
[4]: plot(func, -4.0, 4.0)
```

If we start close enough to the global optimum, we are to find it.

```
[5]: x0 = -3.0
      iter = Newton(func, dfunc, d2func, x0, true, true)
```

```
1. x = -3.0, f(x) = -9.564480032239473
2. x = -2.413309150943051, f(x) = -26.983281625936534
3. x = -2.2051283480850294, f(x) = -28.202989762523757
4. x = -2.1772606979392806, f(x) = -28.219314490011172
5. x = -2.1767739238815578, f(x) = -28.21931925035788
6. x = -2.1767737764195543, f(x) = -28.21931925035831
7. x = -2.176773776419541, f(x) = -28.219319250358318
```

```
[5]: 7×2 Array{Float64,2}:
      -9.56448  -3.0
      -26.9833  -2.41331
      -28.203   -2.20513
      -28.2193  -2.17726
      -28.2193  -2.17677
      -28.2193  -2.17677
      -28.2193  -2.17677
```

```
[6]: iter
```

```
[6]: 7×2 Array{Float64,2}:
      -9.56448  -3.0
      -26.9833  -2.41331
      -28.203   -2.20513
      -28.2193  -2.17726
      -28.2193  -2.17677
      -28.2193  -2.17677
```

```
-28.2193 -2.17677
```

```
[7]: plot(func, -3.2, -2.0, label="Function")
x = -3.0
m(y) = func(x)+dfunc(x)*(y-x)+0.5*d2func(x)*(y-x)^2
m(-3.0)
```

```
[7]: -9.564480032239473
```

```
[8]: plot!(m, -3.2, -2.0, label="Model")
plot!(iter[1:2,2], iter[1:2,1], label="Newton step")
vline!([iter[1,2] iter[2,2]], label = "")
```

```
[9]: plot(func, -3.2, -2.0, label="Function")
plot!(iter[:,2], iter[:,1], label="Newton steps")
```

```
[10]: x0 = -4.0
Newton(func, dfunc, d2func, x0, true)
```

```
1. x = -4.0, f(x) = 99.02720998123172
2. x = -2.942938833739946, f(x) = -12.3872911024757
3. x = -2.3879767853105314, f(x) = -27.24370765699775
4. x = -2.199836344922852, f(x) = -28.20853988049366
5. x = -2.177097516735148, f(x) = -28.219317146168752
6. x = -2.1767738416160567, f(x) = -28.21931925035823
7. x = -2.176773776419543, f(x) = -28.219319250358318
8. x = -2.1767737764195405, f(x) = -28.219319250358314
```

```
[11]: x0 = -2.2
Newton(func, dfunc, d2func, x0, true)
```

```
1. x = -2.2, f(x) = -28.208385615278356
2. x = -2.177102076816536, f(x) = -28.219317086469566
3. x = -2.176773843465362, f(x) = -28.219319250358225
4. x = -2.1767737764195436, f(x) = -28.219319250358314
5. x = -2.176773776419541, f(x) = -28.219319250358318
```

However, if the method converges, it converges to a point where the derivative is equal to zero. This could be a local maximum!

```
[ ]: x0 = 1.0
Newton(func, dfunc, d2func, x0, true)
```

```
[ ]: = 0.1
x = x0- *dfunc(x0)/d2func(x0)
= 0.001
x = x0- *dfunc(x0)/d2func(x0)
```

```
[ ]: f(x)
```

```
[ ]: dfunc(x0)/d2func(x0)
```

```
[ ]: dfunc(x0)
```

We can also converge to local, but not global, minimum.

```
[ ]: x0 = 2.0  
Newton(func, dfunc, d2func, x0, true)
```

1.2 Secant method

```
[ ]: function Secant(f::Function, df::Function, x0::Float64, x1::Float64,  
    verbose::Bool = false, ::Float64 = 1e-6, nmax::Int64 = 1000)  
    k = 1  
    x = x0  
    y = x1  
    if (x0 == x1)  
        println("x0 must different from x1")  
        return  
    end  
    if (verbose)  
        println("0. x0 = $x0, f($x0) = $(f(x0))")  
        println("1. x1 = $x1, f($x1) = $(f(x1))")  
    end  
    dfx = df(x)  
    dfy = df(y)  
    while (abs(dfy) > && k < nmax)  
        k += 1  
        t = y  
        y = y-(x-y)/(dfx-dfy)*dfy  
        x = t  
        if (verbose)  
            println("$k. x = $y, f(x) = $(f(y))")  
        end  
        dfx = dfy  
        dfy = df(y)  
    end  
  
    return y  
end
```

```
[ ]: Secant(func, dfunc, 1.0, 2.0, true)
```

1.3 Root finding using Newton method

Iteration

$$x_{k+1} = x_k - \frac{f(x)}{f'(x)}$$

Consider the function

```
[ ]: f(x) = x-2sin(x)
```

```
[ ]: df(x) = 1-2cos(x)
```

```
[ ]: plot(f, -5.0, 5.0)
```

We are looking for a zero of $f(x)$.

```
[ ]: function NewtonRoot(f::Function, df::Function, xstart::Float64,
    verbose::Bool = false, ::Float64 = 1e-6, nmax::Int64 = 1000)
    k = 1
    x = xstart
    fx = f(x)
    if (verbose)
        println("$k. x = $x, f(x) = $fx")
    end
    while (abs(fx) > 1e-6 && k < nmax)
        k += 1
        dfx = df(x)
        x = x-fx/df(x)
        fx = f(x)
        if (verbose)
            println("$k. x = $x, f(x) = $fx")
        end
    end
    return x
end
```

```
[ ]: x0 = 1.1
    NewtonRoot(f, df, x0, true)
```

It works, but we were close to a disaster! Observe that

```
[ ]: df(1.1)
```

The curve is nearly flat, and we move to a distant point. Luckily we come back, but using many iterations. Consider now the starting point

```
[ ]: x0 = 1/3
```

The derivative at this point is

```
[ ]: df(x0)
```

The Newton method now gives

```
[ ]: NewtonRoot(f, df, x0, true)
```

We are less lucky! In fact, at $\pi/3$,

```
[ ]: df(/3)
```

The Newton recurrence is under trouble as we have a division by zero. The function however proceeds as due to numerical errors, we have avoided the division by zero, but the method diverges, as $x \rightarrow -\infty$.

Take now a point with a significant derivative.

```
[ ]: x0 = 4.0
NewtonRoot(f, df, x0, true)
```

The method now converges very quickly, even if the starting point was further from the solution.

Note the $x - 2 \sin x = 0$ is equivalent to $\frac{1}{\sin x} - \frac{2}{x} = 0$ if we require that $x \neq k\pi$, $k \in \mathcal{Z}$. The function shape is nevertheless quite different around the function zero.

```
[ ]: g(x) = 1/sin(x) - 2/x
plot(g, 0.1, -0.1)
```

The derivative is

```
[ ]: dg(x) = 2/(x*x)-cos(x)/(sin(x)^2)
```

```
[ ]: x0 = 1.1
NewtonRoot(g, dg, x0, true)
```

We now observe a fast convergence, mainly due to the fact that the function does not present flat parts.

```
[ ]: f(x) = exp(x/2)-x-1
```

```
[ ]: df(x) = 0.5*exp(x/2)-1
```

```
[ ]: x0 = 1.0
NewtonRoot(f, df, x0, true)
```

```
[ ]: using Roots
```

```
[ ]: y = fzero(df,0.0, 5.0)
```

```
[ ]: x0 = y
NewtonRoot(f, df, x0, true)
```

1.4 Cycles

Consider now the function

```
[ ]: h(x) = x^3 - 2*x + 2
```

```
[ ]: plot(h, -2.5, 1.5)
```

```
[ ]: dh(x) = 3x^2-2
```

```
[ ]: x0 = 0.0
      NewtonRoot(h, dh, x0, true)
```

```
[ ]: ratio(x) = h(x)/dh(x)
```

```
[ ]: ratio(0.0)
```

```
[ ]: ratio(1.0)
```

The method cycles! However, if we change the starting point, we can converge.

```
[ ]: x0 = 0.5
      NewtonRoot(h, dh, x0, true)
```

1.5 Application: computation of the square root of a positive number.

The square root of a real number can be computed using the Newton method. More precisely, if we look for the square root of s , we can rewrite the problem as the computation of a zero of the function

$$f(x) = x^2 - s$$

Indeed, if $f(x) = 0$, then $s = x^2$ or $x = \sqrt{s}$. Develop the Newton recurrence for this problem.

Choosing a good starting point can however be an issue. (Wikipedia) With s expressed in scientific notation as $a \times 10^{2n}$ where $1 \leq a < 100$ and n is an integer, $\sqrt{s} \approx \sqrt{a}10^n$. The starting point is often computed as

$$\sqrt{s} = \begin{cases} 2 \times 10^n & \text{if } a < 10 \\ 6 \times 10^n & \text{if } a \geq 10 \end{cases}$$

The factors two and six are used because they approximate the geometric means of the lowest and highest possible values with the given number of digits:

$$\sqrt{\sqrt{1}\sqrt{10}} = 10^{\frac{1}{4}} \approx 2$$

and

$$\sqrt{\sqrt{10}\sqrt{100}} = 10^{\frac{3}{4}} \approx 6$$

```
[ ]: sq(x,s) = x*x-s
```

```
[ ]: dsq(x) = 2*x
```

```
[ ]: sqs(x) = x*x-s
```

```
[ ]: s = 6.0
```

```
[ ]: x = NewtonRoot(sqs, dsq, 2.0, true)
```

We can check the solution as

```
[ ]: x^2
```

```
[ ]: x = NewtonRoot(sqs, dsq, 2.0, true, 1e-8)
```

```
[ ]: x^2
```

```
[ ]: x = NewtonRoot(sqs, dsq, 2.0, true, 1e-10)
x^2
```

```
[ ]: x = NewtonRoot(sqs, dsq, 2.0, true, 1e-12)
x^2
```

```
[ ]: x = NewtonRoot(sqs, dsq, 2.0, true, 1e-14)
x^2
```

```
[ ]: x = NewtonRoot(sqs, dsq, 2.0, true, 1e-16)
x^2
```

```
[ ]: x = NewtonRoot(sqs, dsq, 2.0, true, 1e-15)
x^2
```

```
[ ]: 6-x^2
```

```
[ ]: eps()
```

```
[ ]: s = 25
x = NewtonRoot(sqs, dsq, 6.0, true, 1e-15)
x^2
```

```
[ ]: s = 400
x = NewtonRoot(sqs, dsq, 6.0, true, 1e-15)
x^2
```

```
[ ]:
```