Project 2 in FYS3150

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1 ABSTRACT

In this project we have implemented Jacobian algorithm, to find eigenvectors, and their corresponding eigenvalues in tridiagonal matrices. We then used this to model a harmonic oscillator problem in three dimensions, with one and two electrons. This turned out to be a computationally heavy, but relatively accurate method.

2 INTRODUCTION

Finding eigenvectors analytically is complicated, and can be tedious, and this is why it is much more convenient to do so numerically. A common way of doing this is by the application of the Jacobian method. Essentially we rotate one matrix element at a time, always taking the one with the highest absolute value, until all but the diagonal elements are essentially zero. All transformations are also applied to an identity matrix that then turns into our eigenvectors.

3 METHOD

3.a Implementing the Jacobian algorithm

The implementation follows a standard recipe: See the documentation in section 6.a for further explanation.

$$\cot 2\theta = \tau = \frac{a_{ll} - a_{kk}}{2a_{kl}}.$$

We can then define the angle θ so that the non-diagonal matrix elements of the transformed matrix a_{kl} become non-zero and we obtain the quadratic equation (using $\cot 2\theta = 1/2(\cot\theta - \tan\theta)$

$$t^2 + 2\tau t - 1 = 0,$$

resulting in

$$t = -\tau \pm \sqrt{1+\tau^2},$$

and c and s are easily obtained via

$$c = \frac{1}{\sqrt{1+t^2}},$$

and $s = tc$.

$$1 = 2 \tag{1}$$

$$3 = \tag{2}$$

3.b Testing the code

For testing the algorithm we have implemented two tests. One for checking if the largest element in the matrix is correctly located, and one for testing if the resulting eigenvalues are correct. The first one is more useful for development purposes, whilst the second one is essential for validating that our implementation works correctly.

3.c Quantum dots in three dimensions, one electron

3.d Quantum dots in three dimensions, two electrons

4 RESULTS

5 CONCLUSIONS

6 APENDICES

6.a Integration loop from rotator.jl

```
function maxKnotL(a)
    max = 0
    kl = [1,1]
    n = Int64(length(a[1,:]))
    for k = 1:n
        for l = k+1:n
            ma = abs(a[k, l])
            if (ma > max)
                max = ma
                kl = [k, l]
            end
        end
    end
    return kl[1], kl[2]                #k, l
end

function rotate(a, tol)           #the actual integration loop. Takes tridiagonal matrix an tolerance for precission
    n = Int64(length(a[1,:]))     #initiates n for later use
    r = Matrix{Float64}(I, n, n)  #initialising eigenvector matrix
    counter = 0                   #teller antall "similarity transformaitons"
    k, l = maxKnotL(a)            #finds indices of matrix element with highest value
    while abs(a[k ,l]) > tol      #this is the actual loop
        counter += 1
        if (a[k, l] != 0.0)
            #kl = maxKnotL(a)
            tau = (a[l, l] - a[k, k])/2*a[k, l] #blir ikke dette alltid null?
            if (tau > 0)
                t = -tau + sqrt(1.0+tau^2)
            else
                t = -tau - sqrt(1.0+tau^2)
            end
            c = 1/sqrt(1.0+t^2)
            s = c*t
        #end
        else
            c = 1.0
            s = 0.0
        end
        a_kk = a[k, k]
        a_ll = a[l, l]
        #doing stuff with indices k and l
        c2 = c^2
        s2 = s^2
        csakl2 = 2.0*c*s*a[k, l]
        a[k, k] = c2*a_kk - csakl2 + s2*a_ll
        a[l, l] = s2*a_kk + csakl2 + c2*a_ll
        a[k, l] = 0
        a[l, k] = 0
        #doing stuff with the remaing matrix elements
        for i = 1:n
            if ((i != k) && (i != l))
                a_ik = a[i, k]
                a_il = a[i, l]
                a[i, k] = c*a_ik - s*a_il
                a[k, i] = a[i, k]
                a[i, l] = c*a_il + s*a_ik
```

```
            a[l, i] = a[i, l]
        end
        #calculating eigenvectors
        r_ik = r[i, k]
        r_il = r[i, l]
        r[i, k] = c*r_ik - s*r_il
        r[i, l] = c*r_il + s*r_ik
    end
```

7 REFERENCES

References

[1] Computational Physics, Lecture Notes Fall 2015, Morten Hjort-Jensen p.215-220