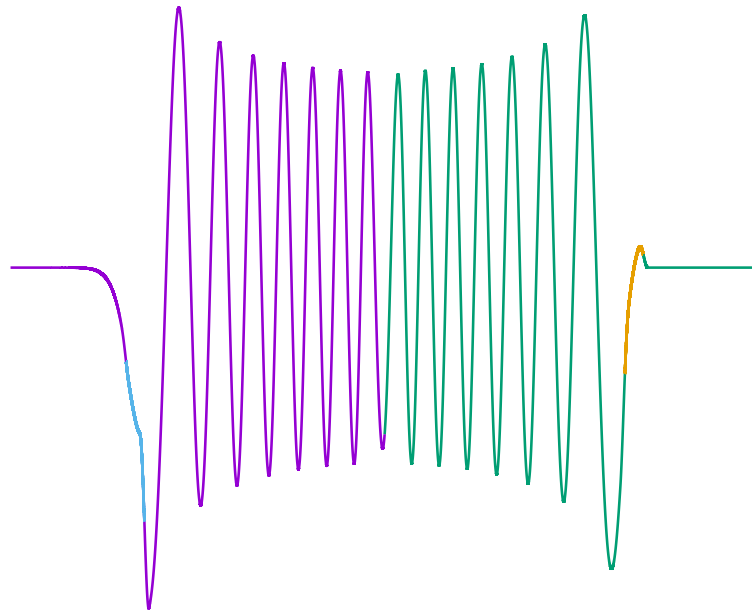


Approximating Solutions of the Time Independent Schrödinger Equation

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1. Introduction

Richard Feynmann one of the core people behind our modern theory of quantum mechanics repeatedly said: “I think I can safely say that nobody understands quantum mechanics.”. Nothing behaves like in our every day lives. Everything is just a probability and nothing certain. Even Schrödinger the inventor of the equation that governs all of those weird phenomena rejected the idea that there are just probabilities.

In this paper we will try to understand this world a little bit better by looking at wave functions in a simplified universe. This universe only has 1 dimension and there will not be any sense of time. This means we will be able to actually see how the wave function looks like in a graph.

1.1. Goals

The goal of this Maturaarbeit is to write a program, `schroeding-approx` that calculates solutions to the time independent Schrödinger equation in 1 dimension for a large variety of potentials. We assume that the wave function, $\Psi(x)$ will diverge to 0 as x goes to infinity.

The program uses Planck units which means $c = \hbar = G = 1$

2. Preliminary

2.1. Schrödinger Equation

In 1926 Erwin Schrödinger changed our understanding of quantum physics with the Schrödinger equation. Based on the observations of de Broglie that particles behave like waves he developed a wave equation which describes how the waves move and change in a given potential $V(x)$ or Hamiltonian \hat{H} .

$$i\hbar \frac{\partial}{\partial t} \Psi(x, t) = \left[-\frac{\hbar^2}{2m} \frac{\partial^2}{\partial x^2} + V(x, t) \right] \Psi(x, t)$$

Or more general

$$i\hbar \frac{\partial}{\partial t} \Psi(x, t) = \hat{H} \Psi(x, t)$$

The time independent version that is going to be used later, ignores the change over time and is much simpler to solve since it is **only** an ordinary differential equation instead of a partial differential equation.

$$E\psi(x) = \hat{H}\psi(x)$$

or

$$-\frac{\hbar^2}{2m} \frac{d^2\psi}{dx^2}(x) + V(x)\psi(x) = E\psi(x)$$

Even with the time independent equation it is very difficult to get analytical solutions, because of this there are mainly three approaches to approximate solutions of $\psi(x)$, perturbation theory, density functional field theory and WKB approximation. Perturbation theory's goal is to give an analytical approximation which means it is extremely difficult to implement for a computer. WKB on the other hand is much better since it is to some degree a step by step manual.

2.2. Rust

Rust is one of the newer programming languages and attempts to replace C/C++ which are notoriously difficult to work with. It supports both functional and object-oriented paradigms. It is much safer in terms of memory and promises the same performance as C. One of the goals of Rust is fearless concurrency which means everybody should be able to write concurrent code without deadlocks and data races. This means calculations can utilize the full potential of the CPU without countless hours of debugging.

Functional programming languages are especially useful for mathematical problems, because they are based on the same mathematics as the problem.

Rust as of the time of writing this document is not yet standardized meaning the code provided might no longer be correct with one of the newer Rust versions.

In case you aren't familiar with Rust you it has excellent documentation on <https://doc.rust-lang.org/book/>.

2.3. Complex Numbers

In quantum mechanics it's customary to work with complex numbers. Complex numbers are an extension to the real numbers, since Rust will do most of the heavy lifting here are the most important things that you should know

$$i^2 = -1$$

$$z = a + bi$$

$$\text{Re}(z) = a$$

$$\text{Im}(z) = b$$

$$\bar{z} = a - bi$$

$$\|z\|^2 = a^2 + b^2$$

$$e^{a+bi} = e^a (\cos(b) + i \sin(b))$$

i is the imaginary unit, z is the general form of a complex number where $\{a, b\} \in \mathbb{R}$, \bar{z} is the complex conjugate and $\|z\|^2$ is the norm square of z . The last equation is the Euler's formula, it rotates a number in the complex plane by b radians.

The complex plane is similar to the real number line, every complex number can be represented on this plane where $\text{Re}(z)$ is the x-coordinate and $\text{Im}(z)$ is the y-coordinate.

2.4. Gnuplot

Gnuplot is a cross platform plotting program that is very simple to use. `schroedinger-approx` will output a file `data.txt`, you can plot the function by typing `gnuplot` and then typing

```
call "plot.gnuplot"
```

to plot the real part of the wave function, or

```
call "plot_3d.gnuplot"
```

to see the full complex wave function.

If you'd like to learn more about Gnuplot you can read there user manual on <http://www.gnuplot.info/>

3. Methods

3.1. Program Architecture

The program has multiple interfaces or traits as they are called in Rust that give the program some abstraction.

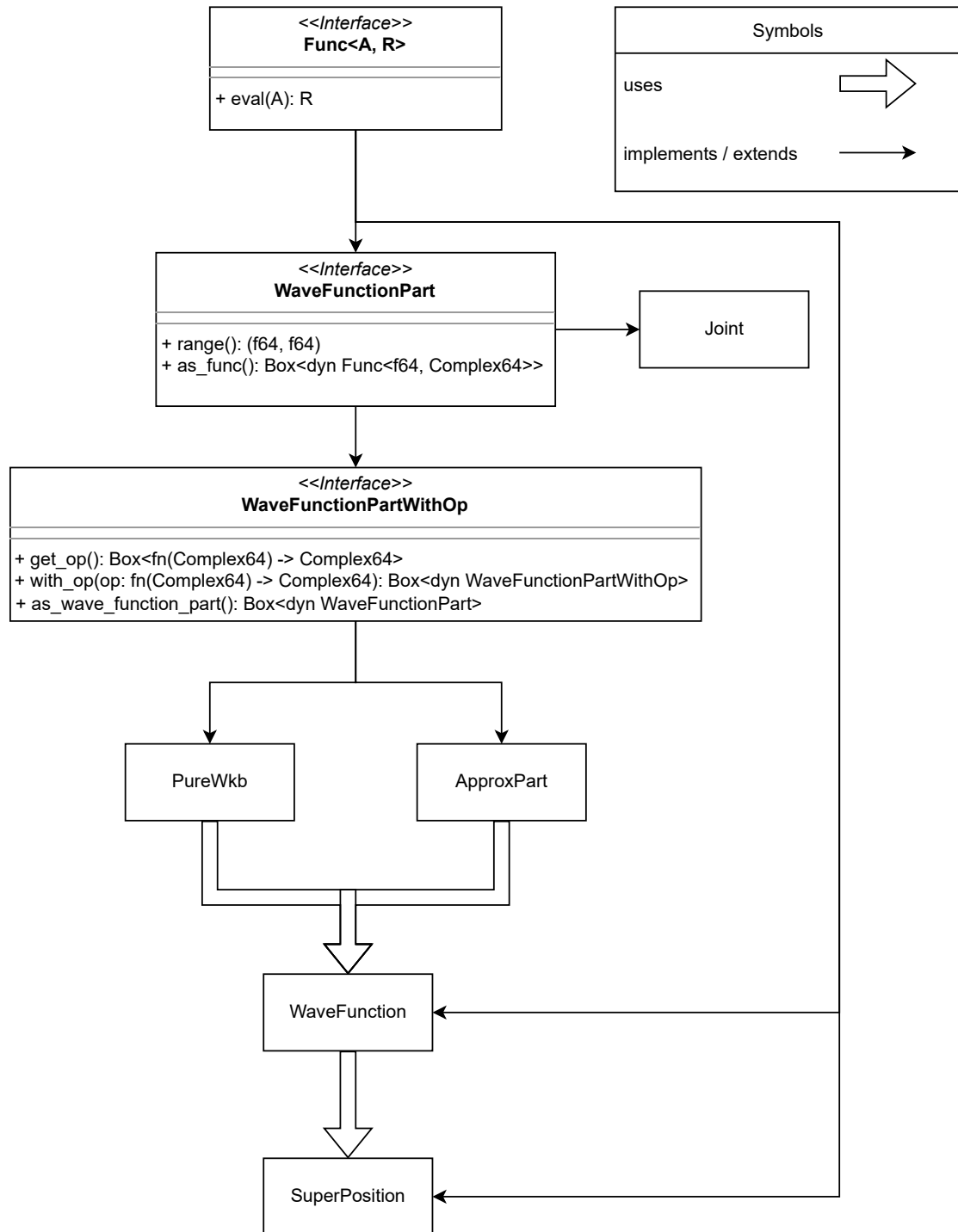


Figure 3.1.: UML diagram of program architecture

Since current version of Rust does not support manual implementations of `std::ops::Fn`

we have to define our own trait for functions `Func<A, R>` where `A` is the type of the argument and `R` is the return type. Later we will use this trait to implement functions for integration, evaluation and more useful utilities.

`WaveFunction` is at the heart of the program, it contains all the functionality to build wave functions. It is composed of `WaveFunctionPart` which represent either a `Joint`, `PureWkb` or an `ApproxPart`. With the `range` function we can check when they are valid.

3.2. Newtons Method

Newton's method, also called the Newton-Raphson method, is a root-finding algorithm that uses the first few terms of the Taylor series of a function $f(x)$ in the vicinity of a suspected root (Weisstein, 2022). It makes a sequence of approximations of a root x_n that in certain cases converges to the exact value where

$$\lim_{n \rightarrow \infty} f(x_n) = 0$$

The sequence is defined as

$$x_0 = a$$

$$x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)}$$

Visually this looks like figure 3.2 $f(x) = (x - 2)(x - 1)(x + 1)$.

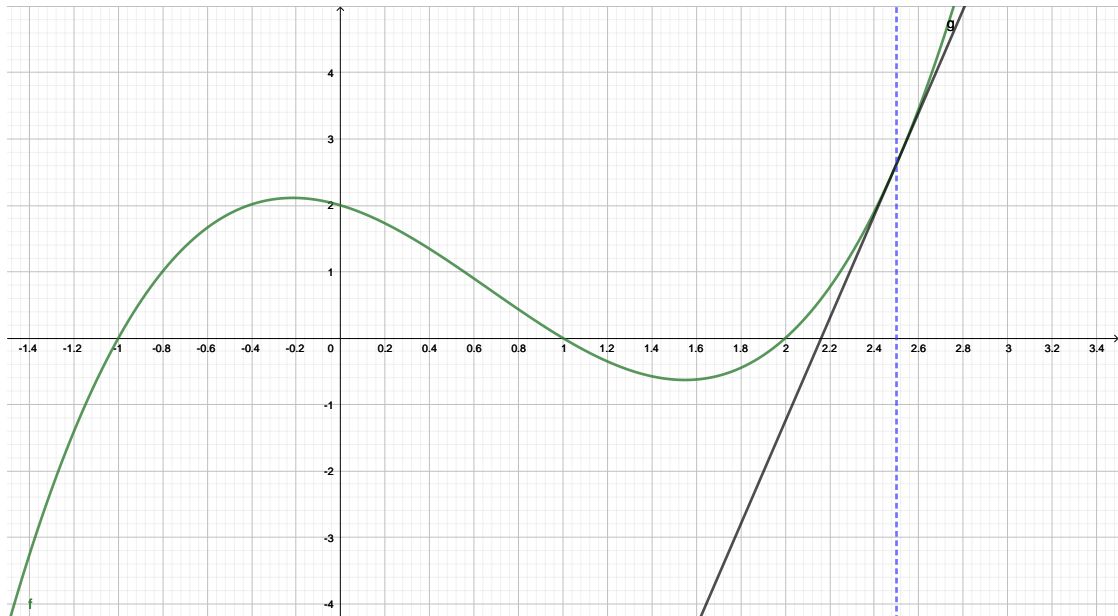


Figure 3.2.: Illustration of Newtons method, $f(x) = (x - 1)(x - 2)(x + 1)$.

The blue line indicates the initial guess which in this case is 2.5 the black line ($g(x)$) is a tangent to $f(x)$ at $(guess, f(guess))$ the next guess will be where the tangent intersects the x-Axis (solution of $g(x) = 0$). This will converge rather quickly compared to other methods such as Regula-Falsi.

```

1 pub fn newtons_method<F>(f: &F, mut guess: f64, precision: f64) -> f64
2     where
3         F: Fn(f64) -> f64,
4     {
5         loop {
6             let step = f(guess) / derivative(f, guess);
7             if step.abs() < precision {
8                 return guess;
9             } else {
10                 guess -= step;
11             }
12         }
13     }

```

In Rust the sequence is implemented with a function that takes a closure f , the initial guess $guess$ and a stop condition $precision$ the function will return if $|\frac{f(x_n)}{f'(x_n)}|$ is less than $precision$.

From the structure of the algorithm it is very tempting to implement it recursively, but by using a loop it is much faster since there are no unnecessary jumps and the precision can (at least in theory) be 0 without causing a stack overflow.

3.3. Derivatives

Derivatives can be calculated numerically as in the C++ library Boost (John Maddock, 2022). The author implemented a analytical system for derivatives in Go. From that experience the benefit is negligible compared to the increase in performance and in development time since every function is a special object.

```

1 pub fn derivative<F, R>(func: &F, x: f64) -> R
2     where
3         F: Fn(f64) -> R + ?Sized,
4         R: Sub<R, Output = R> + Div<f64, Output = R> + Mul<f64, Output = R> + Add<R, Output = R>,
5     {
6         let dx = f64::epsilon().sqrt();
7         let dx1 = dx;
8         let dx2 = dx1 * 2.0;
9         let dx3 = dx1 * 3.0;
10
11         let m1 = (func(x + dx1) - func(x - dx1)) / 2.0;
12         let m2 = (func(x + dx2) - func(x - dx2)) / 4.0;
13         let m3 = (func(x + dx3) - func(x - dx3)) / 6.0;
14
15         let fifteen_m1 = m1 * 15.0;
16         let six_m2 = m2 * 6.0;
17         let ten_dx1 = dx1 * 10.0;
18
19         return ((fifteen_m1 - six_m2) + m3) / ten_dx1;
20     }

```

`f64::epsilon().sqrt()` is approximately 0.000000014901161. `f64::epsilon()` is the smallest double precision floating point number where $1 + \epsilon \neq 1$. this has been chosen for dx because it should be fairly precise.

3.4. Integration

The same principles apply to integrals as to derivative it would not be a great benefit to implement an analytic integration system. Integrals would also be much more difficult to implement than derivatives since integrals can not be broken down in to many smaller integrals that can be computed easily instead it needs to be solved as is.

One approach would be to use the same method as with the derivative, take the definition with the limit and use a small value but this method can be improved in this case, since integrals calculate areas under curves a trapeze is more efficient and accurate then the rectangle that results from the definition.

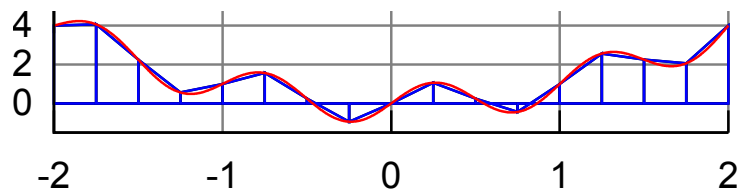


Figure 3.3.: Illustration of integration with trapeze from Wikipedia (2022a).

Figure 3.3 shows visually how the methods work, each blue trapeze from start (a) to end (b) has an area of

$$\int_a^b f(x) dx \approx (b-a)f\left(\frac{a+b}{2}\right).$$

One trapeze would be fairly inaccurate to calculate the area under the function but as the area from a to b is subdivided further the result become better and better.

The general structure of the algorithm can very easily be run in parallel since it doesn't matter in which order the segments are added together and the segments also don't dependent on one another. In Rust this is implemented using rayon. Rayon is an implementation for parallel iterators meaning that normal data structures that implement `std::iter` can be run in parallel *just* by changing `::iter()` to `::par_iter()`. This might not work in all cases because of memory safety but in this case the borrow checker will throw an error and the code wont compile.

```
1 pub trait ReToC: Sync {
2     fn eval(&self, x: &f64) -> Complex64;
3 }
4
5 pub struct Point {
6     pub x: f64,
```

```

7     pub y: Complex64,
8 }

```

These functions were implemented very early and need some refractory. Such that functions with states, like wave functions that store parameters, can be integrated there is a trait ReToC. ReToC describes a function $f : \mathbb{R} \rightarrow \mathbb{C}$ (Fn(f64) -> Complex64).

Point stores both the input (x) and the output (y) of a function.

```

1 pub fn evaluate_function_between(f: &dyn ReToC, a: f64, b: f64, n: usize) -> Vec<Point> {
2     if a == b {
3         return vec![];
4     }
5
6     (0..n)
7         .into_par_iter()
8         .map(|i| index_to_range(i as f64, 0.0, n as f64 - 1.0, a, b))
9         .map(|x| Point { x, y: f.eval(&x) })
10        .collect()
11 }

```

ReToC can be passed to evaluate_function_between it calculates n points between an interval from a to b and returns a vector of Point.

```

1 pub fn trapezoidal_approx(start: &Point, end: &Point) -> Complex64 {
2     return complex(end.x - start.x, 0.0) * (start.y + end.y) / complex(2.0, 0.0);
3 }
4
5 pub fn index_to_range(x: f64, in_min: f64, in_max: f64, out_min: f64, out_max: f64) -> f64 {
6     return (x - in_min) * (out_max - out_min) / (in_max - in_min) + out_min;
7 }
8
9 pub fn integrate(points: Vec<Point>, batch_size: usize) -> Complex64 {
10     if points.len() < 2 {
11         return complex(0.0, 0.0);
12     }
13
14     let batches: Vec<&[Point]> = points.chunks(batch_size).collect();
15
16     let parallel: Complex64 = batches
17         .par_iter()
18         .map(|batch| {
19             let mut sum = complex(0.0, 0.0);
20             for i in 0..(batch.len() - 1) {
21                 sum += trapezoidal_approx(&batch[i], &batch[i + 1]);
22             }
23             return sum;
24         })
25         .sum();
26
27     let mut rest = complex(0.0, 0.0);
28
29     for i in 0..batches.len() - 1 {
30         rest += trapezoidal_approx(&batches[i][batches[i].len() - 1], &batches[i + 1][0]);
31     }
32
33     return parallel + rest;
34 }

```

The actual integration happens in `integrate`, it calculates the areas of the trapezes between the points passed to it. For optimization 1000 trapezes are calculated per thread because it would take more time to create a new thread then to actually do the calculation, this has to be further investigated and 1000 might not be optimal. The calculations performed per thread are called a batch, after all batches have been calculated the boundaries between batches also has to be considered therefor they are added in the end with `rest`

3.5. Transition Regions

The approximation that will be used splits $\Psi(x)$ into multiple parts that do not match perfectly together.

Lets consider an example, in figure 3.5 we can see two Taylor series of cosine. Now we

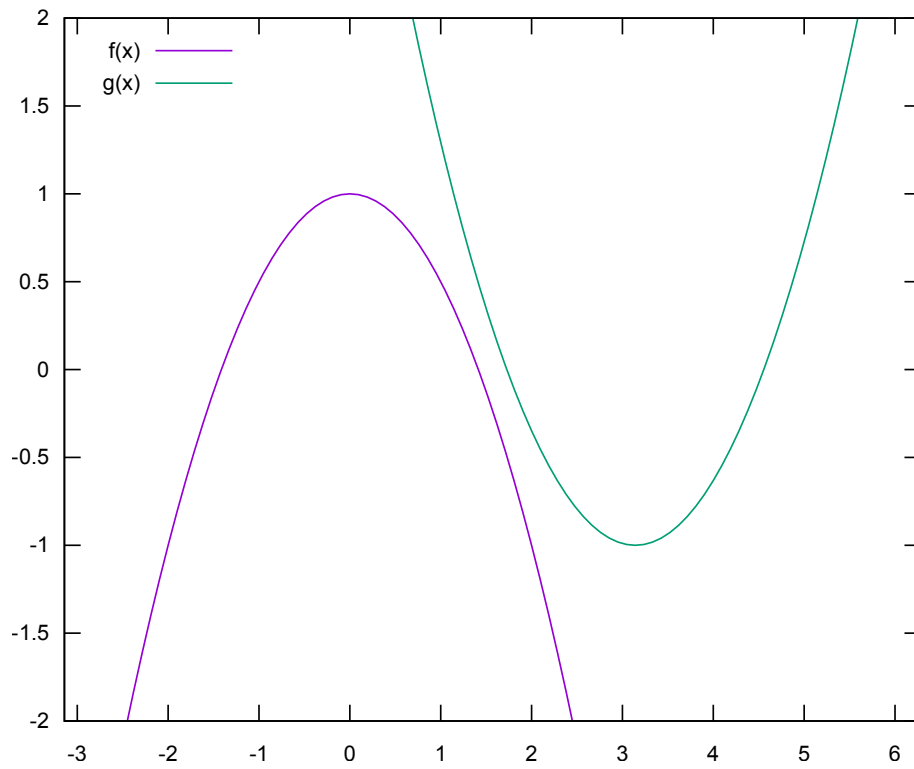


Figure 3.4.: Example for joining functions

have to join the two functions at $x = \pi/2$ such that its a mathematically smooth transition.

$$f(x) = 1 - \frac{x^2}{2} \quad (3.1)$$

$$g(x) = \frac{(x - \pi)^2}{2} - 1 \quad (3.2)$$

As a first guess lets join $f(x)$ and $g(x)$ with a step function, this means that the joint function $h(x)$ will be

$$h(x) = \begin{cases} f(x) & x < \frac{\pi}{2} \\ g(x) & x > \frac{\pi}{2} \end{cases} .$$

This gives us 3.5 which is obviously not smooth.

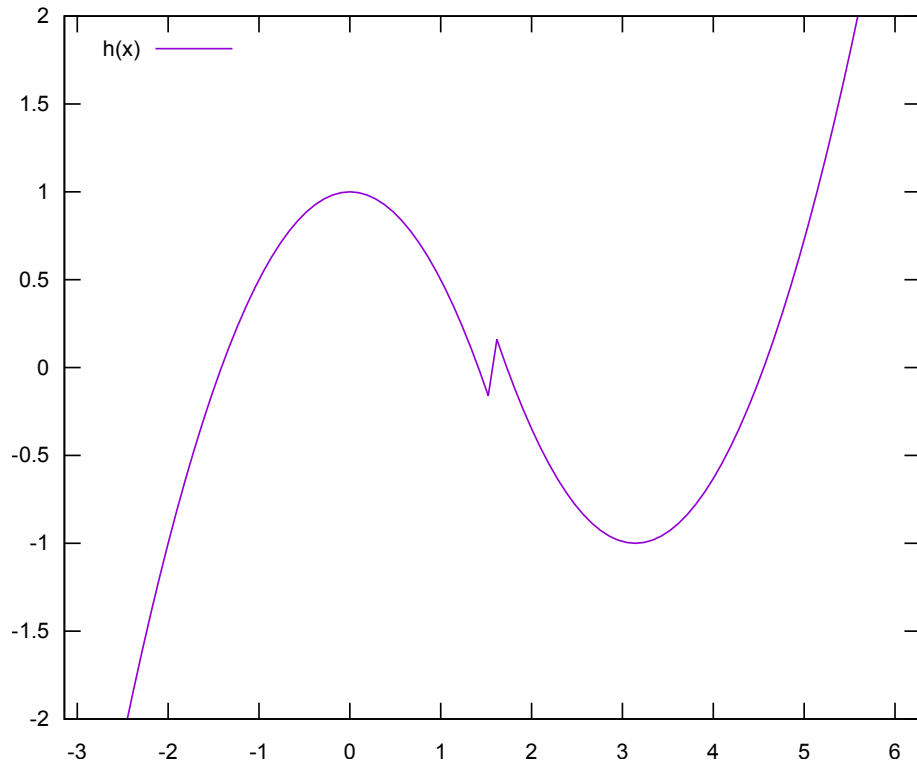


Figure 3.5.: Plot of $h(x)$ with step joint

If we use the formula from (Hall, 2013, p. 325, section 15.6.4) with

$$\begin{aligned}\delta &= 0.5 \\ \alpha &= \frac{\pi}{2} - \frac{\delta}{2} \\ \chi(x) &= \sin^2\left(x \frac{\pi}{2}\right)\end{aligned}$$

this results in

$$h(x) = \begin{cases} f(x) & x < \alpha \\ g(x) & x > \alpha + \delta \\ f(x) + (g(x) - f(x))\chi\left(\frac{x-\alpha}{\delta}\right) & \text{else} \end{cases}$$

which is mathematically smooth as we can see in figure 3.5 (proof in Appendix B.1.1).

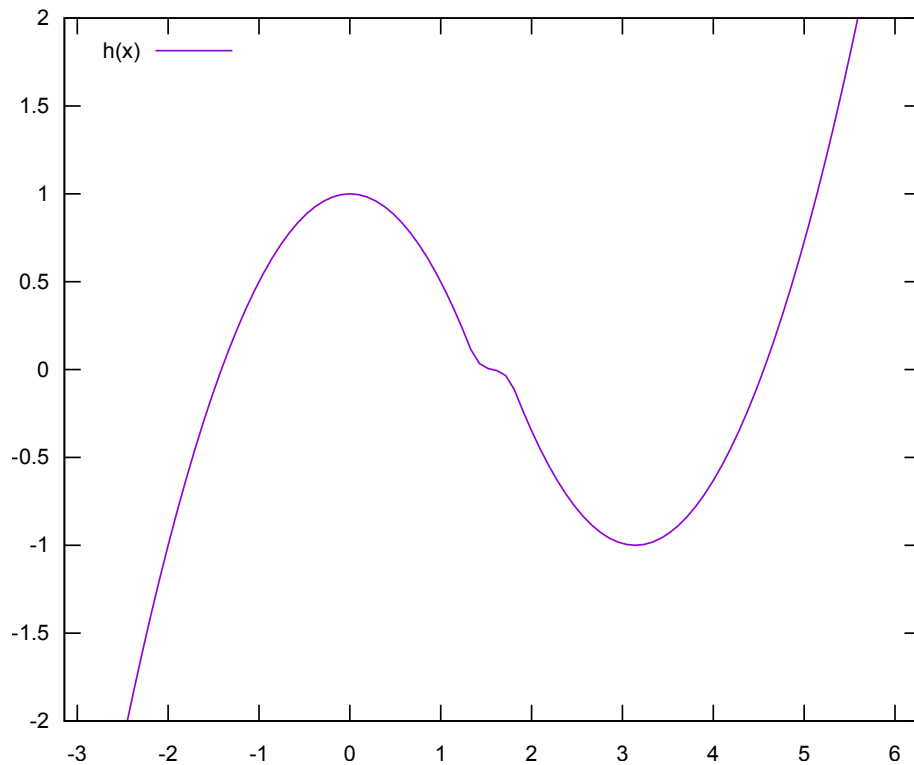


Figure 3.6.: Plot of $h(x)$ with Hall joint

3.5.1. Implementation in Rust

In the program we can define a struct `Joint` that implements `Func<f64, Complex64>`. As in the example we need two functions $f(x)$ and $g(x)$ which we will rename to `left` and `right`. We will also need a variable α and δ which will be named `cut` and `delta`.

```

1  #[derive(Clone)]
2  pub struct Joint {
3      pub left: Arc<dyn Func<f64, Complex64>>,
4      pub right: Arc<dyn Func<f64, Complex64>>,
5      pub cut: f64,
6      pub delta: f64,
7  }
8
9  impl Func<f64, Complex64> for Joint {
10     fn eval(&self, x: f64) -> Complex64 {
11         let chi = |x: f64| f64::sin(x * f64::consts::PI / 2.0).powi(2);
12         let left_val = left.eval(x);
13         return left_val + (right.eval(x) - left_val) * chi((x - self.cut) / self.delta)
14     }
15 }

```


In the proof we assume that $f(x)$ and $g(x)$ are continuous of first order in the interval $(\alpha, \alpha + \delta)$. In the code we will not check this requirement since it would have a major impact on performance to check the derivative on every point.

3.6. Energy Levels

Solving the Schrödinger equation is an eigenvalue problem. This means that only certain energies will result in physically correct results. For an energy to be valid it has to satisfy the Maslov-corrected Bhor-Sommerfeld condition which states that

$$n \in \mathbb{N}_0 \quad (3.3)$$

$$C = \{x \in \mathbb{R} \mid V(x) < E\} \quad (3.4)$$

$$\int_C \sqrt{2m(E - V(x))} dx = 2\pi(n + 1/2) \quad (3.5)$$

this condition does not (in most cases) give the exact energy levels (Hall, 2013). It can be interpreted such that the oscillating part of the wave function has to complete all half oscillation.

To solve this problem for an arbitrary potential in a computer the set C and the fact that n has to be a non negative integer is not really helpful, but the condition can be rewritten to

$$p(x) = \begin{cases} \sqrt{2m(E - V(x))} & V(x) < E \\ 0 & \text{else} \end{cases} \quad (3.6)$$

$$\frac{1}{2\pi} \int_{-\infty}^{\infty} p(x) dx - \frac{1}{2} \mod 1 = 0 \quad (3.7)$$

Unfortunately 3.7 is not continuous which means that Newtons method can't be applied. Further on the bounds of integration have to be finite, this means the user of the program will have to specify a value for the constant APPROX_INF where any value for x out side of that range should satisfy $V(x) > E$. But it shouldn't be too big since the integrate function can only evaluate a relatively small number (default 64000) of trapezes before the performance will suffer enormously. The default value for APPROX_INF is $(-200.0, 200.0)$.

The implementation is quite strait forward we evaluate 3.7 for a number of energies and then check for discontinuities.

```
1 pub fn nth_energy<F: Fn(f64) -> f64 + Sync>(n: usize, mass: f64, pot: &F, view: (f64, f64)) -> f64 {
2     const ENERGY_STEP: f64 = 10.0;
3     const CHECKS_PER_ENERGY_STEP: usize = INTEG_STEPS;
4     let sommerfeld_cond = SommerfeldCond { mass, pot, view };
5
6     let mut energy = 0.0;
7     let mut i = 0;
8
9     loop {
10         let vals = evaluate_function_between(
11             &sommerfeld_cond,
```

```

12         energy,
13         energy + ENERGY_STEP,
14         CHECKS_PER_ENERGY_STEP,
15     );
16     let mut int_solutions = vals
17         .iter()
18         .zip(vals.iter().skip(1))
19         .collect::<Vec<&Point<f64, f64>, &Point<f64, f64>>>();
20     .par_iter()
21     .filter(|(p1, p2)| (p1.y - p2.y).abs() > 0.5 || p1.y.signum() != p2.y.signum())
22     .map(|ps| ps.1)
23     .collect::<Vec<&Point<f64, f64>>>();
24     int_solutions.sort_by(|p1, p2| cmp_f64(&p1.x, &p2.x));
25     if i + int_solutions.len() > n {
26         return int_solutions[n - i].x;
27     }
28     energy += ENERGY_STEP - (ENERGY_STEP / (CHECKS_PER_ENERGY_STEP as f64 + 1.0));
29     i += int_solutions.len();
30 }
31 }

```

First we check over the interval $(0.0, \text{ENERGY_STEP})$ if there are not enough zeros we check the next interval of energies and so on until we found n zeros. It's also possible that 3.7 is negative before the 0th energy there for we also have to check for normal zeros by comparing the signs of the values.

The struct `SommerfeldCond` is a `Func<f64, f64>` that evaluates 3.7.

3.6.1. Accuracy

For a benchmark we will use

$$\begin{aligned}
 m &= 1 \\
 V(x) &= x^2 \\
 (-\infty, \infty) &\approx (-200, 200).
 \end{aligned}$$

To get the actual values we will use Wolfram Language with WolframScript a programming language similar to Wolframalpha that can calculate the integral analytically. In Rust we can rewrite main to

```

1 fn main() {
2     let output_dir = Path::new("output");
3
4     let values = (0..=50)
5         .into_iter()
6         .map(|n: usize| Point::<usize, f64> {
7             x: n,
8             y: energy::nth_energy(n, 1.0, &potentials::square, APPROX_INF),
9         })

```

```

10         .collect::

```

This will output all energy levels from $n = 0$ to $n = 50$. We can implement the same thing in WolframScript

```

1  m = 1
2  V[x_] = x^2
3
4  nthEnergy[n_] = Module[{energys, energy},
5      sommerfeldIntegral[en_] = Integrate[Sqrt[2*m*(en - V[x])],
6                                          {x, -Sqrt[en], Sqrt[en]}]
7      energys = Solve[sommerfeldIntegral[en] == 2*Pi*(n + 1/2), en] // N;
8      energy = en /. energys[[1]];
9      energy
10 ]
11
12 energys = Table[{n, N@nthEnergy[n]}, {n, 0, 50}]
13
14 csv = ExportString[energys, "CSV"]
15 csv = StringReplace[csv, "," -> " "]
16 Export["output/energies_exact.dat", csv]

```

These programs will output two files `energy.txt` (Appendix C.1) for our implementation in Rust and `energies_exact.dat` (Appendix C.1) for WolframScript. As a ruff estimate we would expect an error of $\pm \frac{10}{64000} \approx \pm 1.56 * 10^{-4}$, because the program checks for energies with that step size.

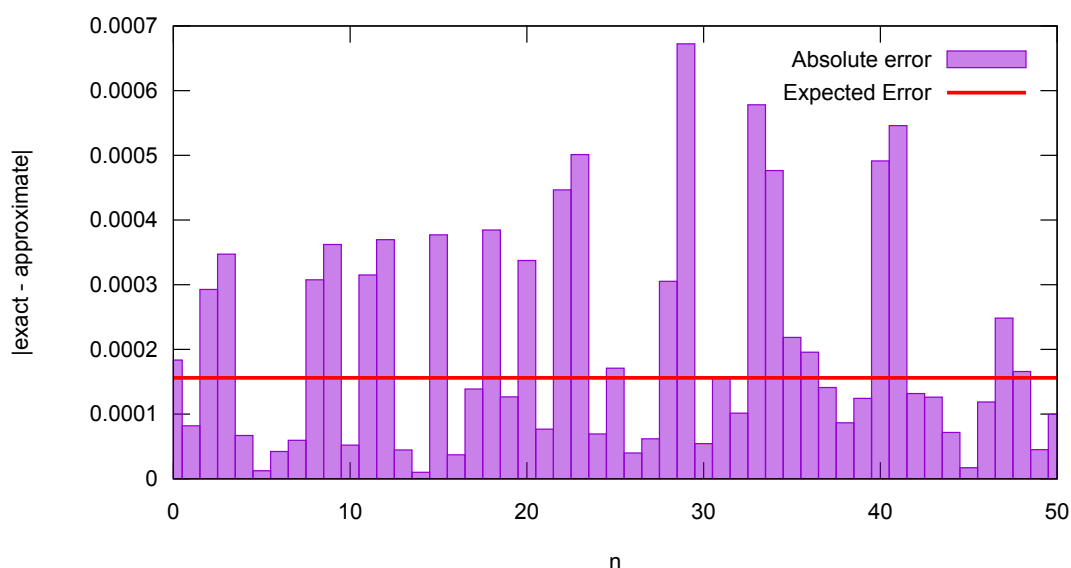


Figure 3.7.: Absolute error of energy levels in square potential

When we plot the absolute error we get figure 3.6.1. The error is a little higher than expected which is probably due to errors in the integral. Still the algorithm should be precise enough. If you'd like you could pick a lower value for `ENERGY_STEP` in `src/energy.rs:49`, but this will impact the performance for calculating energies with higher numbers for n .

3.7. Approximation Scheme

There are mainly three approximation methods used to solve for the actual wave function itself. There is perturbation theory which breaks the problem down in to ever smaller sub-problems that then can be solved exactly. This can be achieved by adding something to the Hamiltonian operator \hat{H} which can then be solved exactly. But *perturbation theory is inefficient compared to other approximation methods when calculated on a computer* (Van Mourik et al., 2014, Introduction).

The second is Density functional field theory, it has evolved over the years and is used heavily in chemistry to calculate properties of molecules and is also applicable for the time dependent Schrödinger equation. It is something that might be interesting to add to the program in the future.

The program uses the third method WKB approximation, it is applicable to a wide verity of linear differential equations and works very well in the case of the Schrödinger equation. Originally it was developed by Wentzel, Kramers and Brillouin in 1926. It gives an approximation to the eigenfunctions of the Hamiltonian \hat{H} in one dimension. The approximation is best understood as applying to a fixed range of energies as \hbar tends to zero (Hall, 2013,

p. 305).

WKB splits $\Psi(x)$ into tree parts that can be connected to form the full solution. The tree parts are described as

$$p(x) = \sqrt{2m(|E - V(x)|)} \quad (3.8)$$

$$V(t) - E = 0 \quad (3.9)$$

$$\psi_{exp}^{WKB}(x) = e^{\delta i} \frac{c_1}{2\sqrt{p(x)}} \exp\left(-\left|\int_x^t p(y)dy\right|\right) \quad (3.10)$$

$$\psi_{osz}^{WKB}(x) = \frac{c_1}{\sqrt{p(x)}} \exp\left(-\left|\int_x^t p(y)dy\right| i + \delta i\right) \quad (3.11)$$

$$u_1 = -2m \frac{dV}{dx}(t) \quad (3.12)$$

$$\psi^{Airy}(x) = e^{(t-x+\delta)i} \frac{c_1 \sqrt{\pi}}{\sqrt[6]{u_1}} \text{Ai}\left(\sqrt[3]{u_1}(t-x)\right). \quad (3.13)$$

Since equation 3.9 might have more than one solution for turning points t , we have to consider each one of them individually and in the end join them into one function.

In figure 3.7 the three parts are visualized. The purple section on the left is the exponential decaying part $\psi_{exp}^{WKB}(x)$, equation 3.10 is a modified version of the original version as described in (Hall, 2013, p. 317, eq. 15.25) where b and a are different solutions for t of equation 3.9. The absolute symbol makes it possible to not differentiate between the case where $x < t$ and $x > t$. Further on a factor of $e^{\delta i}$ was added such that the imaginary part of $\psi_{exp}^{WKB}(x)$ is the same as in $\psi_{osz}^{WKB}(x)$.

The blue part on the right is $\psi_{osz}^{WKB}(x)$. Again equation 3.11 was expanded to result in the more general complex solution and it also works for both ψ_1 and ψ_2 in (Hall, 2013, p. 316-317, Claim 15.7). Hall (2013) assumes that $\delta = \pi/4$ which doesn't work in the simple case of $V(x) = x^2$, in figure 3.7 $\delta = 0$ was used. This will be further discussed in section ??.

3.7.1. Example

This example is from (Wikipedia, 2022b, An example).

Lets solve the ordinary differential equation

$$\epsilon^2 \frac{d^2 y}{dx^2} = Q(x)y$$

where $Q(x)$ is an arbitrary function that is not $Q(x) = 0$ and ϵ is small. Note that this example relates to the Schrödinger equation where $\epsilon = -\frac{\hbar^2}{2m}$ and $Q(x) = E - V(x)$.

Replace y with

$$y(x) = \exp\left(\frac{1}{\delta} \sum_{n=0}^{\infty} \delta^n S_n(x)\right)$$

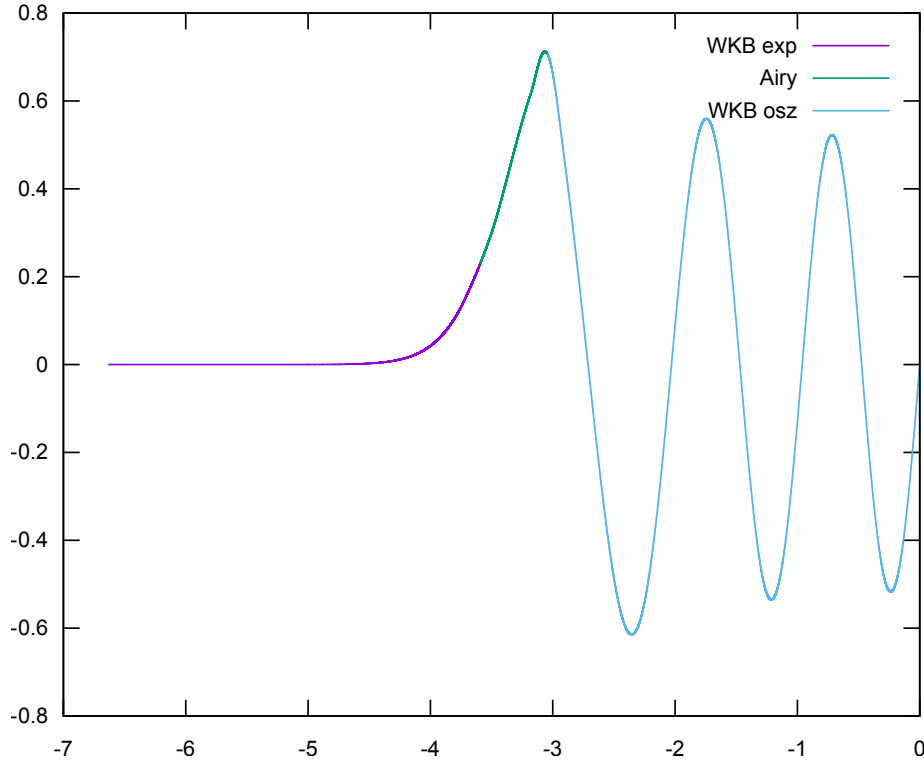


Figure 3.8.: Left half of wave function with $N_{Energy} = 5 \Rightarrow E = 11.0$, $m = 2$, $V(x) = x^2$

resulting in the equation

$$\epsilon^2 \frac{1}{\delta^2} \left(\sum_{n=0}^{\infty} \delta^n S'_n(x) + \frac{1}{\delta} \left(\sum_{n=0}^{\infty} \delta^n S''_n(x) \right) \right) = Q(x) \quad (3.14)$$

As an approximation set the upper bound of the sum to be 2 instead of ∞ , then 3.14 can be written as

$$\frac{\epsilon^2}{\delta^2} S_0'^2 + \frac{2\epsilon^2}{\delta} S_0' S_1' + \frac{\epsilon^2}{\delta} S_0'' = Q(x) \quad (3.15)$$

If we go further and solve for S_0 and S_1 we will get that

$$S_0(x) = \pm \int_{x_0}^x \sqrt{Q(t)} dt \quad (3.16)$$

$$S_1(x) = -\frac{1}{4} \ln Q(x) + k_1 \quad (3.17)$$

this means that our approximate solution of y is

$$y(x) \approx c_1 Q^{-\frac{1}{4}}(x) \exp\left(\frac{1}{\epsilon} \int_{x_0}^x \sqrt{Q(t)} dt\right) + c_2 Q^{-\frac{1}{4}}(x) \exp\left(-\frac{1}{\epsilon} \int_{x_0}^x \sqrt{Q(t)} dt\right)$$

here c_1, c_2 depend on the boundary conditions and can also be used to renormalize the function. x_0 will later be discussed in section (...). It is important to have the right value for x_0 which depends on $V(x)$ and E although I have not yet managed to find out the connection.

3.7.2. Validity

Dieser abschnitt ist momentan nur fürs Probe Kapitel muss ich noch besser nach Rechnen und genau verstehen

The approximation that was made in equation 3.14 to 3.15 that we ignored the terms of $n > 2$ and assumed that they are small.

$$\epsilon \left| \frac{dQ}{dx} \right| \ll Q(x)^2$$

Without the assumption that \hbar is in fact *small* the equation can also be interpreted such that the potential $V(x)$ has to vary slowly compared to $(V(x) - E)^2$.

During calculations I noticed that it is crucial what values x_0 has since some times the results can not be valid.

A. Source Code

At the moment the code is only available on <https://github.com/Gian-Laager/Schroedinger-Approximation>

B. Detailed Calculations

B.1. Proofs

B.1.1. Smoothness of Transitionfunction

Given that

$$f : \mathbb{R} \rightarrow \mathbb{C} \quad (\text{B.1})$$

$$g : \mathbb{R} \rightarrow \mathbb{C} \quad (\text{B.2})$$

$$\{f, g\} \in C^1 \quad (\text{B.3})$$

$$\{\alpha, \delta\} \in \mathbb{C} \quad (\text{B.4})$$

define (Hall, 2013)

$$\chi(x) = \sin^2\left(\frac{\pi(x - \alpha)}{2\delta}\right) \quad (\text{B.5})$$

$$(f \sqcup g)(x) = f(x) + (g(x) - f(x))\chi(x) \quad (\text{B.6})$$

and proof that

$$\frac{d(f \sqcup g)}{dx}(\alpha) = \frac{df}{dx}(\alpha) \quad (\text{B.7})$$

$$\frac{d(f \sqcup g)}{dx}(\alpha + \delta) = \frac{dg}{dx}(\alpha + \delta). \quad (\text{B.8})$$

Calculate derivatives

$$\frac{d\chi}{dx}(x) = \frac{\pi}{2\delta} \sin\left(\frac{\pi(x - \alpha)}{\delta}\right) \quad (\text{B.9})$$

$$\frac{d(f \sqcup g)}{dx}(x) = \frac{df}{dx}(x) + \left(\frac{dg}{dx}(x) - \frac{df}{dx}(x)\right)\chi(x) + (g(x) - f(x))\frac{d\chi}{dx}(x). \quad (\text{B.10})$$

Note that

$$\frac{d\chi}{dx}(\alpha) = 0 \quad (\text{B.11})$$

$$\chi(\alpha) = 0 \quad (\text{B.12})$$

$$\frac{d\chi}{dx}(\alpha + \delta) = 0 \quad (\text{B.13})$$

$$\chi(\alpha + \delta) = 1 \quad (\text{B.14})$$

therefor

$$\frac{d(f \sqcup g)}{dx}(\alpha) = \frac{df}{dx}(\alpha) + 0 \left(\frac{dg}{dx}(\alpha) - \frac{df}{dx}(\alpha) \right) + 0(g(x) - f(x)) = \frac{df}{dx}(\alpha) \quad (\text{B.15})$$

and

$$\frac{d(f \sqcup g)}{dx}(\alpha + \delta) = \frac{df}{dx}(\alpha + \delta) + 1 \left(\frac{dg}{dx}(\alpha + \delta) - \frac{df}{dx}(\alpha + \delta) \right) + 0(g(x) - f(x)) \quad (\text{B.16})$$

$$\frac{d(f \sqcup g)}{dx}(\alpha + \delta) = \frac{df}{dx}(\alpha + \delta) + \frac{dg}{dx}(\alpha + \delta) - \frac{df}{dx}(\alpha + \delta) = \frac{dg}{dx}(\alpha + \delta) \blacksquare. \quad (\text{B.17})$$

C. Data Files

C.1. Energies

energy.txt

```
0 1.4143970999546869
1 4.2427225425397275
2 7.071360490007656
3 9.89984218503414
4 12.727855127619105
5 15.55633682264559
6 18.384818517672073
7 21.213143965139928
8 24.041938165049302
9 26.870419860075785
10 29.69843279777794
11 32.52722700257012
12 35.35570869759661
13 38.18372163529877
14 41.012203335208056
15 43.84099753511743
16 46.66901047281958
17 49.49733591540462
18 52.32628637263825
19 55.15445556278185
20 57.98309351024977
21 60.811106452834736
22 63.64005690518555
23 66.46853860021204
24 69.29639528547274
25 72.1247207329406
26 74.95335868040853
27 77.78168412299357
28 80.61047832778574
29 83.43927252769512
30 86.26697296051438
31 89.09561090798232
32 91.92378010300872
```

energies_exact.dat

```
0 1.4142135623730951
1 4.242640687119286
2 7.0710678118654755
3 9.899494936611665
4 12.727922061357857
5 15.556349186104047
6 18.38477631085024
7 21.213203435596427
8 24.041630560342618
9 26.870057685088806
10 29.698484809834998
11 32.526911934581186
12 35.35533905932738
13 38.18376618407357
14 41.01219330881976
15 43.84062043356595
16 46.66904755831214
17 49.49747468305833
18 52.32590180780452
19 55.15432893255071
20 57.9827560572969
21 60.81118318204309
22 63.63961030678928
23 66.46803743153548
24 69.29646455628166
25 72.12489168102785
26 74.95331880577405
27 77.78174593052023
28 80.61017305526643
29 83.43860018001261
30 86.2670273047588
31 89.095454429505
32 91.92388155425118
```

33 94.75288680780098
34 97.58121225038602
35 100.40938144541242
36 103.23739438311458
37 106.06587607814106
38 108.89435777316754
39 111.72299572551829
40 114.55178992542766
41 117.38027162045414
42 120.2082845630391
43 123.0364537531827
44 125.86493544820918
45 128.69341714323565
46 131.52174259070352
47 134.35053679061292
48 137.17854972831506
49 140.0071876806658
50 142.83566937569228

33 94.75230867899738
34 97.58073580374356
35 100.40916292848975
36 103.23759005323595
37 106.06601717798213
38 108.89444430272833
39 111.72287142747452
40 114.5512985522207
41 117.3797256769669
42 120.20815280171308
43 123.03657992645928
44 125.86500705120547
45 128.69343417595167
46 131.52186130069785
47 134.35028842544403
48 137.17871555019022
49 140.00714267493643
50 142.83556979968262

Bildquellen

Wo nicht anders angegeben, sind die Bilder aus dieser Arbeit selbst erstellt worden.

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Selbständigkeitserklärung

Hiermit bestätige ich, Gian Laager, meine Maturaarbeit selbständig verfasst und alle Quellen angegeben zu haben.

Ich nehme zur Kenntnis, dass meine Arbeit zur Überprüfung der korrekten und vollständigen Angabe der Quellen mit Hilfe einer Software (Plagiaterkennungstool) geprüft wird. Zu meinem eigenen Schutz wird die Software auch dazu verwendet, später eingereichte Arbeiten mit meiner Arbeit elektronisch zu vergleichen und damit Abschriften und eine Verletzung meines Urheberrechts zu verhindern. Falls Verdacht besteht, dass mein Urheberrecht verletzt wurde, erkläre ich mich damit einverstanden, dass die Schulleitung meine Arbeit zu Prüfzwecken herausgibt.

Ort

Datum

Unterschrift