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## Solving the 1D Schrödinger Equation Numerically with Irregular Grid

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### Abstract

There is a great interest for solving the Schrödinger equation, and here is presented an irregular finite difference scheme to do so. This enables to resolve small details in a larger space with logarithmic grids for instance. Both time-dependent and time-independent Schrödinger equation are formulated with this scheme. Non-zero boundary conditions are also studied for the eigenvalue problem, how changing one boundary affects the other boundary instantly due to normalization.

### Introduction

The matrix formulation of quantum mechanics can be used to solve the Schrödinger equation computationally, and with the  $\theta$ -rule, Sec. 8, different schemes can be chosen like the Crank–Nicolson method [1,2]. Sec.4 present an alternative irregular finite-difference scheme, derived by averaging the two-sided limit forms of the derivative, which differs from the standard Taylor-expansion-based non-uniform schemes. New in this paper is matrix formulation, Sec. 9, with irregular grid in the finite difference representation Sec. 4, which enables the use of logarithmic grid to resolve small details Sec. 5. This matrix formulation requires the grid resolution to be included in the state vector, Sec. 6, and Sec. 9 shows that Hamiltonian is self-adjoint [3]. The paper further investigate eigenvalue problem in Sec. 7 with non-zero boundaries and see how changes in one end changes instantly the boundary at the other end, which is due to normalization property of the wave function. A Mathematica script is including in Sec. 11 to show how the irregular grid is implemented.

### Related Work

Finite-difference methods on non-uniform meshes have a long history in numerical analysis. The most widely used three-point non-uniform stencil is obtained by Taylor expansion about the grid point, yielding coefficients that depend asymmetrically on the forward and backward spacings. General formulas for such coefficients on arbitrarily spaced grids were given by Fornberg in his seminal paper on generating finite-difference weights [4]. Earlier, Lyszka and Orkisz developed what is now called the generalized finite difference method, where local derivative approximations are constructed from scattered node distributions [5]. These schemes, and their modern extensions, have been widely applied to partial differential equations on irregular domains. More recent work has focused on compact or self-adjoint difference operators on stretched meshes to preserve conservation or energy properties in time-dependent problems [6].

In contrast to these standard formulations, the irregular finite difference presented here is derived by averaging the two limit definitions of the derivative from the left and the right. This yields a symmetric three-point expression for the second derivative whose coefficients differ algebraically from the standard Taylor-based scheme, while still collapsing to the familiar central difference in the uniform-mesh limit. To our knowledge, this averaged stencil has not been documented in the numerical analysis literature, and its use in discretizing the Schrödinger operator allows for a naturally Hermitian matrix formulation on logarithmic grids.



$$\frac{d^2\Psi_i}{dx^2} \approx \frac{2}{x_{i+1} - x_{i-1}} \left( \frac{\Psi_{i+1} - \Psi_i}{x_{i+1} - x_i} - \frac{\Psi_i - \Psi_{i-1}}{x_i - x_{i-1}} \right) \quad (10)$$

### Logarithmic Grid

To make an logarithmic grid the step length need to increase exponentially, so a grid function  $g(x)$  that starts at  $x_{\min}$  would need to look like

$$g(x) = x_{\min} a^{x-x_{\min}} \quad (11)$$

where the base  $a$  is determined by the end point  $x_{\max}$  of the grid

$$g(x_{\max}) = x_{\max} = x_{\min} a^{x_{\max}-x_{\min}} \quad (12)$$

which yields the grid function

$$g(x) = x_{\min} \left( \frac{x_{\max}}{x_{\min}} \right)^{\frac{x-x_{\min}}{x_{\max}-x_{\min}}} \quad (13)$$

The discrete grid point  $i$  of a total of  $n$  points is then calculated as follows

$$g_i = x_{\min} \left( \frac{x_{\max}}{x_{\min}} \right)^{\frac{i}{n}} = x_{\min}^{1-\frac{i}{n}} x_{\max}^{\frac{i}{n}} \quad (14)$$

### Discretizing Wave Functions

To discretize an wave functions we must use the properties of linear transformation, which is additivity, homogeneity of degree 1 and commutation relation

$$H(\psi_1(x) + \psi_2(x)) = H(\psi_1(x)) + H(\psi_2(x)) \quad (15)$$

$$H(a\psi(x)) = aH(\psi(x)) \quad (16)$$

$$H(\psi(x) x') = H(\psi(x)) x' \quad (17)$$

where  $H$  is a linear transformation,  $\psi$  a function and  $a$  a constant. Lets define

$$\vec{\psi} = \sum_i \psi_i \Delta\chi_i \quad (18)$$

which approaches an integral when  $\Delta\chi_i \rightarrow 0$ .  $\vec{\psi}$  can be made to have the same eigenvalue as  $\psi(x)$  with the operator  $H$ , such that

$$H(\psi_i) \Delta\chi_i = H(\psi_i \Delta\chi_i) \quad (19)$$

where  $H(\psi_i) = E \psi_i$  implies  $H(\psi_i \Delta\chi_i) = E \psi_i \Delta\chi_i$ . Using additivity in (15)

$$\sum_i H(\psi_i) \Delta\chi_i = \sum_i H(\psi_i \Delta\chi_i) = H\left(\sum_i \psi_i \Delta\chi_i\right) \quad (20)$$

This results in the following equality for a linear combination

$$H\left(\sum_{k=1}^n a_k \psi_k(x)\right) = \sum_{k=1}^n a_k H(\psi_k(x)) \quad (21)$$

Using the fact that a vector can be written as a linear combination, means that we can use an integral to represent an infinite dimensional vector

$$\vec{\psi}(x) = \int^x \psi(x) dx \quad (22)$$

Now let the  $dx$  be a non-constant step length which we want to change to a constant step length  $dy$ , then the linear transformation of the above vector can be rewritten to



The inverse iteration method can be used to find an approximate eigenvector  $\vec{\psi}$  to a given eigenvalue  $E$ . We introduce an approximate eigenvalue  $\alpha E$  where we choose  $\alpha$  sufficiently close to 1 such that the inverse iteration method converges to the approximate eigenvector to  $E$ , but  $\alpha$  should also be chosen sufficiently far away from 1 such that the magnitude of  $E - \alpha E$  is sufficiently bigger than 0 to avoid numerical issues in the inverse iteration method. The foundation of the inverse iteration method is given by

$$\mathbf{A} \vec{\psi} = E \vec{\psi} \quad \Leftrightarrow \quad (\mathbf{A} - \alpha E \mathbf{I}) \vec{\psi} = (1 - \alpha) E \vec{\psi} \quad (31)$$

Using the power iteration method we arrive at the inverse iterative method

$$\vec{\psi}_{k+1} = (1 - \alpha) E (\mathbf{A} - \alpha E \mathbf{I})^{-1} \vec{\psi}_k = (1 - \alpha)^k E^k (\mathbf{A} - \alpha E \mathbf{I})^{-k} \vec{\psi}_1 \quad (32)$$

Let  $\vec{\psi}_1 = c \vec{\psi}_0$ , then the norm of  $\vec{\psi}_{k+1}$  are given by

$$\|\vec{\psi}_{k+1}\| = c (1 - \alpha)^k E^k \|(\mathbf{A} - \alpha E \mathbf{I})^{-k} \vec{\psi}_0\| \quad (33)$$

which implies that  $\vec{\psi}_{k+1}$  is normalizable when

$$c = \frac{1}{(1 - \alpha)^k E^k \|(\mathbf{A} - \alpha E \mathbf{I})^{-k} \vec{\psi}_0\|} \quad (34)$$

and we get

$$\vec{\psi}_{k+1} = \frac{(\mathbf{A} - \alpha E \mathbf{I})^{-k} \vec{\psi}_0}{\|(\mathbf{A} - \alpha E \mathbf{I})^{-k} \vec{\psi}_0\|} = \frac{\mathbf{B} \vec{\psi}_0}{\|\mathbf{B} \vec{\psi}_0\|} \quad (35)$$

where  $\mathbf{B} = (1 - \alpha)^k E^k (\mathbf{A} - \alpha E \mathbf{I})^{-k}$ . Following the form of  $\mathbf{A}$  in (26) we have the following form of  $\mathbf{B}$

$$\mathbf{B} = \begin{bmatrix} b_{00} & b_{01} & \cdots & b_{0n} & b_{0(n+1)} \\ b_{10} & b_{11} & \cdots & b_{1n} & b_{1(n+1)} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ b_{n0} & b_{n1} & \cdots & b_{nn} & b_{n(n+1)} \\ b_{(n+1)0} & b_{(n+1)1} & \cdots & b_{(n+1)n} & b_{(n+1)(n+1)} \end{bmatrix} = \begin{bmatrix} 1 & 0 & \cdots & 0 & 0 \\ b_{10} & b_{11} & \cdots & b_{1n} & b_{1(n+1)} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ b_{n0} & b_{n1} & \cdots & b_{nn} & b_{n(n+1)} \\ 0 & 0 & \cdots & 0 & 1 \end{bmatrix} \quad (36)$$

Note the simplification of row 1 and  $n + 1$ , which is due to the fact that  $\mathbf{A} - \alpha E \mathbf{I}$  has only one non-zero element " $(1 - \alpha) E$ " in each of the rows 1 and  $n + 1$ . Let  $\vec{\psi}_{k+1} = [\psi_0^{k+1}, \psi_1^{k+1}, \dots, \psi_{n+1}^{k+1}]$  and  $\vec{\psi}_0 = [a, 0, \dots, 0, b]$  where  $a$  and  $b$  is non-zero, then we can use (35) to setup

$$\|\mathbf{B} \vec{\psi}_0\| = \frac{a}{\psi_0^{k+1}} = \frac{b}{\psi_{n+1}^{k+1}} \quad (37)$$

Expanding the expression for the norm  $\|\mathbf{B} \vec{\psi}_0\|$ , we can then setup

$$\sqrt{\sum_{j=0}^{n+1} (b_{j0} a + b_{j(n+1)} b)^2} = \frac{a}{\psi_0^{k+1}} \quad (38)$$

and using the relation between  $a$  and  $b$  in (37) we get that

$$\sum_{j=0}^{n+1} (b_{j0} \psi_0^{k+1} + b_{j(n+1)} \psi_{n+1}^{k+1})^2 = 1 \quad (39)$$

and solving this equation with regard to  $\psi_{n+1}^{k+1}$  yields

$$\psi_{n+1}^{k+1} = \frac{-\psi_0^{k+1} \sum_{j=1}^n b_{j0} b_{j(n+1)} \pm \sqrt{\left(\psi_0^{k+1} \sum_{j=1}^n b_{j0} b_{j(n+1)}\right)^2 - \sum_{j=1}^{n+1} b_{j(n+1)}^2 \left(\left(\psi_0^{k+1}\right)^2 \sum_{j=0}^n b_{j0}^2 - 1\right)}}{\sum_{j=1}^{n+1} b_{j(n+1)}^2} \quad (40)$$

or with regard to

$$\psi_0^{k+1} = \frac{-\psi_{n+1}^{k+1} \sum_{j=1}^n b_{j0} b_{j(n+1)} \pm \sqrt{\left(\psi_{n+1}^{k+1} \sum_{j=1}^n b_{j0} b_{j(n+1)}\right)^2 - \sum_{j=0}^n b_{j0}^2 \left(\left(\psi_{n+1}^{k+1}\right)^2 \sum_{j=1}^{n+1} b_{j(n+1)}^2 - 1\right)}}{\sum_{j=0}^n b_{j0}^2} \quad (41)$$

When  $k \rightarrow \infty$  these converge to the boundary conditions  $\psi_0 \Delta x_0 = \lim_{k \rightarrow \infty} \psi_0^k$  and  $\psi_{n+1} \Delta x_{n+1} = \lim_{k \rightarrow \infty} \psi_{n+1}^k$ . As we see in (40) and (41) the two boundary conditions are dependent on each other. We also see that given one of the boundary conditions then there are two solutions that is possible for the other boundary condition, making two path ways possible for transition between boundary conditions of the wave function.

### The $\theta$ -Rule and the Time-Dependent Schrödinger Equation

Assuming

$$t_\theta \approx \theta t + (1 - \theta) t_0 \quad (42)$$

then the first order approximation can be written

$$\begin{aligned} \Psi(x, t_\theta) &\approx \Psi(x, t_0) + \frac{\partial \Psi(x, t_0)}{\partial t} (t_\theta - t_0) = \Psi(x, t_0) + \theta \frac{\partial \Psi(x, t_0)}{\partial t} (t - t_0) \\ &\approx \Psi(x, t_0) + \theta (t - t_0) \frac{\Psi(x, t) - \Psi(x, t_0)}{t - t_0} = \theta \Psi(x, t) + (1 - \theta) \Psi(x, t_0) \end{aligned} \quad (43)$$

which is known as the  $\theta$ -rule. The  $\theta$ -rule can be used to approximate the time-dependent Schrödinger equation in (2)

$$\begin{aligned} i \hbar \frac{\Psi(x, t) - \Psi(x, t_0)}{t - t_0} &= i \hbar \frac{\Psi(x, t_\theta) - \Psi(x, t_0)}{t_\theta - t_0} \approx H(\Psi(x, t_\theta)) \\ &= H(\theta \Psi(x, t) + (1 - \theta) \Psi(x, t_0)) + F(\Psi(x, t_\theta)) = \theta H(\Psi(x, t)) + (1 - \theta) H(\Psi(x, t_0)) \end{aligned} \quad (44)$$

where  $\theta = 0$  is Forward Euler scheme (explicit),  $\theta = 1$  is Backward Euler scheme (implicit), and  $\theta = 1/2$  is Crank-Nicolson scheme. Crank-Nicolson have the benefit of second order local truncation error. Therefore the numerical scheme for the  $\theta$ -rule can be approximated

$$i \hbar \frac{\Psi(x, t) - \Psi(x, t_0)}{t - t_0} \approx \theta H(\Psi(x, t)) + (1 - \theta) H(\Psi(x, t_0)) \quad (45)$$

### Time-Dependent Schrödinger Equation and Irregular Finite Difference

Combining the  $\theta$ -rule of the time-dependent Schrödinger equation in (45) and the irregular finite difference of the second derivative in (6), yields the following difference equation with  $i$ -indices are spacial steps and  $j$ -indices are time steps

$$\alpha_j (\Psi_{i,j+1} - \Psi_{i,j}) = \theta \sum_{k=-1}^1 h_{i,k} \Psi_{i+k,j+1} + (1 - \theta) \sum_{k=-1}^1 h_{i,k} \Psi_{i+k,j} \quad (46)$$

where

$$\alpha_j = \frac{i \hbar}{t_{j+1} - t_j}, \quad h_{i,\pm 1} = -\frac{\hbar^2 d_{i,0} d_{i,\pm 1}}{2m}, \quad h_{i,0} = -\frac{\hbar^2 d_{i,0}^2}{2m} + V_i \quad (47)$$

and the  $h_{i,k}$  are elements in the Hamilton operator

$$\mathbf{H} = -\frac{\hbar^2}{2m} \mathbf{D}_2 + \mathbf{V} \quad (48)$$

The equation in (46) can be written as a matrix equation with all the spatial step, where  $\Psi_{j+1}$  is the only unknown giving the wave function in the next time step

$$(\theta \mathbf{H} - \alpha_j \mathbf{I}) \Psi_{j+1} = ((\theta - 1) \mathbf{H} - \alpha_j \mathbf{I}) \Psi_j \quad (49)$$

### Hermitian/Self-Adjoint

Looking at the inner product

$$\langle \mathbf{u}, \mathbf{v} \rangle = \sum_i u_i^* v_i \quad (50)$$

we can show that the discretization used in this paper is self-adjoint

$$\begin{aligned} \langle \mathbf{u}, \mathbf{H} \mathbf{v} \rangle &= \sum_{i,j} u_j^* H_{j,i} v_i = \sum_i (h_{i,0} u_i^* v_i + h_{i,-1} u_{i-1}^* v_i + h_{i,1} u_{i+1}^* v_i) = \sum_i (h_{i,0} u_i^* + h_{i,1} (u_{i-1}^* + u_{i+1}^*)) v_i \\ &= \sum_i (h_{i,0}^* u_i^* + h_{i,1}^* (u_{i-1}^* + u_{i+1}^*)) v_i = \sum_i (h_{i,0}^* u_i^* v_i + h_{i,1}^* u_{i-1}^* v_i + h_{i,-1}^* u_{i+1}^* v_i) = \sum_{i,j} H_{i,j}^* u_j^* v_i = \langle \mathbf{H} \mathbf{u}, \mathbf{v} \rangle \end{aligned} \quad (51)$$

which shows that the eigenvalues are real and the corresponding eigenvectors are orthogonal.

### Mathematica Code

**Algorithm 1** Mathematica code to solve Schrödinger equation for harmonic oscillator with irregular grid where  $\theta = 1$

```

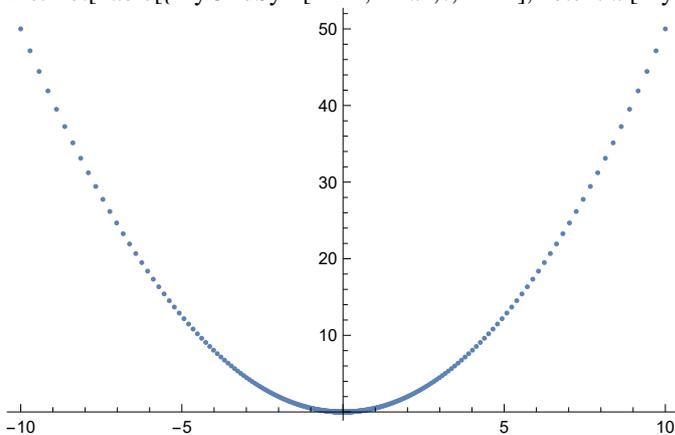
1: Potential[x_]:=m * x * x/2
2: UniformGrid[min_, max_, i_, n_]:=min +(max - min) * i/n
3: LogGrid[min_, max_, i_, n_]:=If[min * max >= 0, min *(max / min)^(i/n), 2*min - min *((min - max)/ min)^(i/n)]
4: CoreGrid[min_, mid_, max_, i_, m_, n_]:=
5:   If[Abs[i]<=m,
6:     UniformGrid[min, mid, i, m],
7:     If[i > 0,
8:       LogGrid[mid, max, i - m, n - m],
9:       2 * min -LogGrid[mid, max, -i - m, n - m]
10:    ]
11: ]
12: Tri[n_, min_, max_, a_, V_, G_]:=Block[
13:   {i, A = SparseArray[{Band[{1, 2}] -> 0, Band[{1, 1}] -> 0, Band[{2, 1}] -> 0}, n],
14:   x0 = Apply[G, {min, max, 0, n + 1}], x1 = Apply[G, {min, max, 1, n + 1}], x2, gm1, g0, g1, b},
15:   For[i = 1, i <= n, i++,
16:     x2 = Apply[G, {min, max, i + 1, n + 1}];
17:     gm1 = 1/(x1 - x0);
18:     g1 = 1/(x2 - x1);
19:     g0 = gm1 * g1(x2 - x0);
20:     b = a * g0/2;
21:     If[i > 1, A[[i - 1, i]] = gm1 * b];
22:     A[[i, i]] = -g0 * b + Apply[V, {x1}];
23:     If[i < n, A[[i + 1, i]] = g1 * b];
24:     x0 = x1;
25:     x1 = x2;
26:   ];
27:   A
28: ]
29: EigenFunction[vec_, min_, max_, G_, order_]:=Block[
30:   {fsum, sum, v, vv = Block[{nmax = Length[vec] + 1},
31:     Block[{x0, x1 = Apply[G, {min, max, 0, nmax}], x2 = Apply[G, {min, max, 1, nmax}],
32:       Join[
33:         {{min, 0}},
34:         Table[x0 = x1; x1 = x2; x2 = Apply[G, {min, max, i + 1, nmax}]; {x1, vec[[i]] * (x2 - x0)/((x2 - x1) *
(x1 - x0))}, {i, 1, nmax - 1}],
35:         {{max, 0}}
36:       ]
37:     ]}
38:   }
39:   fsum = Interpolation[vv, InterpolationOrder -> order];
40:   sum = Sqrt[NIntegrate[Conjugate[fsum[x]] * fsum[x], {x, min, max}]];
41:   v = Table[{vv[[i]][[1]], vv[[i]][[2]]/sum}, {i, 1, Length[vv]}];
42:   Interpolation[v, InterpolationOrder -> order]
43: ]
44: TimeSolve[c_, V_, psi_, xmin_, xmax_, X_, Xsave_, tmin_, tmax_, nt_, T_, Tsave_, order_]:=
45:   Block[{nx = Length[psi] - 2}, Block[
46:     {B = Tri[nx, xmin, xmax, c, V, X], x = Table[Apply[X, {xmin, xmax, i, nx + 1}], {i, 0, nx + 1}],

```

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47:     t0 = Apply[T, {tmin, tmax, 0, nt}], t1,  $\psi_0 = \psi$ , fsum},
48:     Block[{j, k = 2, b,  $\psi_1$ , w = Join[{1}, Table[(x[[i + 1]] - x[[i - 1]])/(2 * (x[[i + 1]] - x[[i]]) * (x[[i]] - x[[i - 1]])), {i, 2, nx + 1}], {1}}],
49:     If[Tsave[[1]] == 0,  $\psi_1 = Table[{x[[Xsave[[i]]]}, t0], Conjugate[\psi_0[[Xsave[[i]]]] * \psi_0[[Xsave[[i]]]]$ ,
50:     {i, 1, Length[Xsave]}],  $\psi_1 = \{\}$ ];
51:      $\psi_0 = Normalize[\psi_0/w]$ ;
52:     For[j = 1, j ≤ nt, j++,
53:     t1 = Apply[T, {tmin, tmax, j, nt}];
54:     b = ConstantArray[i *  $\hbar/(t_0 - t_1)$ , nx];
55:     t0 = t1;
56:      $\psi_0 = Join[\{0\}, LinearSolve[B + DiagonalMatrix[b], b * \psi_0[[2;;nx + 1]]], \{0\}]$ ;
57:     If[Tsave[[k]] == j,
58:     b = w *  $\psi_0$ ;
59:     fsum = Interpolation[Table[{x[[Xsave[[i]]]}, b[[Xsave[[i]]]}], {i, 1, Length[Xsave]}], InterpolationOrder →
order];
60:     b/=Sqrt[NIntegrate[Conjugate[fsum[y]] * fsum[y], {y, xmin, xmax}]];
61:      $\psi_1 = Join[\psi_1, Table[{x[[Xsave[[i]]]}, t0], Re[Conjugate[b[[Xsave[[i]]]] * b[[Xsave[[i]]]}], {i, 1, Length[Xsave]}]]$ ;
62:     k++;
63: ];
64: ];
65: Interpolation[\psi_1, InterpolationOrder → order]
66: ]]
67: GaussianWave[pos___,  $\lambda$ ___,  $\psi$ ___, min___, max___, n___, G___, order___]:=Block[
68: {i = 1, j = 1, x = Table[{Apply[G, {min, max, k, n + 1}], 0], {k, 0, n + 1}],  $\psi = ConstantArray[0, n + 2]$ , fsum, sum},
69: For[j = 1, j ≤ Length[pos], j++,
70: For[i = 1, i ≤ n + 2, i++,
71: x[[i, 2]]+=c[[j]] * Exp[- $\lambda[[j]] * (x[[i, 1]] - pos[[j]])^2$ ];
72:  $\psi[[i]] = x[[i, 2]]$ ;
73: ];
74: ];
75: fsum = Interpolation[x, InterpolationOrder → order];
76:  $\psi/=Sqrt[NIntegrate[Conjugate[fsum[y]] * fsum[y], {y, xmin, xmax}]]$ 
77: ]
78: intorder = 3;
79: m = 1;
80:  $\hbar = 1$ ;
81: pos = {1.5};
82:  $\lambda = \{0.5\}$ ;
83: a = {1};
84: ncore = 0.1;
85: xcore = 0.005;
86: xmin = -10.0;
87: xmax = 10.0;
88: nx = 399;
89: Xsave = Range[1, nx + 2, Floor[(nx + 2)/50]];
90: tmin = 0;
91: tmax = 8;
92: nt = 10000;
93: Tsave = Range[0, nt, Floor[nt/50]];
94: pmin = -4.0;
95: pmax = 4.0;
96: pp = 2;
97: MyGridSym[min___, max___, i___, n___]:=
98: Block[{xc = (max + min)/2}, CoreGrid[xc, xc + xcore * (max - xc), max, i - n/2, ncore * n/2, n/2]];
99: ListPlot[Table[{MyGridSym[xmin, xmax, i, nx + 1], Potential[MyGridSym[xmin, xmax, i, nx + 1]]}, {i, 0, nx + 1}], PlotRange->All]

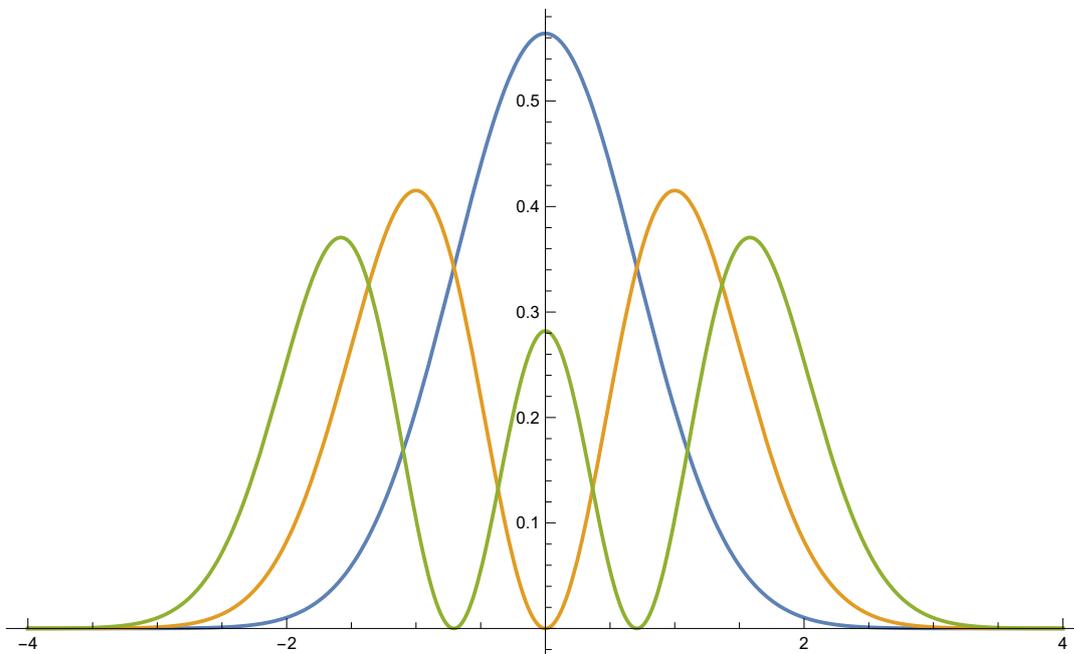
```



```

100: sys1 = Eigensystem[Tri[nx, xmin, xmax, - $\hbar^2/(2 * m)$ , Potential, MyGridSym], -3];
101: Show[Plot[Evaluate[Table[EigenFunction[sys1[[2]]][[Length[sys1[[2]]]-i]], xmin, xmax, MyGridSym, intorder][x]^2,
{i, 0, pp}], {x, pmin, pmax}, PlotRange → All], ImageSize → Large]

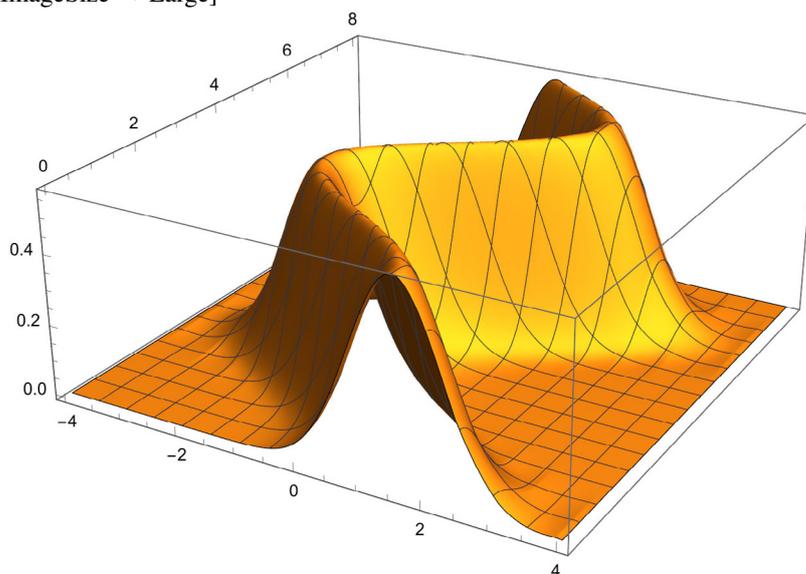
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102:  $\psi_1 = \text{TimeSolve}[-\hbar^2/(2 * m), \text{Potential}, \text{GaussianWave}[\text{pos}, \lambda, a, \text{xmin}, \text{xmax}, \text{nx}, \text{MyGridSym}, \text{intorder}],$ 
       $\text{xmin}, \text{xmax}, \text{MyGridSym}, \text{Xsave}, \text{tmin}, \text{tmax}, \text{nt}, \text{UniformGrid}, \text{Tsave}, \text{intorder}];$ 
103:  $\text{Show}[\text{Plot3D}[\psi_1[x, t], \{x, \text{pmin}, \text{pmax}\}, \{t, \text{tmin}, \text{tmax}\}, \text{PlotRange} \rightarrow \text{All}, \text{PlotPoints} \rightarrow \{200, 200\}],$ 
       $\text{ImageSize} \rightarrow \text{Large}]$ 

```



## Conclusion

This paper shows a successful implementation of irregular grid in the Schrödinger equation, which can be used to resolve smaller detail like the core-potential of atoms. This approach with the irregular grid can be used for other differential equations.

## Declaration of Generative AI and AI-Assisted Technologies in the Writing Process

During the preparation of this work the author used Chat-gpt for literature study. After using this tool/service, the author reviewed and edited the content as needed and takes full responsibility for the content of the publication.

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