# 2D eigensolver

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### 1 What: What are we doing

We are creating a 2D eigensolver in python. Specifically, we are applying a similar method to our 1D eigensolver in an attempt to model a particle confined to some 2D box.

$$\hat{H} |\psi_n\rangle = E_n |\psi_n\rangle \tag{1}$$

Again, we observe the schrodinger equation in Hamiltonian form, but this time we expand it slightly to explain where the '2D' comes in

$$\hat{H} |\psi_n\rangle = (\hat{V}(x,y) + \hat{T}) |\psi_n\rangle \tag{2}$$

$$= (\hat{V}(x,y) + \frac{\vec{p}^2}{2m}) |\psi_n\rangle \tag{3}$$

$$= (\hat{V}(x,y) + \frac{\mathbf{p}_x^2}{2m} + \frac{\mathbf{p}_y^2}{2m}) |\psi_n\rangle \tag{4}$$

## 2 Why: Why are we doing it

In theory we could expand to a full 3D, but the ultimate problem we are concerned with is the propagation of a beam through variable refractive index. Treating the changes in refractive index, the Helmholtz equation has the same form as the schrödinger equation, so the 2 spacial dimensions of each 'slice' are the 2 spacial dimensions the 'particle' finds itself in. The beam propagation direction is represented by time.

## 3 How: What is the design method for this

the state vector  $|\psi\rangle$  can be cast to multiple different forms, most commonly that of the probability amplitude function  $\psi(x)=\langle x|\psi\rangle$ . In the 1D case, we discretize the domain of this function, restricting it to some finite range, and representing it as a 1D numpy array. The utility of using a 1D numpy array is that the Hamiltonian operator can be represented as a 2D matrix, and we can

take advantage of existing libraries and methods to calculate it's eigenvectors.

We want to find a representation for a 2D version of our state vector, some code representation that can allow us to keep the hamiltonian in a form we can reasonably calculate the eigenvectors of. My initial idea was to use a 2D grid,

$$\psi(x,y) = \begin{bmatrix} \psi(x_0, y_0) & \psi(x_0, y_0 + \Delta y) & \psi(x_0, y_0 + 2\Delta y) & \dots \\ \psi(x_0 + \Delta x, y_0) & \psi(x_0 + \Delta x, y_0 + \Delta y) & \psi(x_0 + \Delta x, y_0 + 2\Delta y) & \dots \\ \psi(x_0 + 2\Delta x, y_0) & \psi(x_0 + \Delta x, y_0 + \Delta y) & \psi(x_0 + 2\Delta x, y_0 + 2\Delta y) & \dots \\ \vdots & \vdots & \ddots & \ddots \end{bmatrix}$$
(5)

This allows for the  $\mathbf{p}_x^2$  to remain the same.

$$\mathbf{p}_x^2 = -\hbar^2 \nabla_x^2 = -\hbar^2 \frac{\partial^2}{\partial x^2} \tag{6}$$

In the 1D case we wrote the x-momentum operator as a matrix taking advantage of the finite difference method to write the 2nd derivative

$$\mathbf{p}_{x}^{2} = -\frac{\hbar^{2}}{\Delta x^{2}} \begin{bmatrix} -2 & 1 & 0 & 0 & \dots \\ 1 & -2 & 1 & 0 & \dots \\ 0 & 1 & -2 & 1 & \dots \\ \vdots & \vdots & \vdots & \ddots & \ddots \end{bmatrix}$$
(7)

using a more concise  $\alpha$  and  $\beta$  as shorthand

$$\mathbf{p}_{x}^{2} |\psi\rangle = \begin{bmatrix} \alpha & \beta & 0 & 0 & \dots \\ \beta & \alpha & \beta & 0 & \dots \\ 0 & \beta & \alpha & \beta & \dots \\ \vdots & \vdots & \vdots & \ddots & \ddots \end{bmatrix} \begin{bmatrix} \psi(x_{0}) \\ \psi(x_{0} + \Delta x) \\ \psi(x_{0} + 2\Delta x) \\ \vdots \\ \psi(x_{f}) \end{bmatrix}$$
(8)

we can observe then that this matrix ALSO works for  $\psi(x,y)$ 

$$\begin{aligned} \mathbf{p}_{x}^{2}\psi(x,y) &= \\ \begin{bmatrix} \alpha & \beta & 0 & 0 & \dots \\ \beta & \alpha & \beta & 0 & \dots \\ 0 & \beta & \alpha & \beta & \dots \\ \vdots & \vdots & \vdots & \ddots & \ddots \end{bmatrix} \begin{bmatrix} \psi(x_{0},y_{0}) & \psi(x_{0},y_{0}+\Delta y) & \psi(x_{0},y_{0}+2\Delta y) & \dots \\ \psi(x_{0}+\Delta x,y_{0}) & \psi(x_{0}+\Delta x,y_{0}+\Delta y) & \psi(x_{0}+\Delta x,y_{0}+2\Delta y) & \dots \\ \psi(x_{0}+2\Delta x,y_{0}) & \psi(x_{0}+\Delta x,y_{0}+\Delta y) & \psi(x_{0}+2\Delta x,y_{0}+2\Delta y) & \dots \\ \vdots & \vdots & \ddots & \ddots \end{bmatrix} \end{aligned}$$

At least in the case of  $\mathbf{p}_x^2$ , this representation works. The trouble beings when inspecting the case of  $\mathbf{p}_y^2$ . By observation, all I've found in the way of  $\mathbf{p}_y^2$  is the following result

$$\mathbf{p}_{y}^{2} |\psi\rangle = (\mathbf{p}_{x}^{2} \boldsymbol{\psi}(x, y)^{T})^{T}$$

$$= \boldsymbol{\psi}(x, y) \mathbf{p}_{x}^{2T}$$

$$(10)$$

But at least with the methods we are using, I don't believe we can apply post-multiplication of matrices as an operator, as our Hamiltonian will be interpreted as pre-multiplying. so, instead I propose another method.

Similar to how matrices are stored in memory, we can instead encod all of our 2D data into a single 1D eigenvector (numpy array), keeping track of the 'stride'.

4 How: How is this implemented in code

5 How: How can this be used in the future