

# Project 5 - FYS4150

## The Schrödinger equation

Nanna Bryne, Johan Mylius Kroken, Vetle A. Vikenes  
(Dated: December 7, 2022)

Abstract

Supporting material may be found in the following GitHub repository: <https://github.com/Vikenes/FYS4150/tree/main/project5>.

### NOMENCLATURE

#### Basics

$x \in [0, 1]$ ;  $x \rightarrow x_i = ih$  with  $i = 0, 1, \dots, M - 1$ .

$y \in [0, 1]$ ;  $y \rightarrow y_j = jh$  with  $j = 0, 1, \dots, M - 1$ .

$\mathbf{x} = (x, y)$ ;  $\mathbf{x} \rightarrow \mathbf{x}_{i,j} = h(i, j)$  with  $i, j \in [0, M - 1]$ .

$t \in [0, T]$ ;  $t \rightarrow t_n = n\Delta t$  with  $n = 0, 1, \dots, N_t - 1$ .

$u(t, \mathbf{x}) \rightarrow u(t_n, \mathbf{x}_{i,j}) \equiv u_{i,j}^{(n)}$ .

$U^{(n)}$  is a matrix with elements  $u_{i,j}^{(n)}$ .

$v(\mathbf{x}) \rightarrow v(\mathbf{x}_{i,j}) \equiv v_{i,j}$ .

$V$  is a matrix with elements  $v_{i,j}$ .

#### NB

$M$  is the number of points along  $x$  and  $y$  axis.

$M - 1$  is the number of steps.

$M - 2$  is the number of internal points (excluding boundary points).

#### Dirichlet boundary conditions

$u(t, \mathbf{x}_{0,j}) = u(t, x=0, y) = 0$ .

$u(t, \mathbf{x}_{M-1,j}) = u(t, x=1, y) = 0$ .

$u(t, \mathbf{x}_{i,0}) = u(t, x, y=0) = 0$ .

$u(t, \mathbf{x}_{i,M-1}) = u(t, x, y=1) = 0$ .

### I. INTRODUCTION

Blah blah

For a single, non-relativistic particle with mass  $m_P$  in a two-dimensional potential  $V(t, \mathbf{x})$ , the Schrödinger equation reads<sup>1</sup>

$$i\hbar \frac{\partial}{\partial t} \Psi(t, \mathbf{x}) = -\frac{\hbar^2}{2m_P} \nabla^2 \Psi(t, \mathbf{x}) + V(t, \mathbf{x}) \Psi(t, \mathbf{x}). \quad (1)$$

For a set of initial and boundary conditions, the partial differential equation (PDE) describes the temporal and spatial evolution of the complex-valued function  $\Psi(t, \mathbf{x})$  related to the quantum state of the aforementioned particle. In such a case, at a time  $t$ , the probability density for an experimentalist to locate the particle at  $\mathbf{x}$  ("for detecting ..." is better, but I don't want to copy Anders) is large P or small p???

$$P(\mathbf{x}; t) = |\Psi(t, \mathbf{x})|^2 = \Psi^*(t, \mathbf{x}) \Psi(t, \mathbf{x}), \quad (2)$$

originating from the Born rule; fill me

In this paper we will consider a dimensionless time-independent potential, i.e. we let  $V(t, \mathbf{x}) \rightarrow v(\mathbf{x})$ . The specifics of the scaling do not concern us in this paper, and we simply rewrite equation (1) to the dimensionless equation

$$i \frac{\partial}{\partial t} u(t, \mathbf{x}) = -\nabla^2 u(t, \mathbf{x}) + v(\mathbf{x}) u(t, \mathbf{x}), \quad (3)$$

where we substituted  $\Psi(t, \mathbf{x}) \rightarrow u(t, \mathbf{x})$ . In equation (3) all variables are dimensionless. When demanding the proper normalisation on  $u(t, \mathbf{x})$ , it follows that the Born rule now takes the form of

$$p(\mathbf{x}; t) = |u(t, \mathbf{x})|^2 = u^*(t, \mathbf{x}) u(t, \mathbf{x}). \quad (4)$$

Should maybe rephrase this paragraph.

### II. METHODS

choose one of these:

$$u(t=0, \mathbf{x}) = \exp\{-(\mathbf{x} - \mathbf{x}_c)^T \Sigma^{-1} (\mathbf{x} - \mathbf{x}_c) + i\mathbf{p}^T (\mathbf{x} - \mathbf{x}_c)\}; \\ \Sigma = \text{diag}(\sigma^2) \quad (5)$$

---

<sup>1</sup> In position space, that is. Should we comment on this?

$$u(t=0, \mathbf{x}) = \exp \left\{ -\frac{(x-x_c)^2}{2\sigma_x^2} - \frac{(y-y_c)^2}{2\sigma_y^2} + \mathbf{i}\mathbf{p} \cdot (\mathbf{x} - \mathbf{x}_c) \right\}$$

(6)

### III. RESULTS

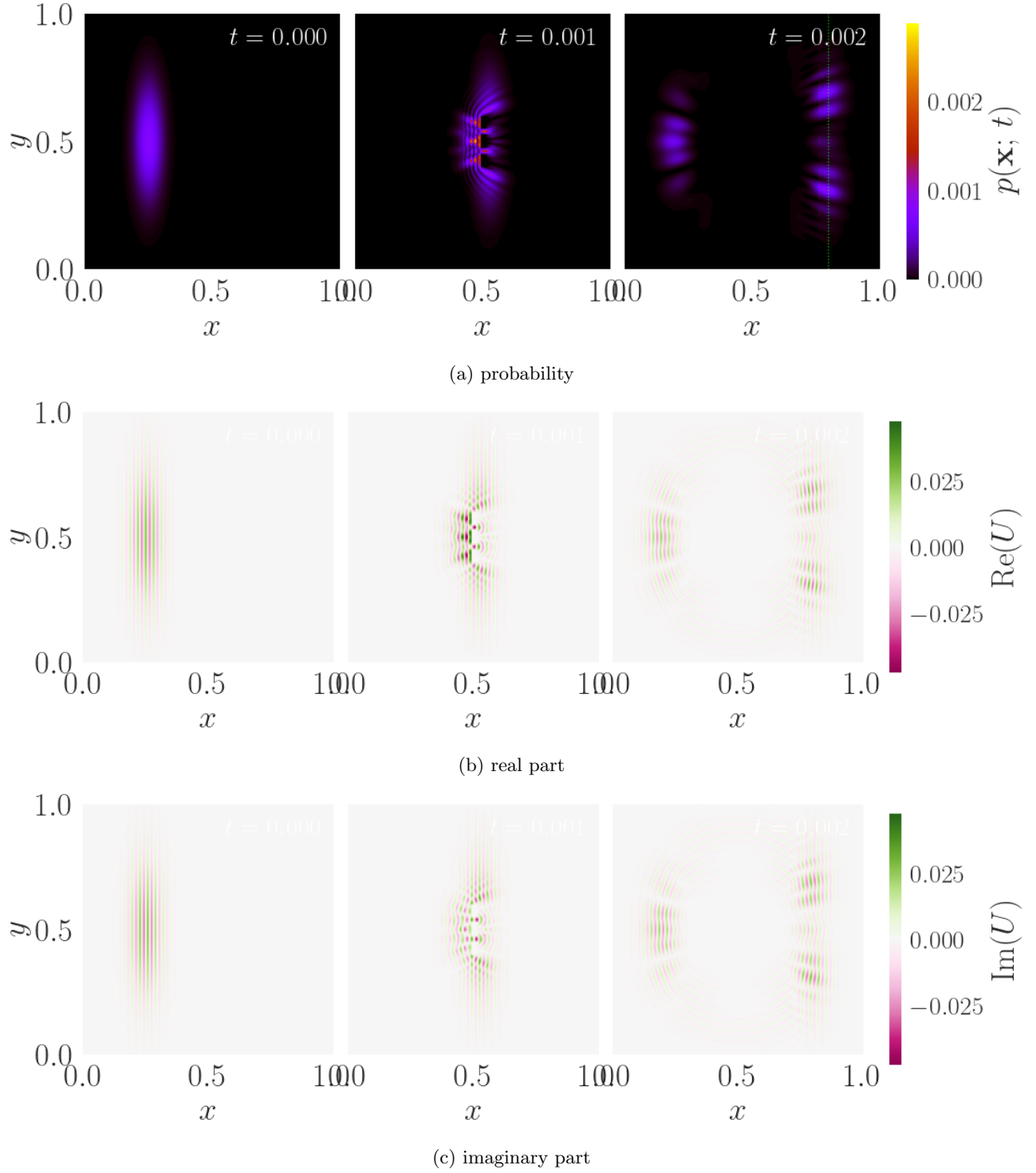


FIG. 1: Colour maps showing blah blah

#### IV. DISCUSSION

#### V. CONCLUSION

#### Appendix A: Discretisation

Suppose you have the (1+1)-dimensional PDE  $\partial u / \partial t = F$  where  $u = u(t, x)$  and  $F = F(t, x, u, \partial u / \partial x, \partial^2 u / \partial x^2)$ . The Crank-Nicolson

scheme reads [ref](#)

$$\frac{u_i^{(n+1)} - u_i^{(n)}}{\Delta t} = \frac{1}{2} \left( F_i^{(n+1)} - F_i^{(n)} \right), \quad (\text{A1})$$

where  $u_i^{(n)} = u(n\Delta t, i\Delta x)$  and  $F_i^{(n)}$  is  $F$  evaluated for  $i, n$  and  $u_i^{(n)}$ . In our (2+1)-dimensional case where  $u = u(t, \mathbf{x})$  we have

$$\frac{\partial u}{\partial t} = F(t, \mathbf{x}, u, \nabla^2 u) = i \left( \nabla^2 u - v(\mathbf{x})u \right), \quad (\text{A2})$$

and this approach translates to

$$\frac{u_{i,j}^{(n+1)} - u_{i,j}^{(n)}}{\Delta t} = \frac{1}{2} \left( F_{i,j}^{(n+1)} - F_{i,j}^{(n)} \right) \quad (\text{A3})$$

where  $u_{i,j}^{(n)} = u(n\Delta t, \mathbf{x}_{i,j})$ ,  $\mathbf{x}_{i,j} = h(i, j)$ , and  $F_{i,j}^{(n)}$  is the right-hand side of equation [\(A2\)](#), explicitly:

$$F_{i,j}^{(n)} = i \left( \left[ \frac{\partial^2 u}{\partial x^2} \right]_{i,j}^{(n)} + \left[ \frac{\partial^2 u}{\partial y^2} \right]_{i,j}^{(n)} - v_{i,j} u_{i,j}^{(n)} \right); \quad (\text{A4})$$

We can approximate the two spatial double derivatives ([correct way to say?](#)) according to ([Don't know what this approximation is called](#)):

$$\left[ \frac{\partial^2 u}{\partial x^2} \right]_{i,j}^{(n)} \approx \frac{1}{h^2} (u_{i+1,j} - 2u_{i,j} + u_{i-1,j})^{(n)}; \quad (\text{A5a})$$

$$\left[ \frac{\partial^2 u}{\partial y^2} \right]_{i,j}^{(n)} \approx \frac{1}{h^2} (u_{i,j+1} - 2u_{i,j} + u_{i,j-1})^{(n)}; \quad (\text{A5b})$$

Define  $r \equiv \frac{i\Delta t}{2h^2}$ . Further, let

$$\begin{aligned} \mathcal{A}^{(n)} = & r (u_{i+1,j} - 2u_{i,j} + u_{i-1,j})^{(n)} \\ & + r (u_{i,j+1} - 2u_{i,j} + u_{i,j-1})^{(n)} \\ & - \frac{i\Delta t}{2} v_{i,j} u_{i,j}^{(n)}. \end{aligned} \quad (\text{A6})$$

Equation [\(A3\)](#) becomes:

$$u_{i,j}^{(n+1)} - \mathcal{A}^{(n+1)} = u_{i,j}^{(n)} + \mathcal{A}^{(n)}; \quad (\text{A7})$$

The final discretisation [\(A7\)](#) is valid for any step in time within the time range ( $n \in [0, N_t - 2]$ ) and all internal points on the grid ( $i, j \in [1, M - 2]$ ).