$$\frac{\partial}{\partial \theta} \operatorname{M} T(\xi) = \frac{\partial}{\partial \theta} \int_{\mathbb{R}_{n}}^{T} T(x) f(x, \theta) dx = \int_{\mathbb{R}_{n}}^{\theta} \frac{\partial}{\partial \theta} T(x) f(x, \theta) dx = \int_{\mathbb{R}_{n}}^{\theta} \frac{\partial}{\partial \theta} T(x) f(x, \theta) dx = \int_{\mathbb{R}_{n}}^{\theta} \frac{\partial}{\partial \theta} T(x) f(x, \theta) dx = \int_{\mathbb{R}_{n}}^{\theta} \int_{\mathbb{R}_{n}}^{\mathbb{R}_{n}} f(x, \theta) dx = \int_{\mathbb{R}_{n}}^$$

Project 3

## FYS3150 - Computational physics

# Quantum dots

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#### Abstract

Here is a short summary of the project.

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#### 1 Introduction

Quantum mechanics is an exciting field.

## 2 Theory

Here is all the theory needed to understand the project.

#### 2.1 The numerical foundation

This is the section explaining the numerical theory upon which the project is built.

#### 2.1.1 Monte Carlo simulations

A Monte Carlo simulation is a way of solving a mathematical or physical problem by generating a random (or pseudorandom  $^1$ ) sequence of numbers and evaluating some quantity on the assumption that our the random sequence of numbers is representative of the domain from which the quantity is evaluated. An example is evaluating the area of the unit circle by randomly placing points in a  $[-1,1] \times [-1,1]$  grid and find the fraction points whose distance to the origin is  $\leq 1$  and multiply this fraction by the area of the grid (i.e. 4). Such a simple Monte Carlo simulation can give the result as shown in figure 2.1.

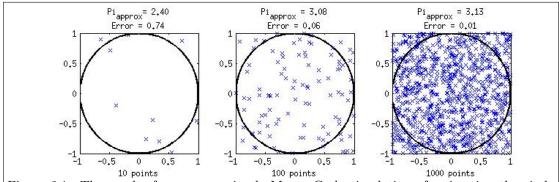


Figure 2.1: The results from a very simple Monte Carlo simulation of estimating the circle constant  $\pi$ . The precision increases with the number of points.

However, the method is not confined to this sort of problem, but can be applied to a variety of mathematical and physical problems. In this report, the method, through the Metropolis algorithm (see section 2.1.3) has been applied to a quantum mechanical system.

<sup>&</sup>lt;sup>1</sup>No electronic random number generator of today is truly random. The sequence of numbers generated will repeat itself after a long period. These periods however, are increadibly long and we will for this report consider the random number generators to be truly random.

#### 2.1.2 Importance sampling

Sometimes, functions are more important in certain domains. When we use a Monte Carlo approach to evaluate a quantity on such a function we can improve the method significantly by ensuring that the probability distribution from which we choose our random values reflect the parts where the function is important. We do of course have to remember that our function lives on its whole domain, so to not get a biased result, we need to weigh our result with respect to the probability function we have chosen. This is called importance sampling. **INSERT EXAMPLE HERE** 

#### 2.1.3 Metropolis algorithm

The metropolis algorithm is a Monte Carlo method discovered in ???? by ????.

#### 2.2 The physical system

This is the section explaining the physics of the system.

- 2.2.1 Quantum mechanics
- 2.2.2 Discretization
- 2.2.3 The virial theorem

#### 3 Method

This is the section explaining what has been done.

#### 3.1 Subsection

These will become apparent as the work boils down.

#### 4 Results and discussion

Section listing results and discussing them.

#### 4.1 Subsection

The same as the method section.

#### 5 Conclusion