

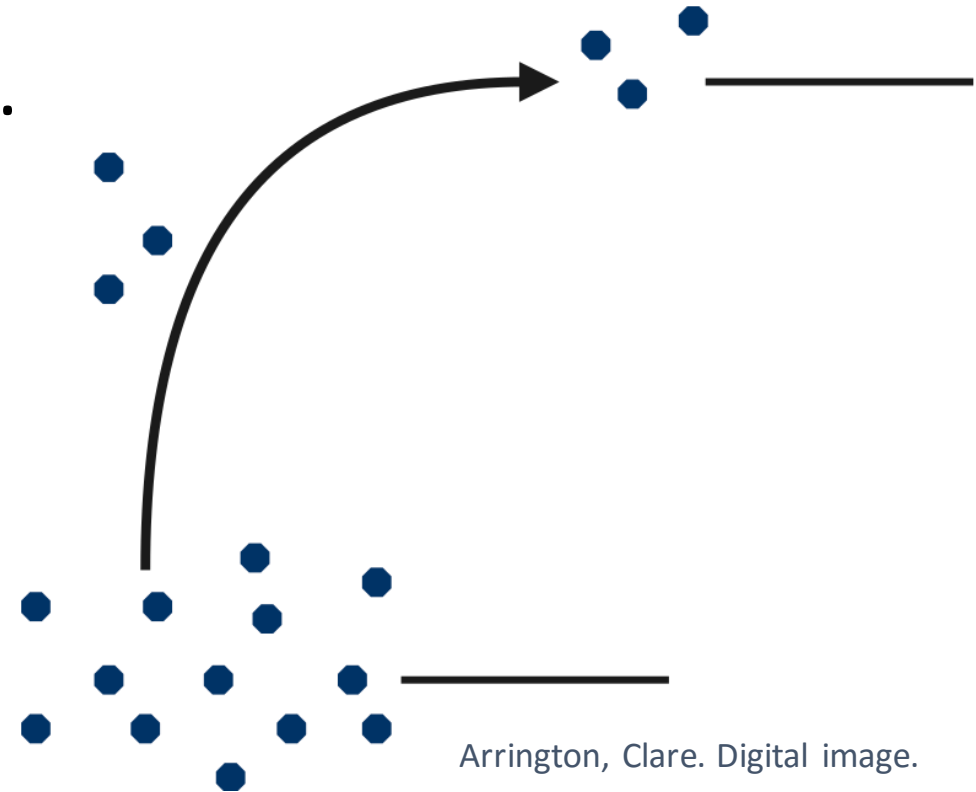
# Coherent Control of Atomic Population Using the Genetic Algorithm

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July 26th 2017

# Research Goal

- Create a program that gives the parameters needed for two lasers to excite a desired amount of  $^{87}\text{Rb}$  to an upper excited state.
- Otherwise known as Coherent Control



Arrington, Clare. Digital image.

# Outline

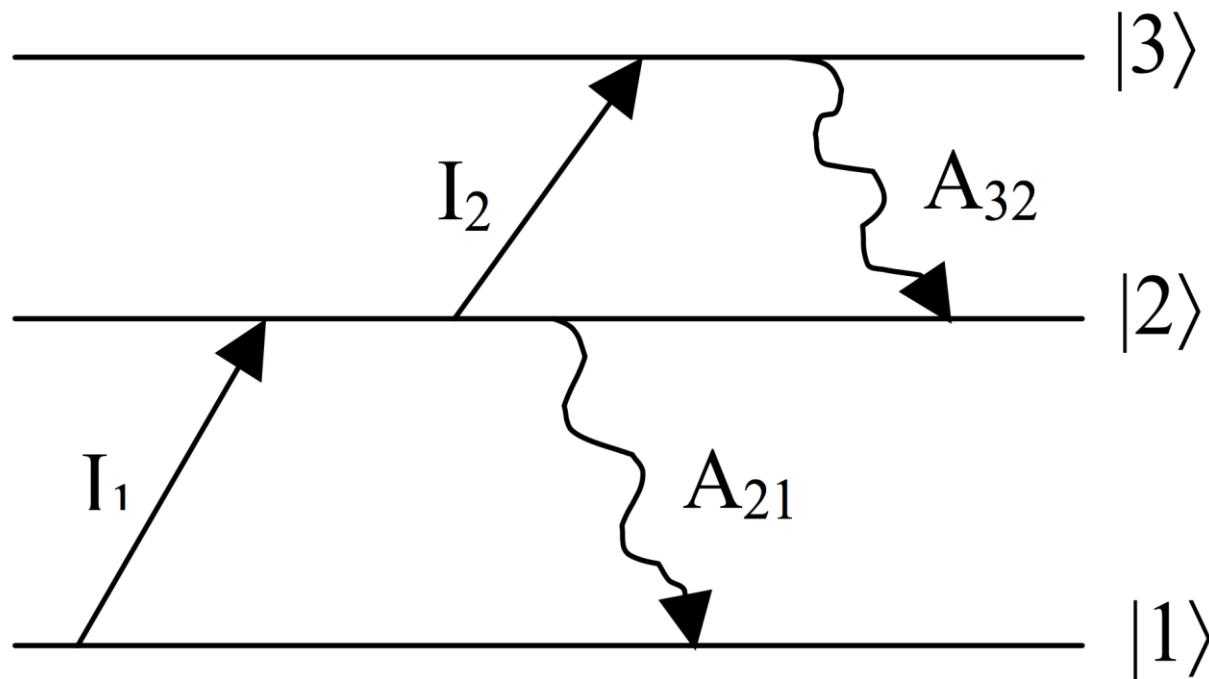
## Background

- An atom's interaction with light
- Technique for excitation
- Goal of coherent control
- Applications of coherent control

## Methods

- Modelling population dynamics
- Implementing the Genetic Algorithm

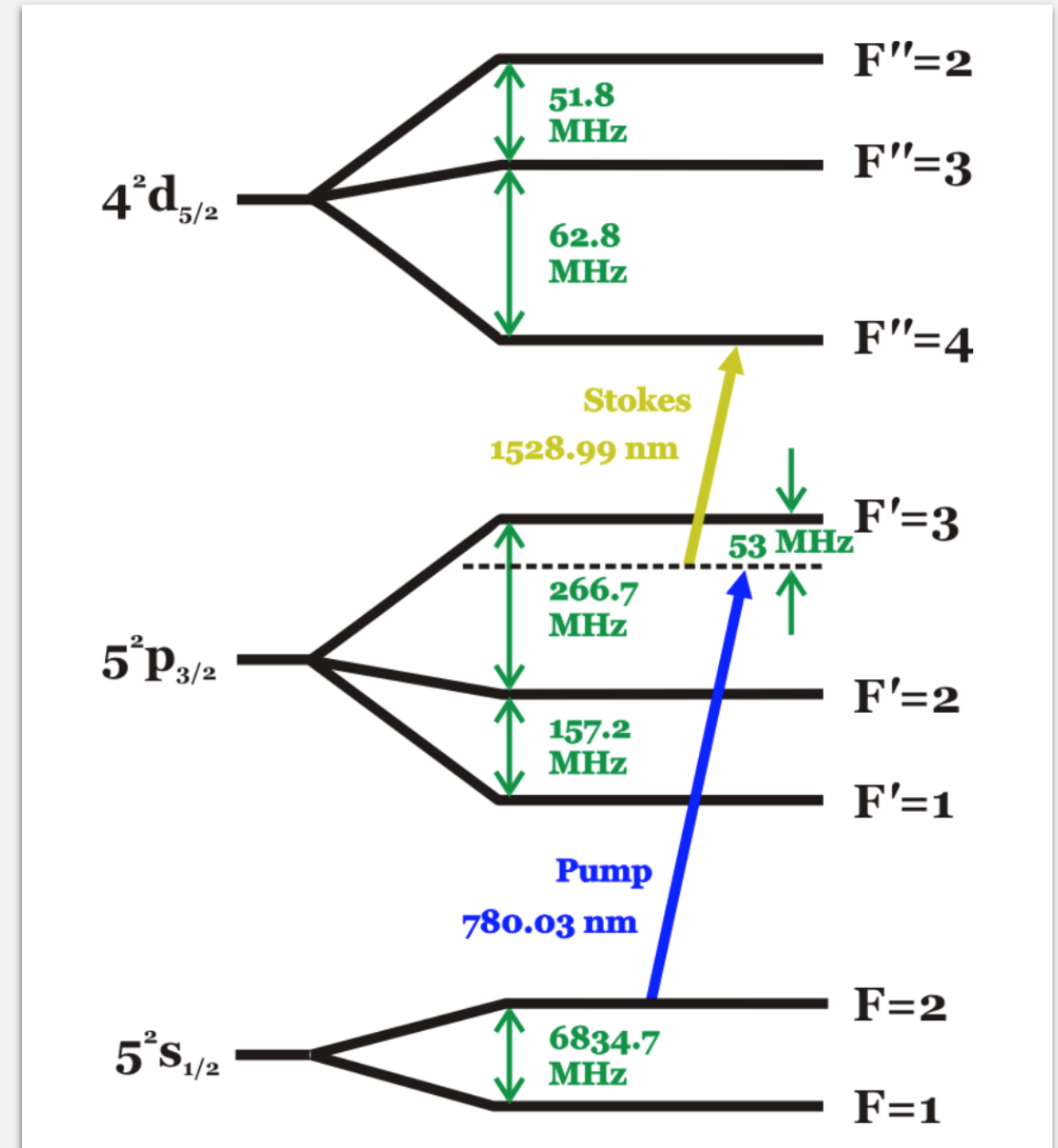
# Atom's Interaction with Light



- Light increases the amount of energy in a system
- Energy levels in an atom are discrete

# Coherent Excitation

- This is a technique to excite atoms efficiently
- The pump couples the ground and the transition state
- The stokes couples the transition state with the upper excited state



# Applications

## Bounded Quantum Storage Model

- Eve can store a limited number of quantum bits
- So Bob and Alice need fine control of the information transfer



Arrington, Clare. Digital image.

The background features a central white rectangular area. This area is bordered by four geometric shapes: a dark grey triangle in the top-left corner, a light grey triangle in the top-right corner, a light grey triangle in the bottom-left corner, and a dark blue triangle in the bottom-right corner. All triangles meet at a central point, creating a white diamond shape in the center.

# Methods

# Modelling Population Dynamics

- The Hamiltonian describes the interaction effects between the different energy levels
- The Liouville equation creates 9 coupled differential equations to work with

$$H = \begin{pmatrix} 0 & \Omega_1[t] & 0 \\ \Omega_1[t] & 2\Delta_1 & \Omega_2[t] \\ 0 & \Omega_2[t] & 2\Delta_2 \end{pmatrix} \quad \frac{\partial \rho}{\partial t} = \frac{1}{i\hbar} [H, \rho]$$

Hamiltonian

Quantum Liouville Equation

Density Matrix

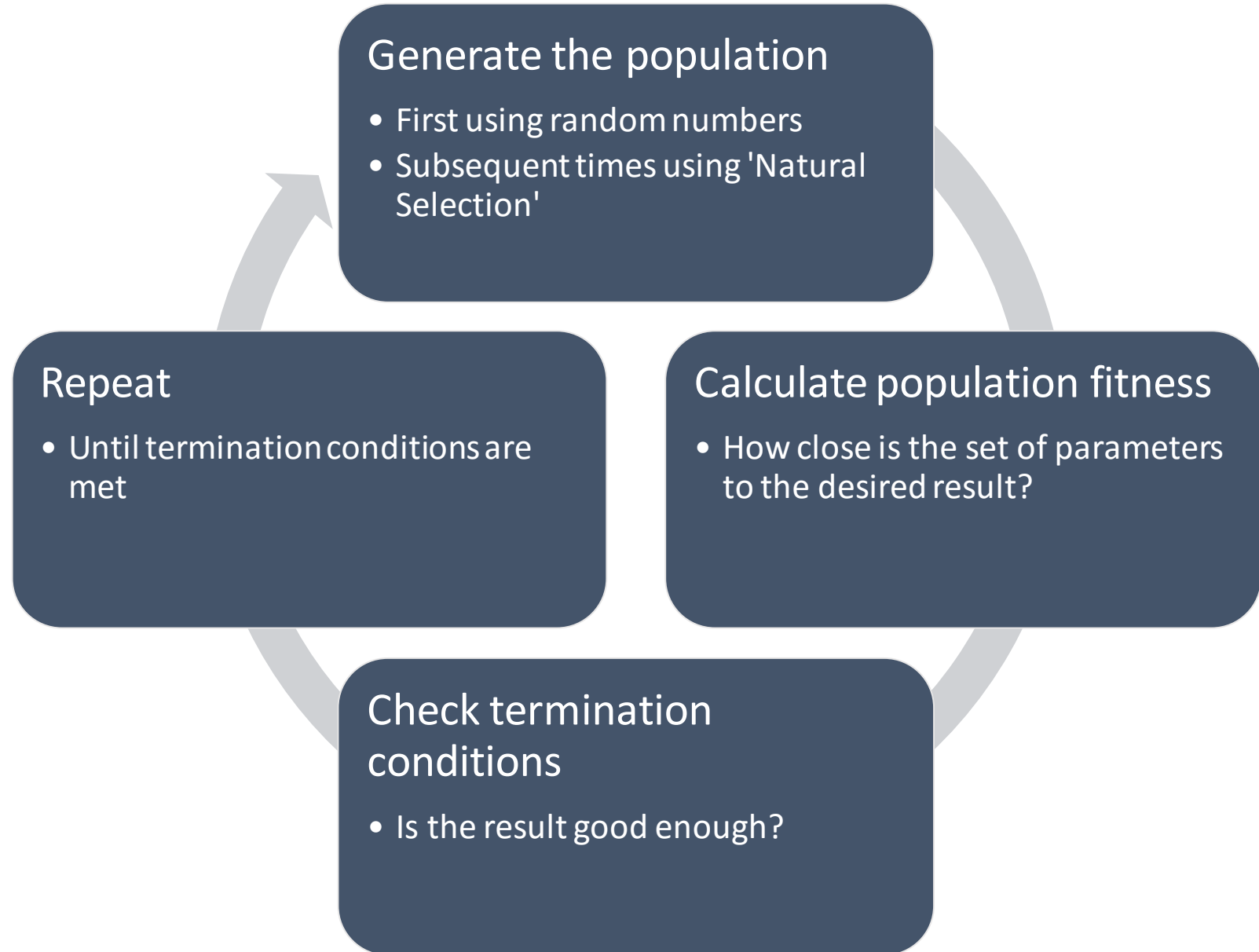


# Why Genetic Algorithm

Intuitive and easy  
to implement

Efficient way of  
searching the  
possible parameter  
space

# How the Genetic Algorithm Works



# Generate the Initial Population

Generate a set of parameters randomly within a valid parameter range

- For example, intensities can be from 0 to 100 mW/cm<sup>2</sup>

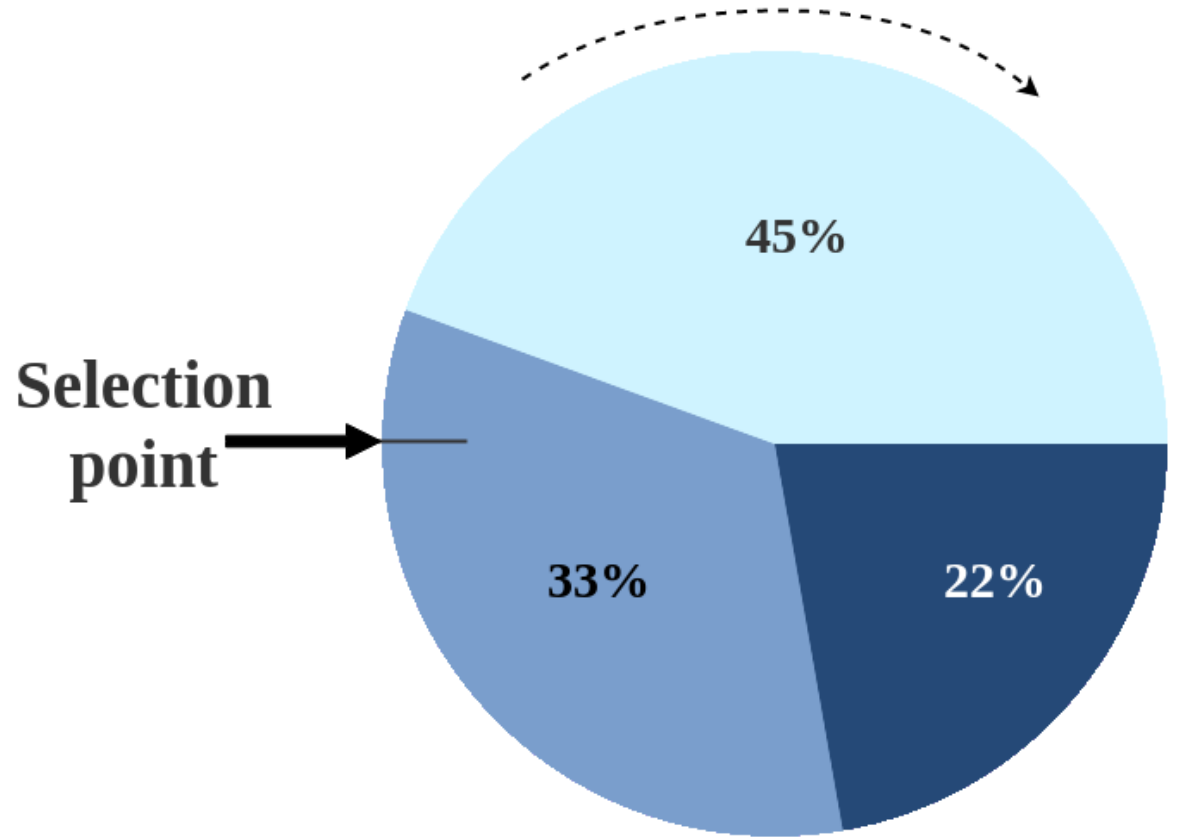
# Calculate the Population Fitness

How close is the amount transferred to the desired amount?

$$fitness = \frac{1}{|target - transferred|}$$

# Obtaining the Parents

- Use the fitness values to assign weightings to the different parameters. The higher the fitness value, the more likely it is to be chosen
- Use the weightings to sample from the existing population to select the parents



Arrington, Clare. Digital image

# New Generation

## Crossover

Use the parent's 'genetic code' to create the new parameter set (child)

## Mutation

Slight probability that a parameter in the child regenerates.

## Repeat

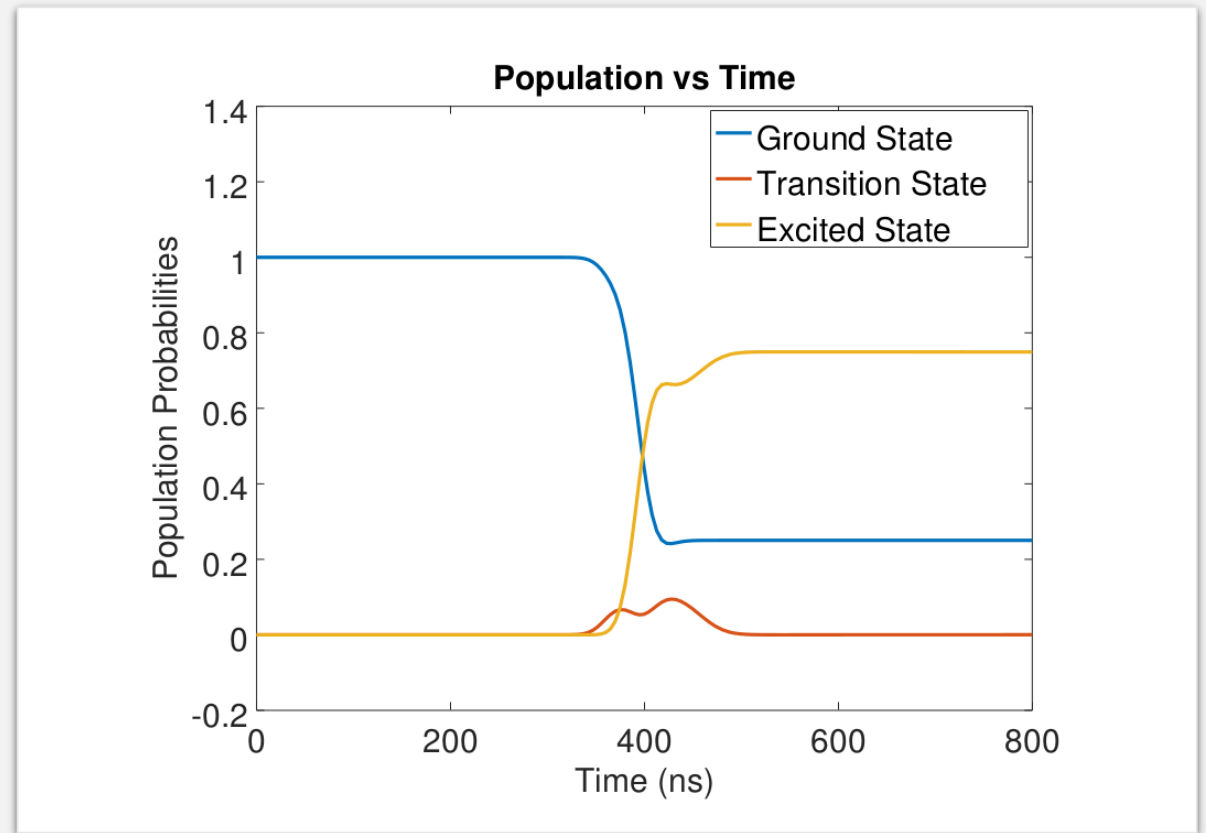
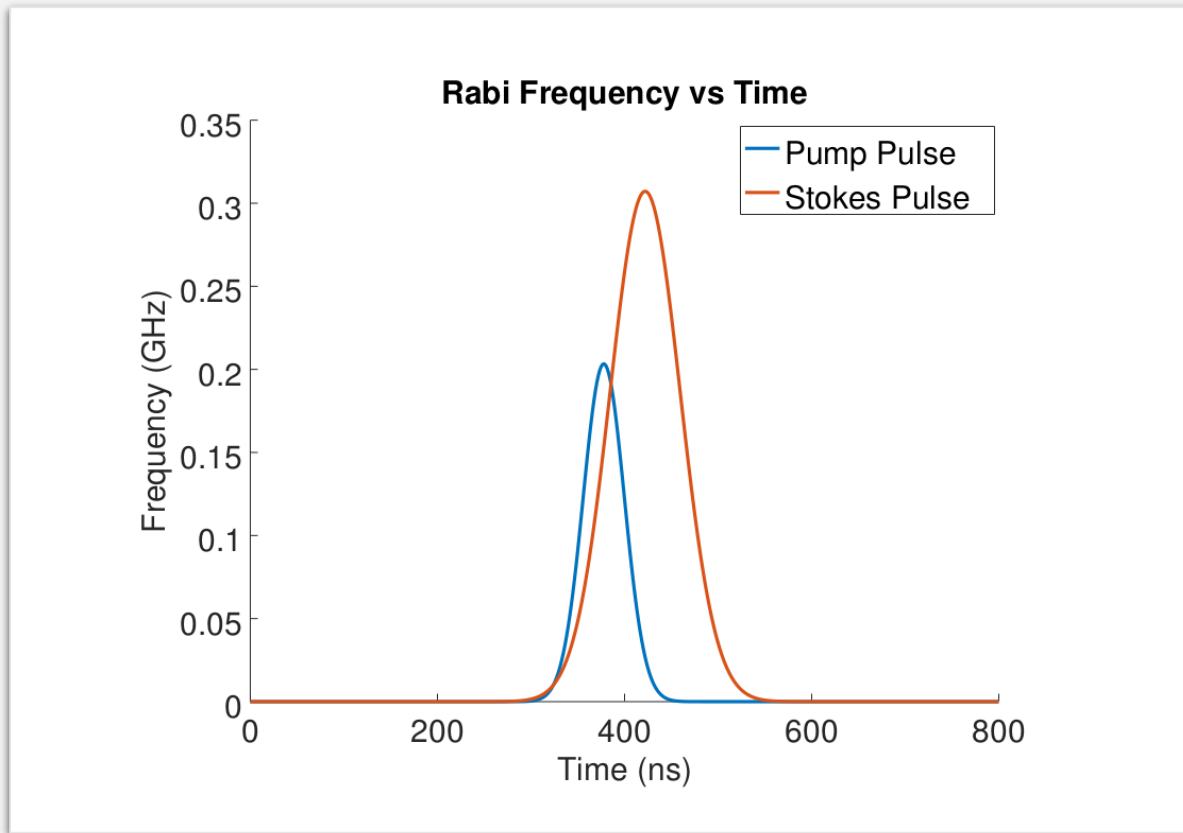
Many times to replace the existing population with the new one

# Termination Condition

This is what stops  
the algorithm

Stops when  
algorithm barely  
improves its  
findings

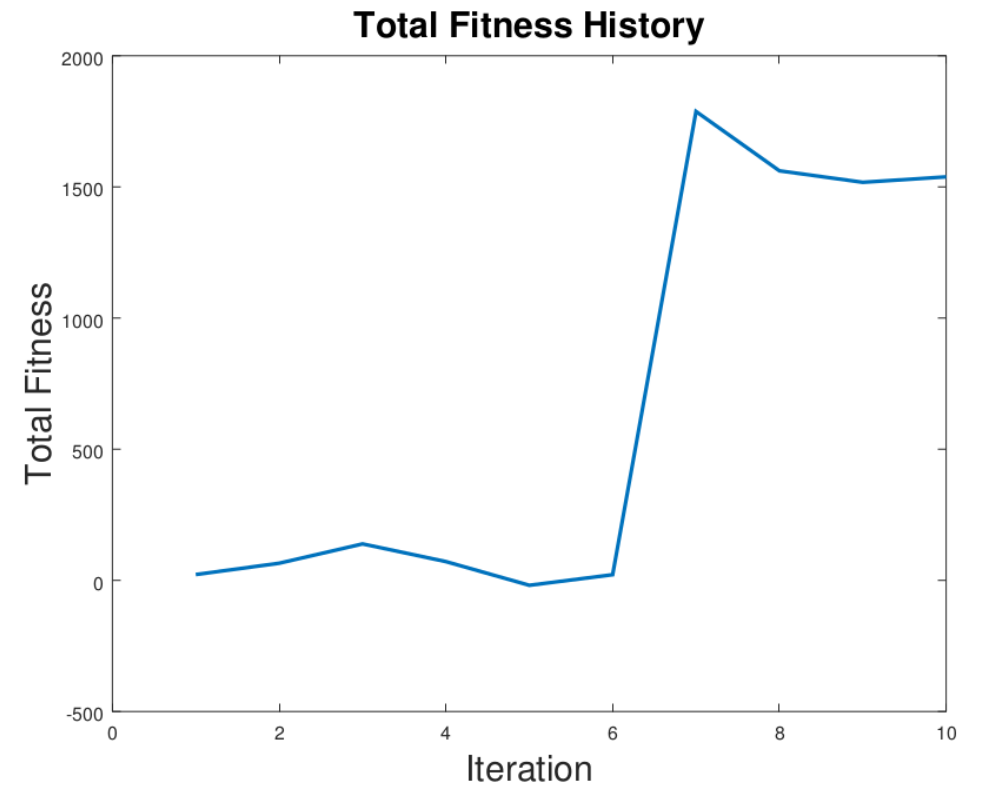
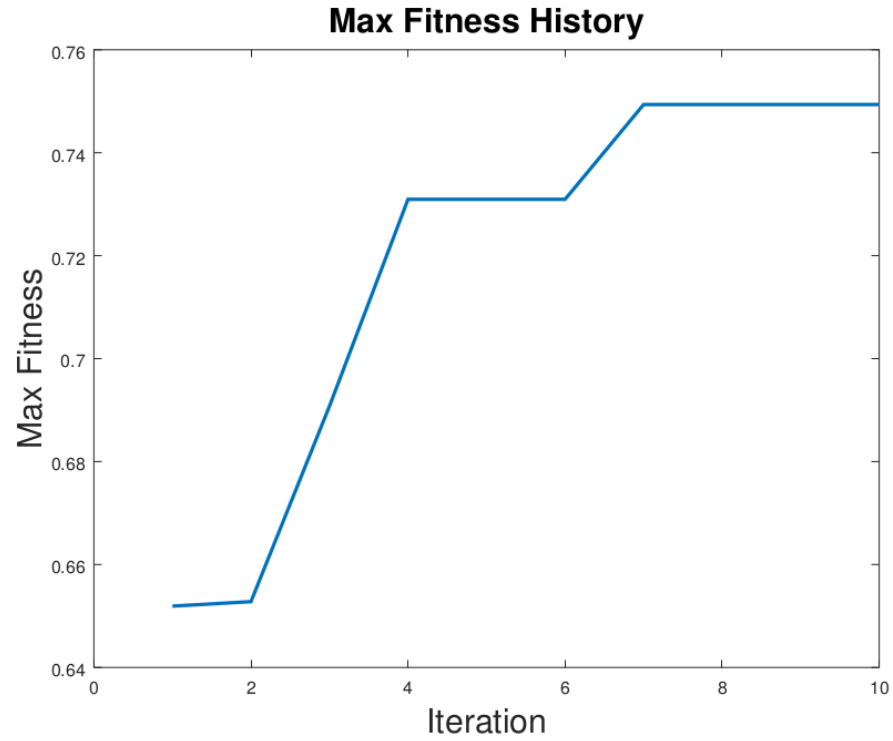
Report back the  
best result found  
when met



# Sample Result

Finding the parameters needed for 75% population transfer





# Algorithmic Efficiency

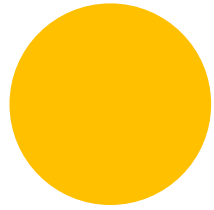
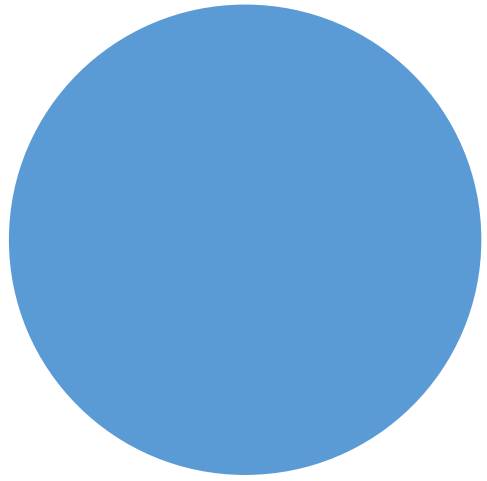
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# Acknowledgements

- UMW Summer Science Institute
- UMW Department of Physics
- Dr. Hai Nguyen, Faculty advisor
- Hannah Killian, Fellow researcher



Questions?