

# Final Year Project

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# Chapter 1

## Introduction

The Prisoners Dilemma is a classic game theory topic

### 1.1 Background

#### 1.1.1 Game Theory

fill this section in when I do the course?

#### 1.1.2 Iterated Prisoners Dilemma

The Prisoners Dilemma is a well known game theory problem based on the example of a pair of prisoners and their subsequent interrogation. The game is as follows:

Something about the PD

The single game itself is very basic and is modeled in the following way:

*give – a – model – here*

The Iterated Prisoners Dilemma is the iterated version of the Prisoners Dilemma<sup>1</sup>. The iteration of the game is what makes the game an interesting concept, as now **learn the technical stuff and put it here!!!!**

#### 1.1.3 Machine Learning Concepts

This section will briefly provide a background to machine learning algorithms implemented in the axelrod-doj<sup>2</sup>. This is by no means a comprehensive look

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<sup>1</sup>reference this stuff dude, come on..

<sup>2</sup>referencing opportunity here

into these subjects but will provide sufficient background on technical discussion later on.

## Genetic Algorithms

Genetic Algorithms are a description of techniques for generating solutions complex problems such as searching and, in our case, optimization<sup>3</sup>. The basis of a genetic algorithm is focused on a cycle of evolution. Like nature, we create a survival of the fittest concept<sup>4</sup> to evaluate a population, kill off the weakest members and create offspring from the most successful population.

Put a figure of the cycle here.

Initially we create a heuristic function, say our fitness function, which is a measure of how successful a candidate in our population is. Then we run our whole population through this function, ranking each one by how successful their score is. At this point we can create a cut off<sup>5</sup> to decide which of the population not to put through to the next round.

## Bayesian Optimization

### 1.2 Brief Overview

In this document I will be looking at the creation of sequences to beat given players in The Iterated Prisoners Dilemma<sup>6</sup>. My research looked into just the single opponent use case, but the idea of designing a sequence for a given number of opponents is looked at in the further study of the report

Problem:

Given a certain opponent, O, (with a provided strategy, S) what is the best possible sequence of moves, in a game of n turns, made by my strategy to maximise my players score?

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<sup>3</sup>Mitchell, Melanie (1996). An Introduction to Genetic Algorithms. Cambridge, MA: MIT Press. ISBN 9780585030944. learn to reference soon

<sup>4</sup>need to reference Darwin?

<sup>5</sup>Can often be referred to as the bottleneck

<sup>6</sup>Reference this for some background

## Chapter 2

# Literature Review

### 2.1 Background

### 2.2 Strategies Of Interest

#### 2.2.1 Tit for Tat

this is a sentence

#### 2.2.2 Cyclor

this is a type of opponent

#### 2.2.3 Other

add at least two here

## Chapter 3

# Approach To Problems

### 3.1 In Depth Definition Of Task

### 3.2 Solution Form

The sequence archetype will use the `Cycler()` player for our strategy each time, only editing the input to improve our score. To this model we can apply an optimised input of length  $n$  (tbd) to the player, this sequence, as per the design of the strategy will then be repeated until the games end (if  $n = \text{len}(\text{game})$  then we are just calculating the sequence for the whole game).

The input sequence itself will be created using a genetic optimisation. Starting with a set of randomly generated sequences, we will have each one play the opponent and return with a score. These sequences will be ranked and the lowest  $x\%$  will be discarded, resulting in a fitter, but smaller, population than before. This smaller population will then create offspring using a —X TBD method X— pairing algorithm before mutating with —X TBD method X—. This new set of offspring will be included in the next scoring round and the process repeats for  $k$  number of rounds

This sequence of Play-Rank-Create-LOOP will be the basis of creating the optimal strategy for each other opponent.

## Chapter 4

# Implementation Of Sequence Discovery

## Chapter 5

# Results and Discussion



## Chapter 6

# Practical Applications for Solution Sequences

## Chapter 7

# Summary and Future Research