

# CS540 Introduction to Artificial Intelligence

## Lecture 20

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# Genetic Algorithm

## Description

- Start with a fixed population of initial states.
- Find the successors by:
  - ① Cross over.
  - ② Mutation.

# Reproduction Probability

## Definition

- Each state in the population has probability of reproduction proportional to the fitness. Fitness is the opposite of the cost: higher cost means lower fitness. Use  $F$  to denote the fitness function, for example,  $F(s) = \frac{1}{f(s)}$  is a valid fitness function.

$$p_i = \frac{F(s_i)}{\sum_{j=1}^N F(s_j)}, i = 1, 2, \dots, N$$

- A pair of states are selected according to the reproduction probabilities (using CDF inversion).

# Cross Over

## Definition

- The states need to be encoded by strings.
- Cross over means swapping substrings.
- For example, the children of 10101 and 01010 could be the same as the parents or one of the following variations.

(11010, 00101) , (10010, 01101)

(10110, 01001) , (10100, 01011)

# Mutation

## Definition

- The states need to be encoded by strings.
- Mutation means randomly updating substrings. Each character is changed with small probability  $q$ , called the mutation rate.
- For example, the mutated state from 000 could stay the same or be one of the following.

one of 001, 010, 100, with probability  $q(1 - q)^2$

one of 011, 101, 110, with probability  $q^2(1 - q)$

and 111, with probability  $q^3$

# Cross Over, Modifications

## Definition

- The previous cross over method is called 1 point cross over.
- It is also possible to divide the string into  $N$  parts. The method is called  $N$  point cross over.
- It is also possible to choose each character from one of the parents randomly. The method is called uniform cross over.

# Mutation, Modifications

## Definition

- For specific problems, there are ways other than flipping bits to mutate a state.
- ① Two-swap: ABCDE to EBCDA
- ② Two-interchange: ABCDE to EDCBA

# Genetic Algorithm TSP Example

## Discussion



# Genetic Algorithm, Part I

## Algorithm

- Input: state space  $S$  represented by strings  $s$  and cost function  $f$  or fitness function  $F$ .
- Output:  $s^* \in S$  that minimizes  $f(s)$ .
- Randomly generate  $N$  solutions as the initial population.

$$s_1, s_2, \dots, s_N$$

- Compute the reproduction probability.

$$p_i = \frac{F(s_i)}{\sum_{j=1}^N F(s_j)}, i = 1, 2, \dots, N$$

# Genetic Algorithm, Part II

## Algorithm

- Randomly pick two states according to  $p_i$ , say  $s_a, s_b$ .  
Randomly select a cross over point  $c$ , swap the strings.

$$s'_a = s_a [0...c) s_b [c...m)$$

$$s'_b = s_b [0...c) s_a [c...m)$$

- Randomly mutate each position of each state  $s_i$  with a small probability (mutation rate).

$$s'_i [k] = \begin{cases} s_i [k] & \text{with probability } 1 - q \\ \text{random} & \text{with probability } q \end{cases}, k = 1, 2, \dots, m$$

- Repeat with population  $s'$ .

# Variations

## Discussion

- Parents can survive.
- Use ranking instead of  $F(s)$  to compute reproduction probabilities.
- Cross over random bits instead of chunks.

# Genetic Algorithm Performance

## Discussion

- Use hill-climbing first.
- State design is the most important.
- In theory, cross over is much more efficient than mutation.

# Game Tree

## Motivation

- The initial state is the beginning of the game.
- There are no goal states, but there are multiple terminal states in which the game ends.
- Each successor of a state represents a feasible action (or a move) in the game.
- The search problem is to find the terminal state with the lowest cost (or usually the highest reward).

# Adversarial Search

## Motivation

- The main difference between finding solutions of games and standard search problems or local search problems is that part of the search is performed by an opponent adversarially.
- Usually, the opponent wants to maximize the cost or minimize the reward from the search. This type of search problems is called adversarial search.
- In game theory, the solution of a game is called an equilibrium. It is a path in which both players do not want to change actions.

# Backward Induction

## Motivation

- Games are usually solved backward starting from the terminal states.
- Each player chooses the best action (successor) given the (already solved) optimal actions of all players in the subtrees (called subgames).