CS540 Introduction to Artificial Intelligence Lecture 19

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Based on lecture slides by Jerry Zhu, Yingyu Liang, and Charles Dyer

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Summary Discussion

- Search:
- Uninformed.
- Informed.
- Solution Local Search: Hill Climbing (Valley Finding).
- Iccal Search: Simulated Annealing.
- **o** Local Search: Genetic Algorithm.
- Adversarial.

Genetic Algorithm Description

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

$$p_{i} = \frac{F(s_{i})}{\sum_{i=1}^{N} F(s_{i})}, i = 1, 2, ..., N$$

• 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.

 $\begin{array}{c} (11010,00101) \ , (10010,01101) \\ (10110,01001) \ , (10100,01011) \end{array}$

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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

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Cross Over, Modifications

- 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.

Genetic Algorithm

Mutation, Modifications

• For specific problems, there are ways other than flipping bits to mutate a state.

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- Two-swap: ABCDE to EBCDA
- Two-interchange: ABCDE to EDCBA

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.

$$\textit{s}_1,\textit{s}_2,...,\textit{s}_N$$

• Compute the reproduction probability.

$$p_{i} = \frac{F(s_{i})}{\sum_{j=1}^{N} F(s_{j})}, i = 1, 2, ..., N$$

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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).

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

• Repeat with population s'.

Variations Discussion

- Parents can survive.
- Use ranking instead of F(s) to compute reproduction probabilities.

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• Cross over random bits instead of chunks.

Genetic Algorithm

Genetic Algorithm Performance Discussion

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

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Summary Discussion

- Search:
- Uninformed.
- Informed.
- Local Search: Hill Climbing (Valley Finding): Start at a random state → Move to the best successor → Repeat.
- Local Search: Simulated Annealing: Start at a random state
 → Generate a random successor → Move if better, Move with
 small probability if worse → Repeat.
- Search: Genetic Algorithm: Start with many random states → Cross-over according to fitness → Mutation → Repeat.
- Adversarial.