

Running AGP2 iteration 1/20

Author Annotation: Elitism

```
def selection(population, fitnesses):
```

```
    """
```

```
    Selects parents for crossover based on their fitness.
```

```
    Parameters:
```

```
    population (list): A list of individuals.
```

```
    fitnesses (list): A list of fitness values corresponding to the individuals in the population.
```

```
    Returns:
```

```
    individual: An individual from the population selected as a parent.
```

```
    """
```

```
    # Combine the population and fitnesses into a numpy array for efficient indexing
```

```
    individuals = np.array(population)
```

```
    # Select the index of the best individual based on its fitness (minimum error)
```

```
    idx_best = np.argmin(fitnesses, axis=0)[0]
```

```
    # Return the best individual as a parent
```

```
    return individuals[idx_best]
```

Running AGP2 iteration 2/20

Author Annotation: Elitism

```
def selection(population, fitnesses):
```

```
    """
```

Selects an individual from a population based on its fitness.

Parameters:

- population (list): A list of individuals in the current population.
- fitnesses (list): A list of fitness values corresponding to each individual in the population.

Returns:

- selected_individual (individual): An individual chosen for reproduction based on its fitness value.

```
    """
```

```
    # Combine the population and their fitness values into a list of tuples
```

```
    population_fitness = list(zip(population, fitnesses))
```

```
    # Sort the population by fitness in ascending order
```

```
    sorted_population = sorted(population_fitness, key=lambda x: np.mean(x[1]))
```

```
    # Select the individual with the best fitness (i.e., minimum error)
```

```
    selected_individual = sorted_population[0][0]
```

```
return selected_individual
```

Running AGP2 iteration 3/20

Author Annotation: Elitism

```
def selection(population, fitnesses):  
    # Calculate the total errors for each individual in the population  
    total_errors = [sum(fit) for fit in fitnesses]  
  
    # Choose the index of the fittest individual with the lowest total error  
    best_index = np.argmin(total_errors)  
  
    return population[best_index]
```

Running AGP2 iteration 4/20

Author Annotation: Elitism

```
def selection(population, fitnesses):  
    # Calculate the total errors of each individual in the population  
    total_errors = [sum(f) for f in fitnesses]  
  
    # Select the indices of individuals with the lowest total error  
    min_error_idx = np.argmin(total_errors)
```

```
# Select one individual from those with the lowest total error based on their original order
of appearance in the population
```

```
# (to avoid favoring some over others due to random initialization or early convergence)
```

```
return np.random.choice([individual for idx, individual in enumerate(population) if idx ==
min_error_idx])
```

Running AGP2 iteration 5/20

Author Annotation: Elitism

```
import numpy as np
```

```
def selection(population, fitnesses):
```

```
    # Convert lists to numpy arrays for efficient computation
```

```
    population = np.array(population)
```

```
    fitnesses = np.array(fitnesses)
```

```
    # Select the individual with the best fitness (lowest error)
```

```
    selected_individual_index = np.argmin(fitnesses, axis=0)[0]
```

```
    return population[selected_individual_index]
```

Running AGP2 iteration 6/20

Author Annotation: Elitism

```
def selection(population, fitnesses):
```

```
    """
```

Select parent from population based on fitness.

Args:

population (list): List of individual in the population.

fitnesses (list): List of fitness values corresponding to individuals in the population.

Returns:

The selected parent as an individual from the population.

"""

Combine population and fitnesses into a list of tuples for easier manipulation

pop_with_fitness = list(zip(population, fitnesses))

Sort the population based on fitness (lower is better)

sorted_pop_with_fitness = sorted(pop_with_fitness, key=lambda x: np.mean(x[1]))

Select first 10 individuals with lowest fitness as parents

parents = [x[0] for x in sorted_pop_with_fitness[:10]]

Select one parent randomly from the selected parents

parent = np.random.choice(parents)

return parent

Running AGP2 iteration 7/20

Author Annotation: Elitism

```
def selection(population, fitnesses):
```

```
    """
```

```
    Selects the fittest individual from a population based on their fitness.
```

```
    Args:
```

```
        population (list): A list of individuals in the GP system.
```

```
        fitnesses (list): A list of fitness values corresponding to each individual in the
    population.
```

```
    Returns:
```

```
        Individual: The fittest individual in the population.
```

```
    """
```

```
    # Combine the population and fitnesses into a single data structure for easier access
```

```
    combined = [(population[i], fitnesses[i]) for i in range(len(population))]
```

```
    # Sort the population by their fitness (lower is better)
```

```
    combined.sort(key=lambda x: np.mean(x[1]))
```

```
    # Select the fittest individual as the parent for reproduction
```

```
    return combined[0][0]
```

Running AGP2 iteration 8/20

Author Annotation: Elitism

```
import numpy as np
```

```
def selection(population, fitnesses):
```

```
    """
```

```
    Parent selection for Genetic Programming.
```

```
    Args:
```

```
        population (list): A list of individuals.
```

```
        fitnesses (list): A list of fitness values corresponding to each individual in the
        population.
```

```
    Returns:
```

```
        Individual: The selected parent.
```

```
    """
```

```
    # Combine individuals and their fitnesses into a list of tuples
```

```
    individuals = list(zip(population, fitnesses))
```

```
    # Sort individuals based on their fitness values
```

```
    sorted_individuals = sorted(individuals, key=lambda x: np.sum(x[1]))
```

```
    # Select the individual with the lowest fitness value (i.e., highest probability)
```

```
    selected_parent = sorted_individuals[0][0]
```

```
    return selected_parent
```

Author Annotation: Elitism

```
def selection(population, fitnesses):
```

```
    """
```

```
    Parent Selection Function for Genetic Programming.
```

```
    Parameters:
```

```
    population (list): A list of individuals in the current generation.
```

```
    fitnesses (list): A list of fitness values corresponding to each individual in the population.
```

```
    Returns:
```

```
    individual: An individual from the population selected for reproduction.
```

```
    """
```

```
    # Combine population and fitnesses into a single data structure
```

```
    combined = np.column_stack((population, fitnesses))
```

```
    # Select the index of the best individual based on minimum total error
```

```
    best_index = np.argmin(combined[:, 1])
```

```
    # Return the best individual
```

```
    return population[best_index]
```

Running AGP2 iteration 10/20

Author Annotation: Fitness-proportionate


```

def selection(population, fitnesses):

    # Calculate the cumulative probability distribution of the population fitnesses.
    probabilities = np.cumsum(fitnesses)

    # Generate a random number between 0 and 1 to simulate a roulette wheel spin.
    select_prob = np.random.uniform(0.0, 1.0)

    # Perform binary search on the cumulative probability distribution to find the individual
    that would be selected by this spin of the roulette wheel.

    lo = 0
    hi = len(population) - 1
    while lo < hi:
        mid = (lo + hi) // 2
        if probabilities[mid] <= select_prob:
            lo = mid + 1
        else:
            hi = mid

    # Return the individual that was selected by this spin of the roulette wheel.
    return population[lo-1]

```

Running AGP2 iteration 11/20

Author Annotation: Elitism

import numpy as np

```

def selection(population, fitnesses):
    # Combine population and fitnesses into a list of tuples
    combined = list(zip(population, fitnesses))

    # Sort the combined list based on fitness
    sorted_combined = sorted(combined, key=lambda x: sum(x[1]), reverse=False)

    # Select the best individual (with lowest fitness) as the parent
    best_individual = sorted_combined[0][0]

    return best_individual

```

Running AGP2 iteration 12/20

Author Annotation: Fitness-proportionate

```

def selection(population, fitnesses):

```

```

    """

```

Selects parents from a population based on their fitness.

Args:

population: A list of individuals, each represented as a tree-based structure.

fitnesses: A list of fitness values corresponding to the individuals in the population.

Returns:

An individual selected for breeding.

"""

Calculate the total fitness for each individual (assuming we want to maximize it)

total_fitness = [sum(fit) for fit in fitnesses]

Calculate the cumulative sum of the total fitness

cum_sum = np.cumsum(total_fitness)

Generate a random number between 0 and the sum of all total fitness values

rand_num = np.random.uniform(0, cum_sum[-1])

Find the index of the first cumulative sum that is greater than or equal to the random number

idx = np.where(cum_sum >= rand_num)[0][0]

return population[idx]

Running AGP2 iteration 13/20

Author Annotation: Elitism

def selection(population, fitnesses):

"""

Selects an individual for reproduction based on its fitness.

Parameters:

population (list): List of individuals.

fitnesses (list): Corresponding list of fitness values for each individual.

Returns:

individual: One of the elements in the population.

"""

Calculate the total number of errors across all test cases

num_errors = [sum(f) for f in fitnesses]

Find the index of the individual with the lowest error (i.e., best fitness)

min_error_idx = np.argmin(num_errors)

Select this individual for reproduction

return population[min_error_idx]

Running AGP2 iteration 14/20

Author Annotation: Elitism

def selection(population, fitnesses):

Calculate the error for each individual in the population

errors = [sum(f) for f in fitnesses]

Use tournament selection to select an individual from the population

best_individual_index = np.argmin(errors)

```
return population[best_individual_index]
```

Running AGP2 iteration 15/20

Author Annotation: Elitism

```
def selection(population, fitnesses):
```

```
    """
```

Select parents for crossover and mutation based on their fitness.

Parameters:

population (list): List of individuals.

fitnesses (list): Corresponding list of fitness values.

Returns:

individual: Chosen parent from the population.

```
    """
```

```
# Combine individuals and their fitnesses into a list of tuples
```

```
ind_fit = list(zip(population, fitnesses))
```

```
# Sort the list in ascending order based on fitness
```

```
ind_fit.sort(key=lambda x: np.mean(x[1]))
```

```
# Select the best individual (with the minimum mean error)
```

```
chosen_parent = ind_fit[0][0]
```

```
return chosen_parent
```

Running AGP2 iteration 16/20

Author Annotation: Elitism

```
def selection(population, fitnesses):
```

```
    """
```

Selects a parent for crossover based on the tournament selection method.

Parameters:

population (list): A list of individuals in the current generation.

fitnesses (list): A list of lists where each sublist contains the errors
on test cases for an individual in the population.

Returns:

individual: The selected parent from the population.

```
    """
```

```
# Initialize variables to keep track of the best individual and its fitness
```

```
best_of_run_f = float('inf')
```

```
best_individual = None
```

```
# Iterate over each individual in the population
```

```
for i, individual in enumerate(population):
```

```
    # Get the fitness of the current individual
```

```
    individual_fitness = fitnesses[i]
```

```
    # Check if this is a better individual than the current best
```

```
if np.mean(individual_fitness) < np.mean(best_of_run_f):  
    best_individual = individual  
    best_of_run_f = np.mean(individual_fitness)  
  
return best_individual
```

Running AGP2 iteration 17/20

Author Annotation: Elitism

```
def selection(population, fitnesses):  
    # Combine population and fitness into a list of tuples for easier manipulation  
    individuals = list(zip(population, fitnesses))  
  
    # Sort the individuals based on their fitness (lower is better)  
    individuals.sort(key=lambda x: np.sum(x[1]))  
  
    # Select the best individual  
    selected_individual = individuals[0]  
  
    return selected_individual[0] # Return the individual (not its fitness)
```

Running AGP2 iteration 18/20

Author Annotation: Elitism

```
def selection(population, fitnesses):
```

```
"""
```

Selects parents for the next generation based on their fitness.

Parameters:

population (list): A list of individuals in the current population.

fitnesses (list): A list of fitness values corresponding to each individual in the population.

Returns:

Individual: The selected parent, i.e., one of the elements in the population.

```
"""
```

```
# Combine population and fitness into a numpy array for easier manipulation
```

```
pop_fit = np.column_stack((population, fitnesses))
```

```
# Get indices of individuals with the best fitness (smallest error)
```

```
best_of_run_f = min(fitnesses)
```

```
best_indices = np.where(pop_fit[:, 1:] == best_of_run_f)[0]
```

```
# Select one parent at random from the ones with the best fitness
```

```
parent_idx = np.random.choice(best_indices, size=1)[0]
```

```
return population[parent_idx]
```

Running AGP2 iteration 19/20

Author Annotation: Elitism

```
def selection(population, fitnesses):
```

```
"""
```


Selects a parent from the population based on their fitness.

Args:

population (list): A list of individuals in the current generation.

fitnesses (list): A list of corresponding fitness values for each individual in the population.

Returns:

The selected individual with the best fitness.

"""

Combine the population and fitness lists into a list of tuples

population_fitness = list(zip(population, fitnesses))

Sort the combined list based on the fitness value (lower is better)

sorted_population_fitness = sorted(population_fitness, key=lambda x: np.mean(x[1]))

Select the individual with the best fitness (i.e., the first element in the sorted list)

selected_individual, _ = sorted_population_fitness[0]

return selected_individual

Running AGP2 iteration 20/20

Author Annotation: Elitism

def selection(population, fitnesses):

Combine population and fitnesses into a list of tuples for easier manipulation

```
individuals = list(zip(population, fitnesses))
```

```
# Sort the population by their fitness values in ascending order
```

```
individuals.sort(key=lambda x: np.sum(x[1]))
```

```
# Select the individual with the best fitness
```

```
return individuals[0][0]
```