

COLUMBIA UNIVERSITY

MECE 4510 EVOLUTIONARY COMPUTATION AND DESIGN AUTOMATION

Symbolic Regression

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Grace Hours Used: 53
Grace Hours Remaining: 43

October 22, 2018

1 Result Summary

| Result Table | | |
|--|-------------------|------------------|
| Method | Evaluation Number | Best Error (MAE) |
| Random Search | 300000 | 0.0847 |
| Hill Climber | 300000 | 0.0863 |
| GP (Conventional Selection) | 14000 | 0.00002537 |
| GP (Conventional Selection) with Larger Population | 43000 | 0.00002537 |
| GP (Deterministic Crowding) | 129000 | 0.00002537 |

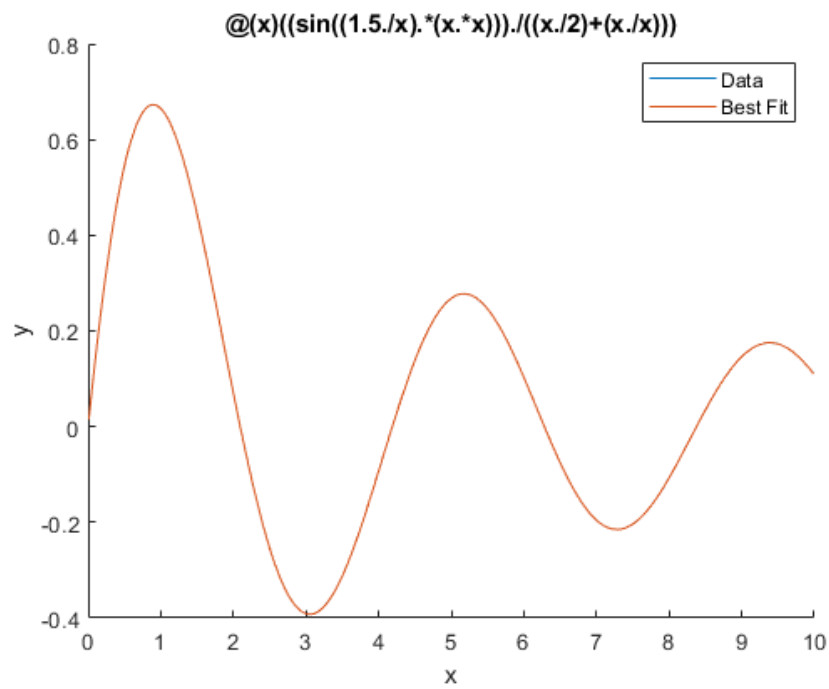


Figure 1: Performance Plot

After simplification, the function can be expressed as:

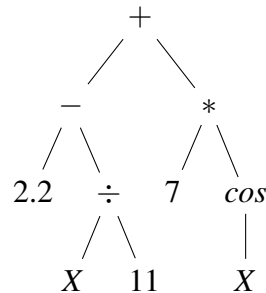
$$f(x) = \sin(1.5x)/(0.5x + 1) \quad (1)$$

2 Methods

For this homework, we have been asked to use genetic programming to perform symbolic regression. Our goal is to find the symbolic algebraic expression in the form of $y = f(x)$ that best fits a set of given 1000 (x,y) points. Assume the symbolic regression only uses algebraic operators: x , $-$, $*$, $/$, *sine and cosine*, and real constants (in the range ± 10), and the variable x .

2.1 Representation

The key part of genetic programming is to convert the program into a high level tree structure compare to genetic programming which the program was converted into simpler chromosome type. The tree structure is more powerful in terms of computer programming since trees can be easily evaluated in a recursive manner. For example, a function can be represented as a tree structure as following:



can represent the equation

$$(2.2 - \frac{x}{11}) + (7 * \cos(x)) \quad (2)$$

For this assignment, we assume the maximum depth of the tree structure is less and equal than 5.

2.2 Random Search

Once we have our representation setup, the random search algorithm is straight forward. For each evaluation of the random search algorithm, a function tree is generated with random depth from two to five. The random search is the baseline for performance comparison between the hill climber algorithm and our genetic programmings.

2.3 Random Hill Climber

A Random Hill Climber is basically a random search with simple decision making capability. Between each evaluation, the mutation process is applied to the function. During the mutation process, it will randomly select a valid mutation point, generate a new sub-tree and replace into new function tree. In my implementation, the probability decide whether it always go to a better solution or not. Sometime go to a worse solution can avoid hill climber stuck at the local maximum.

2.4 Genetic Programming with Variations

Similar to Evolutionary Algorithm, we can implement following types of operators into the Genetic Programming: selection, crossover, and mutation. For variation, Deterministic Crowding and different size of population was applied during my implementation. Here are some details about the techniques I used in my implementation:

- **Selection:** For conventional selection Genetic Programming, during each generation, a default of 50% parents will be selected to generate 50 offsprings.
- **Crossover:** The crossover was applied into both conventional selection method and deterministic crowding method. The algorithm will first choose two valid crossover node from each parent and switch the subtrees between two parents in order to generate two offSprings.
- **Mutation:** During the mutation process, the algorithm will first randomly pick a valid node for mutation, then generate a new tree based on the depth of the mutation node.
- **Deterministic Crowding:** In order to maintain the diversity for genetic programming, we need some methods to maintain useful diversity for genetic programming to work better. Crowding is one of the popular method, it only replace individuals that are similar. More specific, deterministic crowding compare the similarity between two parents and two offsprings, replace the one has higher similarity.

Algorithm 1 Deterministic Crowding

```
1: procedure MYPROCEDURE
2:   if  $d(p_1, c_1) + d(p_2, c_2) < d(p_1, c_2) + d(p_2, c_1)$  then
3:     compare  $c_1$  to  $p_1$  and  $c_2$  to  $p_2$  and replace parents if offspring better
4:   else
5:     compare  $c_1$  to  $p_2$  and  $c_2$  to  $p_1$  and replace parents if offspring better
```

2.5 Analysis of Performance

Overall, all GPs performed very well since they all can find the optimal solution. However, the random search and hill climber did not perform as expected. The main reason for the failure of hill climber because it was badly guided. Like random search, hill climber would perform well if it is lucky. On the other hand, the deterministic crowding for genetic programming is extremely powerful. From both performance plot as well as the diversity plot, the crowding helped to maintain diversity in a better manner compare to conventional selection. Therefore, on the performance plot, the crowding starts slow than conventional selection, but it could find better solution eventually.

3 Performance

3.1 Performance Plots

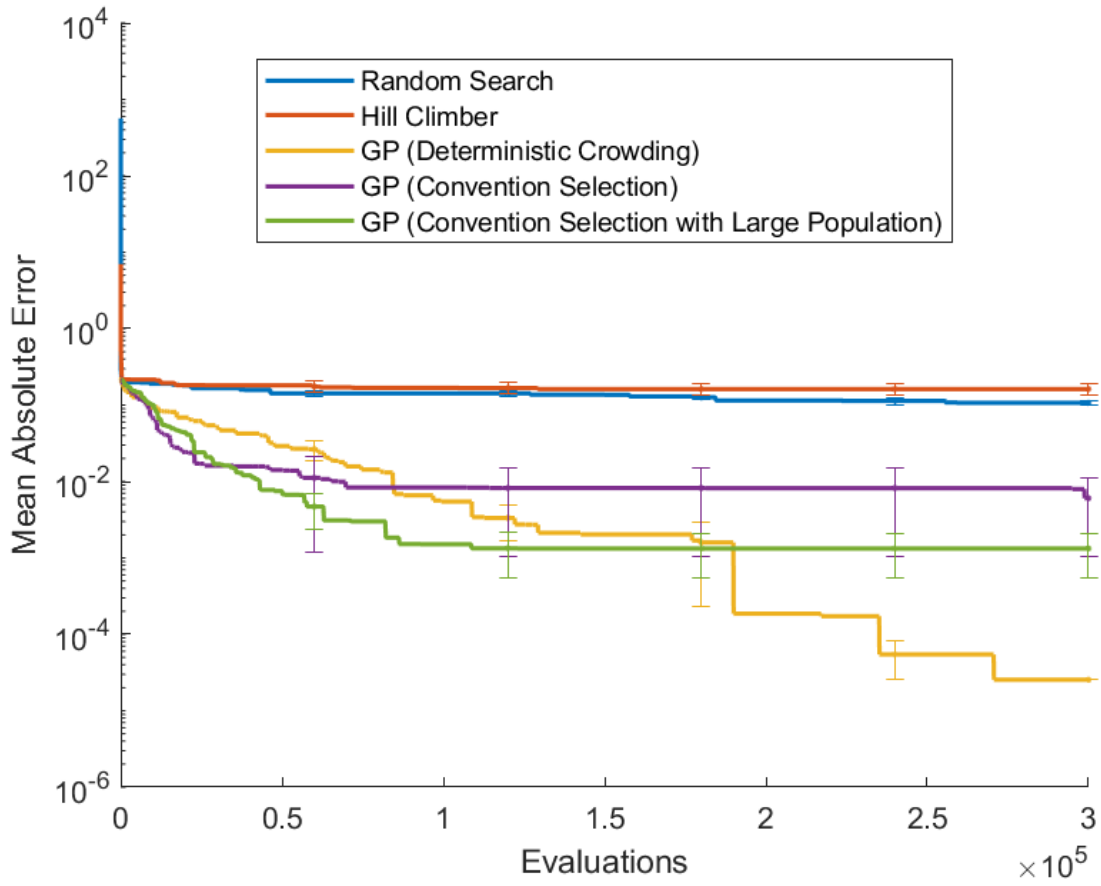


Figure 2: Performance Plot

3.2 Dot Plot

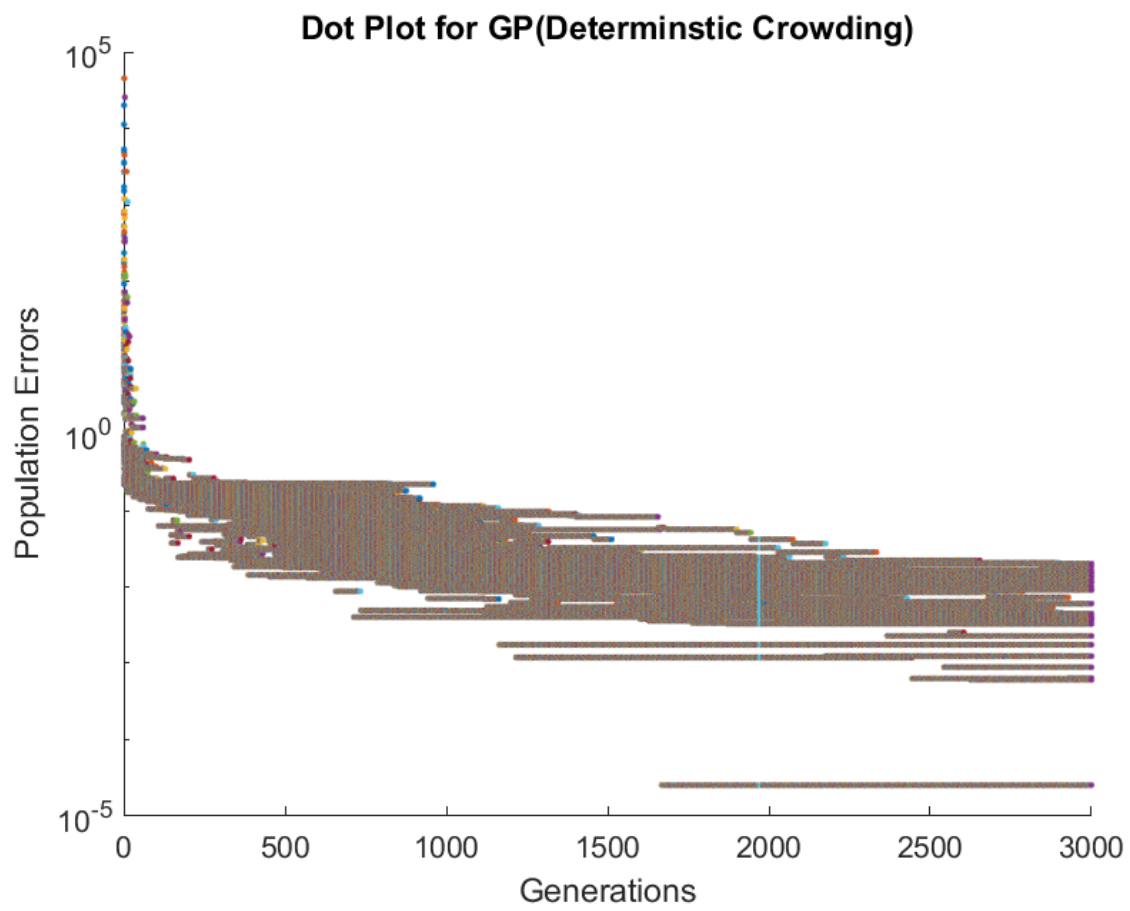


Figure 3: Dot Plot

3.3 Diversity Plot

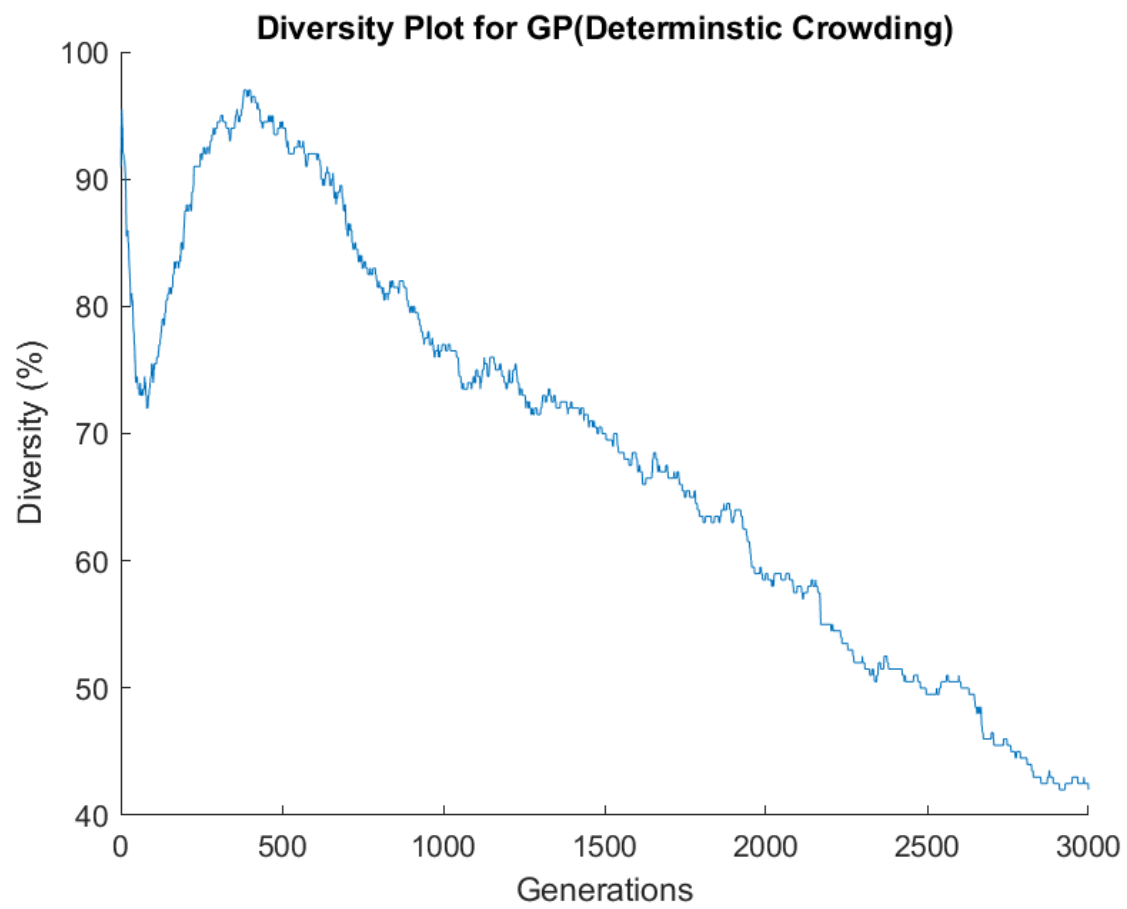


Figure 4: Diversity Plot

3.4 Convergence Plot

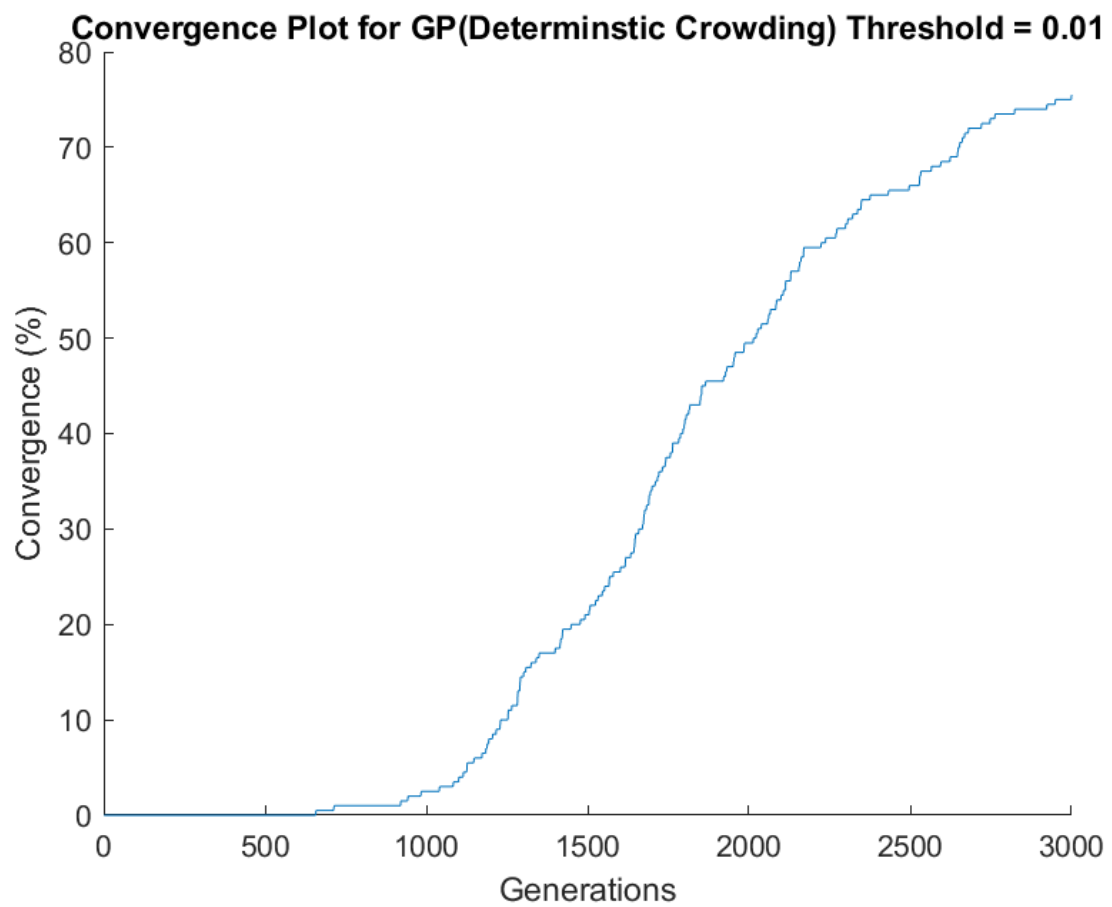


Figure 5: Convergence Plot

3.5 Simpler Problem Tested

During the debugging process of Genetic Programming, simpler problem was tested to ensure the GP can run without any bug. For example, a test data set of $\sin(x)$ was used. In addition, all GPs can find the optimal solution within 10 generations.

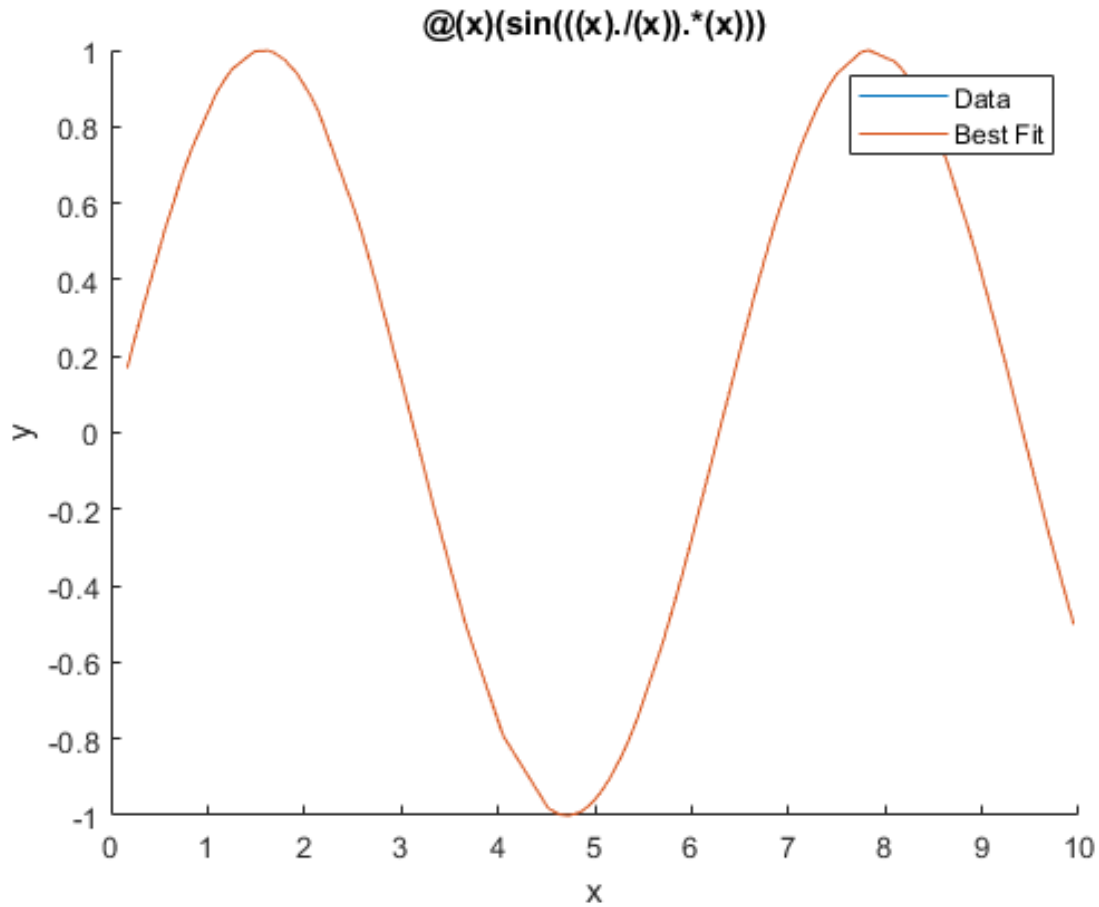


Figure 6: Simple Problem Test

3.6 Validation

The validation plot did not show much difference between training data and testing data since the data we were given has negligible noise.

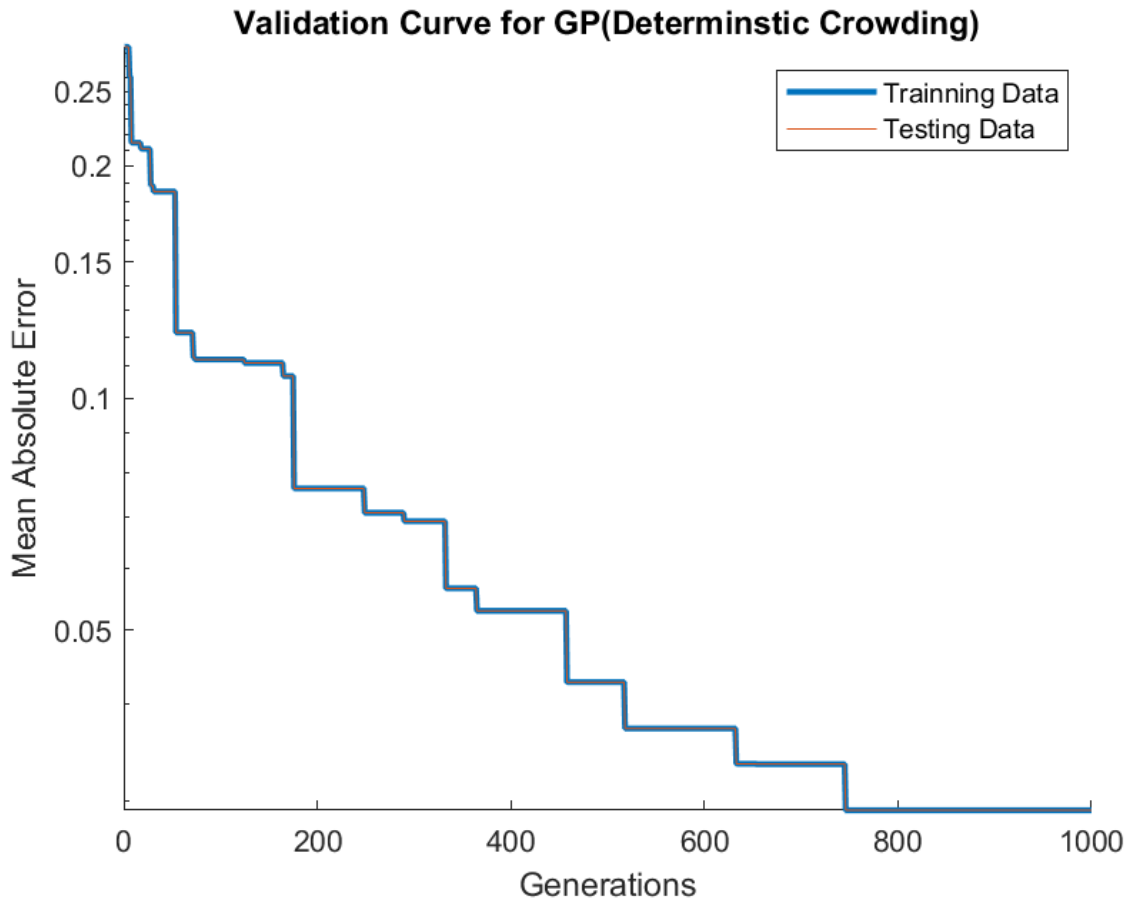


Figure 7: Validation Plot

4 Appendix

```
1 function [] = HW2()
2 operator = {'+', '-', '.*', './', 'sin', 'cos'};
3 varConst = {'c', 'x'};
4 maxLevel = 4;
5 tic
6 rng('shuffle')
7 data = csvread('function1.csv');
8 %trainingIndex = sort(randperm(1000, 100));
9 trainingIndex = 1:500;
10 trainingData = data(trainingIndex,:);
11 validationData = data(setdiff(1:length(data), trainingIndex),:);
12 randomSearch(operator, varConst, maxLevel, trainingData, 300000, 4);
13 hillClimber(operator, varConst, maxLevel, trainingData, 300000, 4);
14 GP1(operator, varConst, maxLevel, trainingData, validationData, 200, 3000, 4, 'true', 'dot');
15 GP2(operator, varConst, maxLevel, trainingData, 200, 3000, 4, 'true');
16 GP2_LP(operator, varConst, maxLevel, trainingData, 400, 1500, 4, 'true');
17 toc
18 end
19
20 function [] = GP2_LP(operator, varConst, maxLevel, data, popSize, n, repeat, print)
21 parfor r = 1:repeat
22     if strcmp(print, 'true')
23         fileID = fopen(strcat('GP2_LP_', int2str(r), '.txt'), 'wt');
24     end
25     population = cell([popSize, 2^maxLevel]);
26     for ii = 1:popSize
27         population(ii, 1:end-1) = heapGeneration(operator, varConst, maxLevel);
28     end
29     popError = zeros(popSize, 1);
30     bestError = inf;
31     for i = 1:n
32         for ii = 1:popSize
33             population(ii, 2^maxLevel) = MAE_Cal(population(ii, 1:end-1), data);
34             popError(ii) = population(ii, 2^maxLevel);
35             %heapString(population(ii, 1:end-1))
36         end
37         currentError = min(popError);
38         population = sortrows(population, 2^maxLevel);
39         selectPop = population(1:popSize/2, 1:end-1);
40         offSpringPop = cell([popSize/2, 2^maxLevel-1]);
41         bestError = min(currentError, bestError);
42         if strcmp(print, 'true')
43             fprintf(fileID, '%d %10.8f %10.8f \n', i*popSize/2, currentError, bestError);
44         end
45         for jj = 1:popSize/4
46             parent1 = selectPop(randi(length(selectPop)), :);
47             parent2 = selectPop(randi(length(selectPop)), :);
48             [offSpring1, offSpring2] = crossover(parent1, parent2, maxLevel);
49             offSpring1 = mutation(offSpring1, operator, varConst, maxLevel);
50             offSpring2 = mutation(offSpring2, operator, varConst, maxLevel);
51             offSpringPop(jj*2-1:jj*2, :) = [offSpring1; offSpring2];
52         end
53         population(:, 1:end-1) = [selectPop; offSpringPop];
54     end
55     population = population(:, 1:end-1);
56     for i = 1:popSize
57         popError(i) = MAE_Cal(population(i, :), data);
58     end
59     % bestHeap = population(1, :);
60     % bestHeapString = heapString(bestHeap);
61     % figure
62     % hold on
63     % plot(data(:, 1), data(:, 2))
64     % plot(data(:, 1), evalHeap(bestHeap, data))
65     % legend('Data', 'Best Fit')
```

```

66 %     xlabel('x')
67 %     ylabel('y')
68 %     title(bestHeapString)
69 index = find(popError == min(popError));
70 fprintf("GP2: Eval: %d, Error: %10.8f.\n", n, popError(index(1)))
71 end
72 end
73
74 function [] = GP2(operator, varConst, maxLevel, data, popSize, n, repeat, print)
75 parfor r = 1:repeat
76     if strcmp(print, 'true')
77         fileId = fopen(strcat('GP2_', int2str(r), '.txt'), 'wt');
78     end
79     population = cell([popSize, 2^maxLevel]);
80     for ii = 1:popSize
81         population(ii, 1:end-1) = heapGeneration(operator, varConst, maxLevel);
82     end
83     popError = zeros(popSize, 1);
84     bestError = inf;
85     for i = 1:n
86         for ii = 1:popSize
87             population(ii, 2^maxLevel) = MAE_Cal(population(ii, 1:end-1), data);
88             popError(ii) = population(ii, 2^maxLevel);
89             %heapString(population(ii, 1:end-1))
90         end
91         currentError = min(popError);
92         population = sortrows(population, 2^maxLevel);
93         selectPop = population(1:popSize/2, 1:end-1);
94         offSpringPop = cell([popSize/2, 2^maxLevel-1]);
95         bestError = min(currentError, bestError);
96         if strcmp(print, 'true')
97             fprintf(fileId, '%d %10.8f %10.8f \n', i*popSize/2, currentError, bestError);
98         end
99         for jj = 1:popSize/4
100             parent1 = selectPop(randi(length(selectPop)), :);
101             parent2 = selectPop(randi(length(selectPop)), :);
102             [offSpring1, offSpring2] = crossOver(parent1, parent2, maxLevel);
103             offSpring1 = mutation(offSpring1, operator, varConst, maxLevel);
104             offSpring2 = mutation(offSpring2, operator, varConst, maxLevel);
105             offSpringPop(jj*2-1:jj*2, :) = [offSpring1; offSpring2];
106         end
107         population(:, 1:end-1) = [selectPop; offSpringPop];
108     end
109     population = population(:, 1:end-1);
110     for i = 1:popSize
111         popError(i) = MAE_Cal(population(i, :), data);
112     end
113     %     bestHeap = population(1, :);
114     %     bestHeapString = heapString(bestHeap);
115     %     figure
116     %     hold on
117     %     plot(data(:, 1), data(:, 2))
118     %     plot(data(:, 1), evalHeap(bestHeap, data))
119     %     legend('Data', 'Best Fit')
120     %     xlabel('x')
121     %     ylabel('y')
122     %     title(bestHeapString)
123     index = find(popError == min(popError));
124     fprintf("GP2: Eval: %d, Error: %10.8f.\n", n, popError(index(1)))
125 end
126 end
127
128 function [] = GP1(operator, varConst, maxLevel, data, validationData, popSize, n, repeat, print,
    type)
129 dotData = zeros(n, popSize);
130 errorAndValidation = zeros(n, 2);
131 for r = 1:repeat
132     if strcmp(print, 'true')

```

```

133     fileID = fopen(strcat('GP1_',int2str(r),'.txt'),'wt');
134 end
135 population = cell([popSize,2^maxLevel-1]);
136 for ii = 1:popSize
137     population(ii,:) = heapGeneration(operator, varConst, maxLevel);
138 end
139 popError = zeros(popSize,1);
140 validationError = zeros(popSize,1);
141 bestError = inf;
142 for i = 1:n
143     for ii = 1:popSize
144         popError(ii) = MAE_Cal(population(ii,:),data);
145         validationError(ii) = MAE_Cal(population(ii,:),validationData);
146         %heapString(population(ii,:))
147     end
148     mean(popError,'omitnan');
149     currentError = sort(popError);
150     validationError = sort(validationError);
151     bestError = min(currentError(1),bestError);
152     errorAndValidation(i,:) = [currentError(1) validationError(1)];
153     if strcmp(print,'true')
154         fprintf(fileID, '%d %10.8f %10.8f \n', i*popSize/2, currentError(1), bestError);
155     end
156     if strcmp(type,'dot')
157         dotData(i,:) = popError;
158     end
159     % random select two parents
160     for j = 1:popSize/2
161         index = randperm(popSize,2);
162         parent1 = population(index(1),:);
163         parent2 = population(index(2),:);
164         % crossover to get two offsprings
165         [offSpring1, offSpring2] = crossOver(parent1, parent2, maxLevel);
166         % perfome mutation
167         offSpring1 = mutation(offSpring1, operator, varConst, maxLevel);
168         offSpring2 = mutation(offSpring2, operator, varConst, maxLevel);
169         % perfome deterministic crowding
170         dp1c1 = similarCal(parent1, offSpring1, data);
171         dp2c2 = similarCal(parent2, offSpring2, data);
172         dp1c2 = similarCal(parent1, offSpring2, data);
173         dp2c1 = similarCal(parent2, offSpring1, data);
174         c1 = MAE_Cal(offSpring1, data); p1 = MAE_Cal(parent1, data);
175         c2 = MAE_Cal(offSpring2, data); p2 = MAE_Cal(parent2, data);
176         if dp1c1+dp2c2 < dp1c2+dp2c1
177             if c1 < p1
178                 population(index(1),:) = offSpring1;
179             end
180             if c2 < p2
181                 population(index(2),:) = offSpring2;
182             end
183         else
184             if c1 < p2
185                 population(index(2),:) = offSpring1;
186             end
187             if c2 < p1
188                 population(index(1),:) = offSpring2;
189             end
190         end
191     end
192 end
193 for i = 1:popSize
194     popError(i) = MAE_Cal(population(i,:),data);
195 end
196 % sortError = sort(popError);
197 % [~,errorIndex] = ismember(sortError,popError);
198 % sortPop = population(errorIndex,:);
199 % bestHeap = sortPop(1,:);
200 % bestHeapString = heapString(bestHeap);

```

```

201 % figure
202 % hold on
203 % plot(data(:,1),data(:,2))
204 % plot(data(:,1),evalHeap(bestHeap, data))
205 % legend('Data','Best Fit')
206 % xlabel('x')
207 % ylabel('y')
208 % title(bestHeapString)
209 index = find(popError == min(popError));
210 fprintf("GP1: Eval: %d, Error: %10.8f.\n", n, popError(index(1)))
211 save('dotData','dotData')
212 save('errorAndValidation2','errorAndValidation')
213 end
214 end
215
216 function sim = similarCal(heap1, heap2, data)
217 sim = abs(MAE_Cal(heap1,data)-MAE_Cal(heap2,data));
218 end
219
220 function [offSpring1, offSpring2] = crossover(parent1, parent2, maxLevel)
221 maxCrossOverLimit = min(max(find(~cellfun(@isempty,parent1))),max(find(~cellfun(@isempty,parent2)
    )));
222 offSpring1 = parent1;
223 offSpring2 = parent2;
224 maxCrossOverLevel = floor(log2(maxCrossOverLimit))+1;
225 while 1
226 crossOverLevel = randi(maxCrossOverLevel);
227 selection = 2^(crossOverLevel-1):2^(crossOverLevel)-1;
228 crossOverPoint1 = selection(randi(length(selection)));
229 crossOverPoint2 = selection(randi(length(selection)));
230 if ~isempty(parent1{crossOverPoint1}) && ~isempty(parent2{crossOverPoint2})
231     if isnumeric(parent1{crossOverPoint1}) && isnumeric(parent2{crossOverPoint2})
232         break
233     elseif strcmp(parent1{crossOverPoint1},'x') && strcmp(parent2{crossOverPoint2},'x')
234         break
235     elseif isnumeric(parent1{crossOverPoint1}) && strcmp(parent2{crossOverPoint2},'x')
236         break
237     elseif isnumeric(parent2{crossOverPoint2}) && strcmp(parent1{crossOverPoint1},'x')
238         break
239     elseif ischar(parent1{crossOverPoint1}) && ischar(parent2{crossOverPoint2})
240         break
241     end
242 end
243 end
244 crossOverLocations1 = searchChildren(crossOverPoint1, maxLevel);
245 crossOverLocations2 = searchChildren(crossOverPoint2, maxLevel);
246 subHeap1 = parent1(crossOverLocations1);
247 subHeap2 = parent2(crossOverLocations2);
248 for i = 1:length(crossOverLocations1)
249     offSpring1(crossOverLocations1(i)) = subHeap2(i);
250     offSpring2(crossOverLocations2(i)) = subHeap1(i);
251 end
252 end
253
254 function [] = hillClimber(operator, varConst, maxLevel, data, n, repeat)
255 parfor r = 1:repeat
256     fileID = fopen(strcat('RMHC_',int2str(r),'.txt'),'wt');
257     heap = heapGeneration(operator, varConst, maxLevel);
258     oldError = MAE_Cal(heap,data);
259     bestError = inf;
260     for i = 1:n
261         newHeap = mutation(heap, operator, varConst, maxLevel);
262         newError = MAE_Cal(newHeap, data);
263         if newError < oldError
264             heap = newHeap;
265             oldError = newError;
266             bestError = newError
267         end

```

```

268         fprintf(fileID, '%d %10.8f %10.8f \n', i, newError, bestError);
269     end
270     fprintf("RMHC: Eval: %d, Error: %10.8f.\n", n, bestError)
271 end
272 end
273
274 function heap = mutation(heap, operator, varConst, maxLevel)
275 % the mutation should be able happen non-empty node
276 mutateIndex = find(~cellfun(@isempty,heap));
277 mutationPoint = mutateIndex(randi(length(mutateIndex)-1)+1);
278 mutateLocations = searchChildren(mutationPoint, maxLevel);
279 % determine the mutation level
280 mutationLevel = floor(log2(mutationPoint))+1;
281 % delete the original heap nodes
282 for i = 1:length(mutateLocations)
283     heap{mutateLocations(i)} = [];
284 end
285 if mutationLevel == maxLevel
286     heap(mutateLocations) = varConst(randi(length(varConst)));
287 else
288     % generate a subtree
289     subHeap = heapGeneration(operator, varConst, (maxLevel-mutationLevel+1));
290     for i = 1:length(mutateLocations)
291         heap(mutateLocations(i)) = subHeap(i);
292     end
293 end
294 % replace c with constant
295 heap = replaceC(heap);
296 end
297
298 function operateLocations = searchChildren(operatPoint, maxLevel)
299 searchQueue = [operatPoint];
300 operateLocations = [operatPoint];
301 while ~isempty(searchQueue)
302     currentIndex = searchQueue(1);
303     if currentIndex*2+1 <= 2^maxLevel - 1
304         operateLocations(end+1) = currentIndex*2;
305         operateLocations(end+1) = currentIndex*2+1;
306         searchQueue(end+1) = currentIndex*2;
307         searchQueue(end+1) = currentIndex*2+1;
308     end
309     searchQueue(1) = [];
310 end
311 end
312
313 function [] = randomSearch(operator, varConst, maxLevel, data, n, repeat)
314 parfor r = 1:repeat
315     fileID = fopen(strcat('Random_',int2str(r),'.txt'),'wt');
316     bestError = inf;
317     %bestHeap = cell([2^maxLevel-1,1]);
318     for i = 1:n
319         % maximum level of heap can vary from 2 to 4
320         heap = heapGeneration(operator, varConst, maxLevel);
321         error = MAE_Cal(heap, data);
322         if error < bestError
323             bestError = error
324             %bestHeap = heap;
325         end
326         fprintf(fileID, '%d %10.8f %10.8f \n', i, error, bestError);
327     end
328     fprintf("RM: Eval: %d, Error: %10.8f.\n", n, bestError)
329 end
330 end
331
332 function eval = evalHeap(heap, data)
333 heapStr = heapString(heap);
334 fh = str2func(heapStr);
335 eval = fh(data(:,1));

```

```

336 end
337
338 function error = MAE_Cal(heap, data)
339 heapStr = heapString(heap);
340 fh = str2func(heapStr);
341 y = fh(data(:,1));
342 error = sum(abs(data(:,2) - y))/length(y);
343 end
344
345 function heap = heapGeneration(operator, varConst, maxLevel)
346 heapSize = 2^maxLevel - 1;
347 heap = cell([heapSize,1]);
348 opSize = 2^(maxLevel-2)-1;
349 operatorQueue = [1];
350 while ~isempty(operatorQueue)
351     % pick the current index
352     currentIndex = operatorQueue(1);
353     % assign current operator
354     heap(currentIndex) = operator(randi(length(operator)));
355     % make sure the operator assignment does not exceed the limit
356     if currentIndex <= opSize
357         if rand < 0.5
358             operatorQueue(end+1) = currentIndex*2;
359         end
360         if rand < 0.5
361             operatorQueue(end+1) = currentIndex*2+1;
362         end
363     end
364     % delete the first in queue
365     operatorQueue(1) = [];
366 end
367 % record the operator index
368 opIndex = find(~cellfun(@isempty, heap));
369 varConstIndex = [];
370 for i = 1:length(opIndex)
371     if isempty(heap{opIndex(i)*2})
372         varConstIndex(end+1) = opIndex(i)*2;
373     end
374     if isempty(heap{opIndex(i)*2+1})
375         varConstIndex(end+1) = opIndex(i)*2+1;
376     end
377 end
378 % assign x and c to the rest of the tree
379 for i = 1:length(varConstIndex)
380     heap(varConstIndex(i)) = varConst(randi(length(varConst)));
381 end
382 % replace the 'c' with actual constant
383 heap = replaceC(heap);
384 end
385
386 function heap = replaceC(heap)
387 heapSize = length(heap);
388 for i = 1:heapSize
389     const = -10:0.1:10;
390     if strcmp(heap{i}, 'c')
391         heap{i} = const(randi(length(const)));
392     end
393 end
394 end
395
396 function heapStr = heapString(heap)
397 heapStr = heap;
398 % reverse order from 15 to 1
399 for i = fliplr(1:floor(length(heap)/2))
400     % make sure current node is not empty
401     if ~isempty(heapStr{i})
402         % if current node is sin or cos, only combine its right child
403         if strcmp(heapStr{i}, 'sin') || strcmp(heapStr{i}, 'cos')

```



```

404         %heapStr{i} = strcat(num2str(heapStr{i*2}),'*',num2str(heapStr{i}),'(',num2str(
            heapStr{2*i+1}),')');
405         %heapStr{i} = strcat(num2str(heapStr{i}),'(',num2str(heapStr{2*i+1}),')');
406         heapStr{i} = strcat('(',num2str(heapStr{i}),'(',num2str(heapStr{2*i}),'.*',num2str(
            heapStr{2*i+1}),')','')');
407     else
408         % else combine the left child, current node and right child
409         heapStr{i} = strcat('(',num2str(heapStr{2*i}),num2str(heapStr{i}),num2str(heapStr{2*i
            +1}),')');
410     end
411 end
412 end
413 % take the top of the heap as output
414 heapStr = strcat('@(x) ',heapStr{1});
415 end

```