COMP SCI 5401 FS2017 Assignment 2c

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MOEA Explained

For this experiment I decided to go with a fairly simple approach for my CoEA. So I have a configurable variable called CoevolutionaryFitnessSamplePercent that can be set to any percent a user likes, this variable should be in the form of % (so 100 would be what it would be set at for 100%), it will get converted into a percentage during the execution of the program automatically. Once I had that variable all that was left was to get the number of opponents each bracket would have. In order to do that I simply make a variable that gets set to the formula $(\mu + \lambda - 1) * CoevolutionaryFitnessSamplePercent$ in order to obtain the number of opponents for that experiment. After getting that value all we have left to do have the program finish multiple evaluations for each bracket of the experiment, take the final fitness values and see which is the Best Composite Fitness out of the bracket as well as finding the Average Composite Fitness of the bracket. To do all of this we simply get the highest fitness value of the bracket and that becomes the Best Composite Fitness Value, we also take all of the fitness values, add them up, and divide by the number of opponents fighting in that round to obtain the Average Composite Fitness value. After that we are pretty much finished with the CoEA, it will compare multiple evaluations at a time in a very simple and straightforward way.

Experiment parameters and graphs

1 IPD Results Using Configuration 1

1.1 EA Experimental Setup for Configuration 1

In order to recreate this Experiment all that needs to be done is set the configuration file variables to the following values. In order to change each experiment I decided to use different percentages for the CoevoluationaryFitness-SamplePercent. The reason I did this was to change the number of opponents in each bracket which in turn creates entirely different calculations for the Average Composite Fitness. I also increased random variables to see if the results of the experiment differed by much or if they were fairly close together. The Best Composite Fitness seems to stay the same (Pretty much always sitting at 5), however the Average Composite Fitness does seem to change quite a lot depending on the parameters set up.

NOTE: If you want to get the same results you have to change the newSeed variable to 0 (Zero) in the configuration file in order to use the previous seed.

Using config1.txt

```
runs = 30
_{2} fitness = 10000
_{4} k = 5
5 d = 10
6 \ 1 = 30
7 n = 5
8 \text{ mu} = 0.01
9 \text{ lambda} = 5
10 parentNumber = 10
terminationEvals = 3
13 CoevolutionaryFitnessSamplePercent = 100
14
  prob_log_file = log_s/log_1.txt
  prob_solution_file = solutions/solution1.txt
17
18
19
   Initialize: Ramped_halfandhalf = 1
21
  parentSelection: Fitness_Proportional_Selection = 0, Over_Selection
       = 1
23
  Recombination: \ subTree\_Crossover\_Recombination = 1
24
25
  Mutation: subTree_Crossover_Mutation = 1
27
  survivalSelectionStrategy: Plus = 0, comma = 1
```

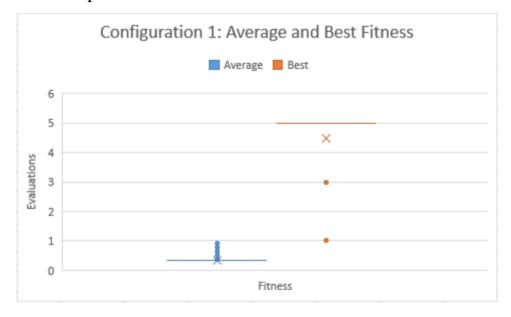
```
survivalSelection: Truncation = 1, kTournament = 0

bloatControl: parsimonyPressure = 1

Termination: numEvals = 1, noChange = 0

newSeed = 1
```

1.2 Graphs



2 IPD Results Using Configuration 2

2.1 EA Experimental Setup for Configuration 2

In order to recreate this Experiment all that needs to be done is set the configuration file variables to the following values. In order to change each experiment I decided to use different percentages for the CoevoluationaryFitness-SamplePercent. The reason I did this was to change the number of opponents in each bracket which in turn creates entirely different calculations for the Average Composite Fitness. I also increased random variables to see if the results of the experiment differed by much or if they were fairly close together. The Best Composite Fitness seems to stay the same (Pretty much always sitting at 5), however the Average Composite Fitness does seem to change quite a lot depending on the parameters set up.

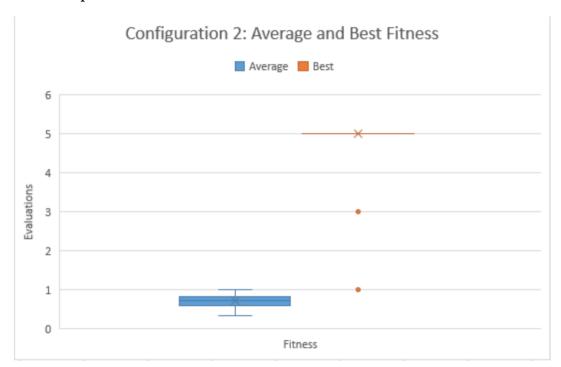
NOTE: If you want to get the same results you have to change the newSeed variable to 0 (Zero) in the configuration file in order to use the previous seed.

Using config2.txt

```
runs = 30
_2 fitness = 10000
_4 k = _5
5 d = 10
6 \ 1 = 30
7 n = 5
8 \text{ mu} = 0.05
9 \text{ lambda} = 5
10 parentNumber = 20
p = 1
terminationEvals = 3
  CoevolutionaryFitnessSamplePercent = 50
14
prob_log_file = logs/log2.txt
  prob_solution_file = solutions/solution2.txt
18
19
  Initialize: Ramped_halfandhalf = 1
20
21
  parentSelection: Fitness_Proportional_Selection = 1, Over_Selection
22
24
  Recombination: subTree_Crossover_Recombination = 1
25
  Mutation: subTree_Crossover_Mutation = 1
26
27
  survivalSelectionStrategy: Plus = 0, comma = 1
28
  survivalSelection: Truncation = 0, kTournament = 1
30
31
  bloatControl: parsimonyPressure = 1
32
Termination: numEvals = 0, noChange = 1
```

```
35
36 newSeed = 1
```

2.2 Graphs



3 IPD Results Using Configuration 3

3.1 EA Experimental Setup for Configuration 3

In order to recreate this Experiment all that needs to be done is set the configuration file variables to the following values. In order to change each experiment I decided to use different percentages for the CoevoluationaryFitness-SamplePercent. The reason I did this was to change the number of opponents in each bracket which in turn creates entirely different calculations for the Average Composite Fitness. I also increased random variables to see if the results of the experiment differed by much or if they were fairly close together. The Best Composite Fitness seems to stay the same (Pretty much always sitting at 5), however the Average Composite Fitness does seem to change quite a lot depending on the parameters set up.

NOTE: If you want to get the same results you have to change the newSeed variable to 0 (Zero) in the configuration file in order to use the previous seed.

Using config3.txt

```
runs = 30
  fitness = 10000
_4 k = _5
5 d = 10
6 \ 1 = 30
7 n = 5
8 \text{ mu} = 0.03
9 \text{ lambda} = 3
10 parentNumber = 15
p = 1
terminationEvals = 3
  CoevolutionaryFitnessSamplePercent = 25
14
prob_log_file = logs/log3.txt
  prob_solution_file = solutions/solution3.txt
18
19
  Initialize: Ramped_halfandhalf = 1
20
21
  parentSelection: Fitness_Proportional_Selection = 1, Over_Selection
22
24
  Recombination: subTree_Crossover_Recombination = 1
25
  Mutation: subTree_Crossover_Mutation = 1
26
27
  survivalSelectionStrategy: Plus = 1, comma = 0
28
  survivalSelection: Truncation = 0, kTournament = 1
30
31
  bloatControl: parsimonyPressure = 1
32
Termination: numEvals = 0, noChange = 1
```

```
35
36 newSeed = 1
```

3.2 Graphs



4 Results

4.1 Result Tables

final results

	А	В	С	D	E	F		
1	Averages and Standard Deviations							
2	Configuration 1		Configuration 2		Configuration 3			
3	Average Fitness	0.346364	Average Fitness	0.706485	Average Fitness	1.538589		
4	Best Fitness	4.491465	Best Fitness	4.9948	Best Fitness	4.75808		
5	Absolute Fitness	5	Absolute Fitness	5	Absolute Fitness	5		
6	Average S Deviation	0.036478	Average S Deviation	0.157204	Average S Deviation	0.457684		
7	Best S Deviation	1.332133	Best S Deviation	0.129098	Best S Deviation	0.855997		

Data Summary

,								
	Samples							
	1	2	3	4	5	Total		
N	300000	300000	300000			900000		
ΣΧ	1347198	1498150	1427162			4272510		
Mean	4.4907	4.9938	4.7572			4.7472		
ΣX^2	6583190	7487958	7010358			21081506		
Variance	1.7779	0.0215	0.7368			0.8877		
Std.Dev.	1.3334	0.1466	0.8584	NaN	NaN	0.9422		
Std.Err.	0.0024	0.0003	0.0016	NaN	NaN	0.001		

standard weighted-means analysis								
ANOVA Summary Correlated Samples k=3								
Source	SS	df	MS	F	Р			
Treatment [between groups]	38022.2708	2	19011.1354	22511.71	<.0001			
Error	506708.3959	599998	0.8445					
Ss/Bl	254173.4443	299999			Graph Maker			
Total	798904.111	899999						

Tukey HSD Test

```
HSD[.05]=0.01; HSD[.01]=0.01
M1 vs M2 P<.01
M1 vs M3 P<.01
M2 vs M3 P<.01
```

M1 = mean of Sample 1 M2 = mean of Sample 2 and so forth.

HSD = the absolute [unsigned] difference between any two sample means required for significance at the designated level. HSD[.05] for the .05 level; HSD[.01] for the .01 level.

5 Discussion Section

So for the most part I expected the results I obtained in the end of this experiment. I figured that the Best Composite Fitness would end up being 5, and I knew the Absolute Fitness would be 5. I will say that I did not expect the Average Composite Fitness to fluctuate so much between the three experiments, I also kind of expected the Average Composite Fitness to be closer to 3.

For the statistical Analysis I used an online calculator for an anova test. According to the final results I ended up with all the P values being under 0.01, this means that all of the results were statistically significant since anything under 0.05 is statistically significant. The test ran each Fitness Group against the other two fitness groups, obtaining the p-value for each test.

After finishing the experiments I believe that I can say giving the experiments a larger parent population help improve the Average Composite Fitness because there is a larger chance of getting higher fitness offspring the more children there are to choose from, meaning that it is easier to get out of a local Maximum because there are more recombination and mutation options to choose from.

6 Conclusion

After all is said and done I would say this was a successful experiment, while I am unsure whether the values I got were appropriate or not I cannot say fully, however I do know that obtaining a final value for the Best Composite Fitness most of the time should be correct, and I know the Absolute Value of 5 is correct. I was a little surprised when it came to the Average Composite fitness however, I expected the average to be closer to 3 for all of the exeriments, I also expected the averages between the 3 experiments to be closer together. It turns out that the changes I made to the configurations did impact the results quite a bit. Those are the main findings that surprised me from this experiment overall.