

Parent Selection and Diversification in Genetic Programming

Thomas Helmuth
Washington and Lee Univ
Someplace nice
A city, Virginia
helmuth@wlu.edu

Nicholas Freitag McPhee
Div of Sci and Math
Univ of MN, Morris
Morris, MN 56267
mcphee@morris.umn.edu

Lee Spector
Cognitive Science
Hampshire College
Amherst, MA
lspector@hampshire.edu

ABSTRACT

More things!

Keywords

lexicase selection, hyperselection, PushGP, other stuff

1. INTRODUCTION

I bet we start here!

Lexicase selection [3] is nifty, eh?

2. EXPERIMENTAL DESIGN

Previous work has shown that using lexicase selection results in higher population error diversity than tournament selection across a variety of problems [1, 2]. These examples examined the diversity of entire GP runs, each starting with a different initial population and random number seed.

Here we examine the effects of these parent selection methods on population diversity starting from specific population conditions besides a random initial population. In particular, we want to see how each method changes diversity in populations that occur naturally during an evolutionary run.

In order to produce the populations on which to experiment, we started GP runs and let them continue until they met certain conditions; we then stored those populations and later conducted multiple trials with different random number seeds starting with those stored populations. We used three different stopping conditions in order to generate naturally occurring populations with interesting properties:

1. In a run using lexicase selection, we stopped if the population error diversity was greater than 0.9. This results in very diverse populations, allowing us to observe whether evolution is able to maintain such high diversity in the following generations.
2. In a run using tournament selection, we stopped if the population error diversity was less than 0.15. These

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populations allow us to see if methods promote diversification starting from such undiverse populations. They also allow us to see if methods perform differently on a population produced by tournament selection versus one produced by lexicase selection.

3. As described above, we were initially motivated here by observations of runs using lexicase selection undergoing major drops in diversity following hyperselection events, where one or a few individuals in the population received the majority of the parent selections in a generation. We had anecdotally noticed rapid diversity recovery following these events, but not examined them systematically.

In this condition, we stopped a run using lexicase selection when the error diversity reached a level at least 0.25 less than it had been at some point in the previous 10 generations. This allowed us to detect populations that had recently undergone large drops in diversity. We do not know for sure whether those drops are related to hyperselection events, but we expect that they are.

In all three conditions, we only considered populations occurring after generation 10 in order to give evolution a chance to settle down after the extreme shifts that can happen at the beginning of a run.

- We ran each for 20 generations more - 2 problems

3. CONCLUSIONS

I'm hoping we have conclusions.

Acknowledgments

Lots of cool people helped us.

4. REFERENCES

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