

Visualizing Genetic Programming Ancestries

McPhee¹, Casale², Finzel, Helmuth, & Spector

¹Division of Science and Mathematics
University of Minnesota, Morris
Morris, Minnesota, USA

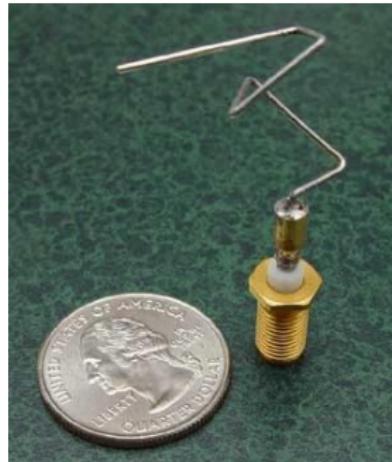
²Design Center Inc.
St. Paul, Minnesota, USA

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VizGEC

Why?

Evolutionary Computation allows us to:

- Create algorithmic solutions by simulating biological evolution.
- Derive solutions that humans wouldn't necessarily find.
 - Like NASA's antenna! →

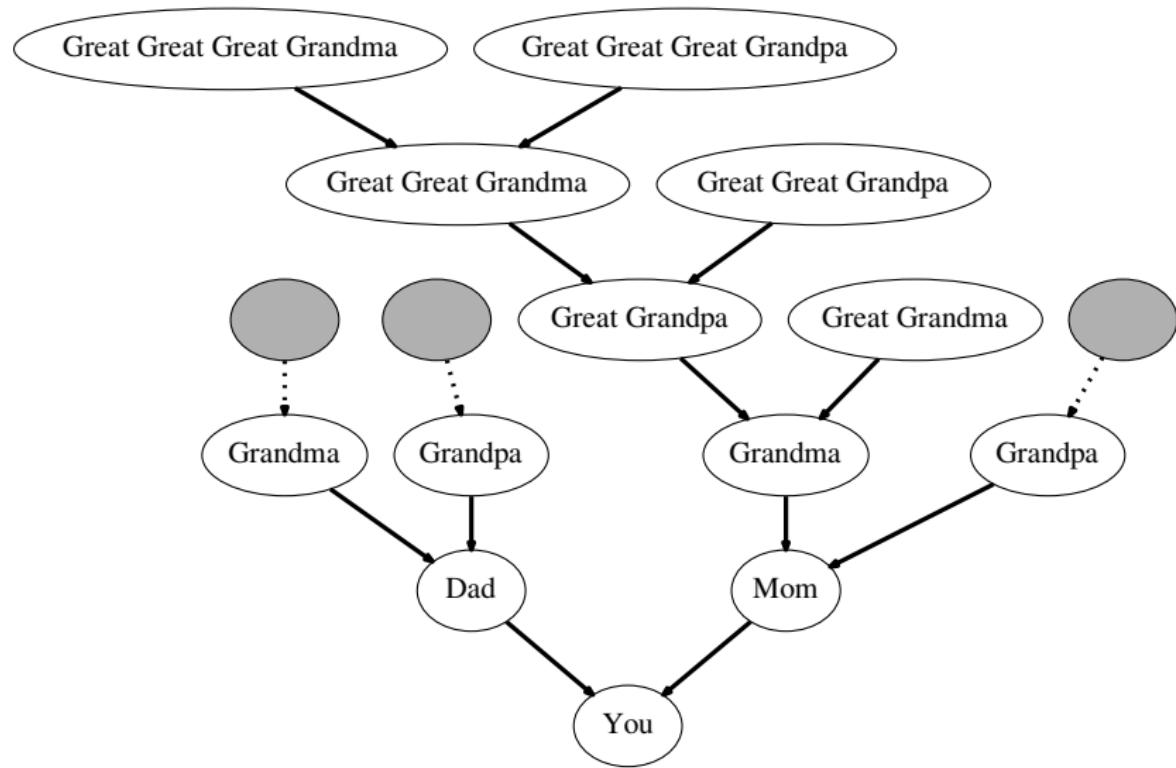


Our Research has allowed us to:

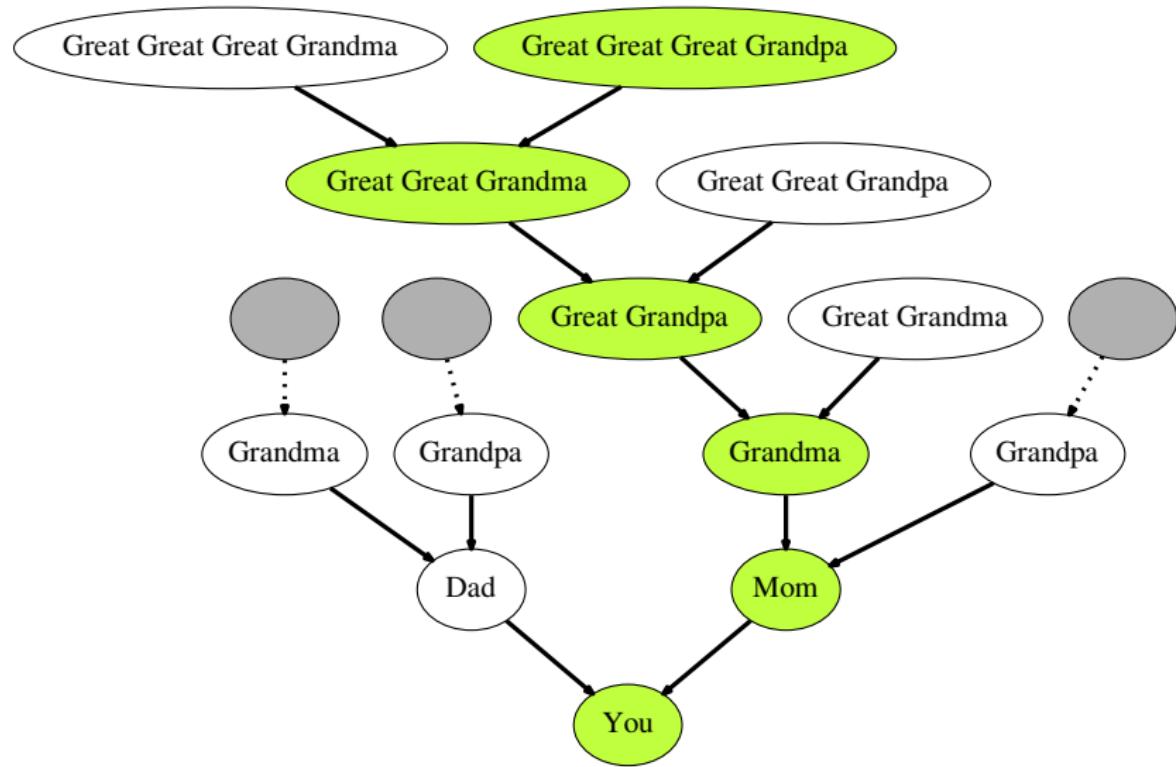
- Turn our complex runs into organized ancestry trees.
- Modify aspects of both nodes and edges to visualize analytics.
- Find patterns in data that couldn't be seen in previous work.

[https://en.wikipedia.org/
wiki/Evolved_antenna](https://en.wikipedia.org/wiki/Evolved_antenna)

Family Tree



Highlighted Individuals



Outline

- 1 Background
- 2 Graph Structure
- 3 What did I learn?
- 4 Conclusions

Outline

- 1 **Background**
 - Evolutionary Computation
 - Runs
 - Ancestry Trees
- 2 Graph Structure
- 3 What did I learn?
- 4 Conclusions

Evolutionary Computation Versus Biological Evolution

Evolutionary Computation

- Programs
- Population growth occurs until solution

```
31 (define create-snake []
32   (body (list [3 0] [2 0] [1 0] [0 0])
33     direction [1 0]
34     type :snake
35     color (Color. 15 160 70)))
36 )
37
38 (define create-apple []
39   (location ([rand-int field-width) (rand-int field-height)
40     color (Color. 210 50 90)
41     type :apple))
42 )
43
44 (define point-to-screen-rect [[pt-x pt-y]]
45   [(+ pt-x point-size) (* pt-y point-size) point-size point-size])
46 )
47
48 (define move [(ways {body direction} :as snake) & grow]
49   (unless snake body
50     (cons
51       (let [[head-x head-y] (#first body)
52             [dir-x dir-y] direction]
53         {[+ head-x dir-x] [- head-y dir-y]})
54       (if grow body (#rest body))))
55     )
56 )
57
58 (define turn [snake direction]
59   (assoc snake direction direction))
```

<https://i.ytimg.com/vi/-XzSGPJRBSw/>

maxresdefault.jpg

Biological Evolution

- Organisms
- Continuous growth



<https://www.flickr.com/photos/keesey/>

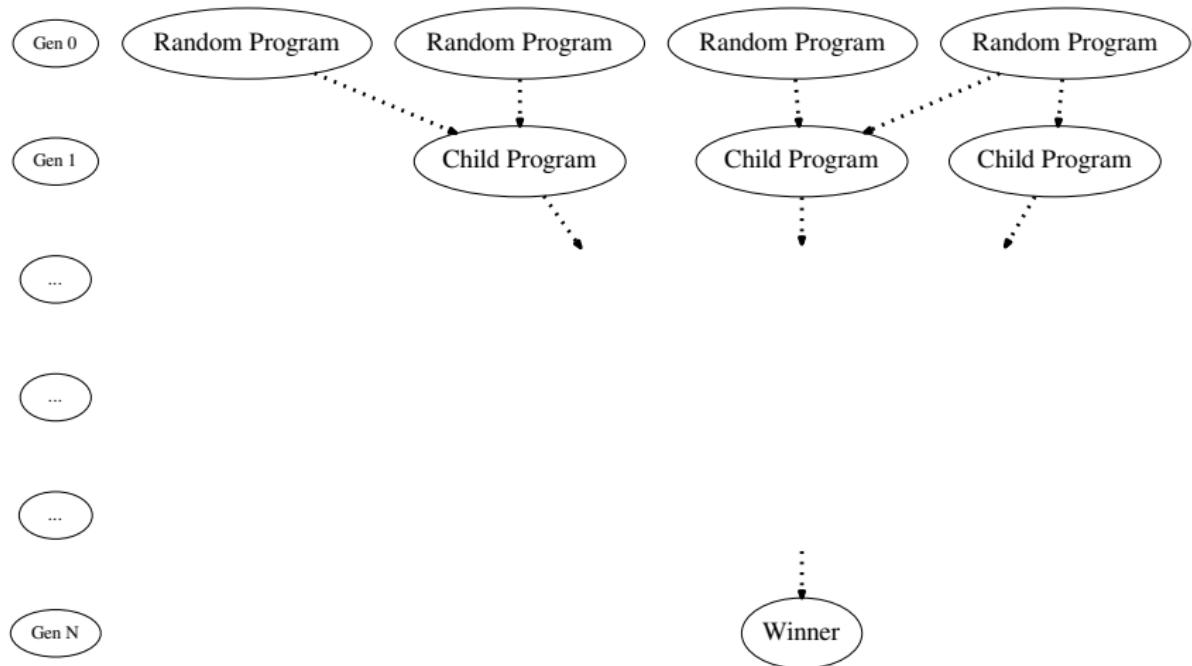
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Run Structure

In a run, we have many generations of individuals or programs.

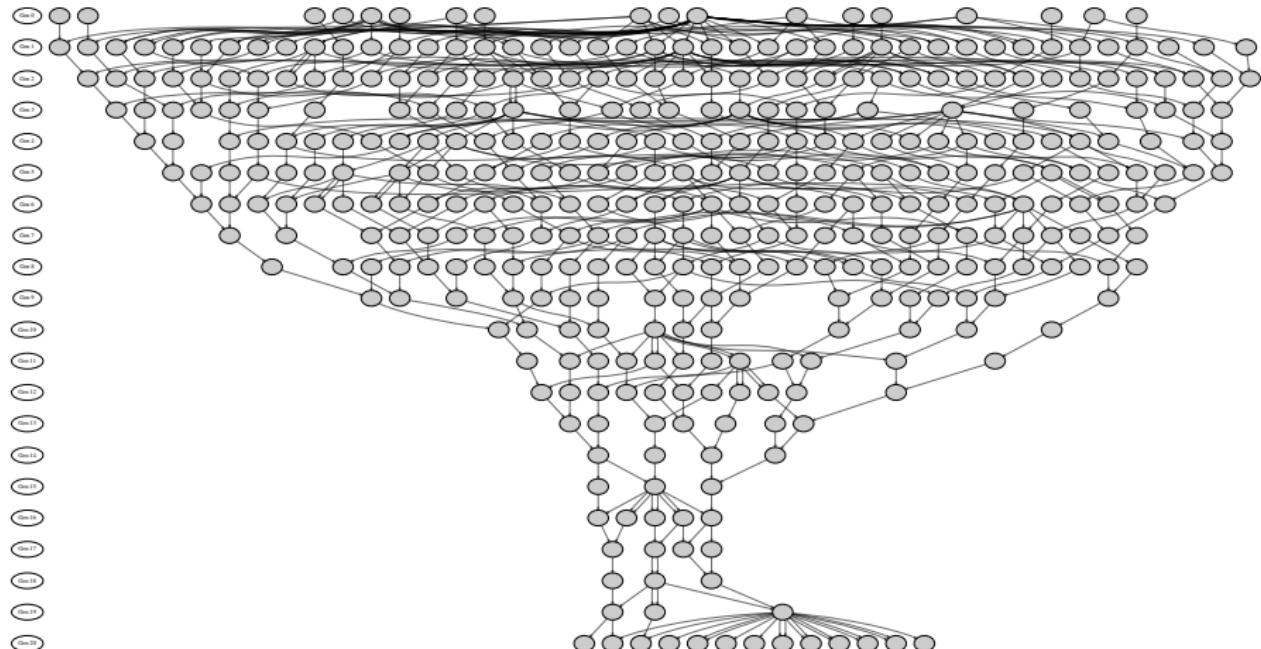
- Start with 1,000 individuals
- Test each individual on 200 test cases
- Assign error based on test results
- If a solution is found, stop
- Otherwise, make a new generation

Ancestry Trees: Basic Structure



Basic Run 0

Random Programs ↓



Winning Program(s) ↑

Outline

1 Background

2 Graph Structure

- Nodes
- Edges

3 What did I learn?

4 Conclusions

Node Basics

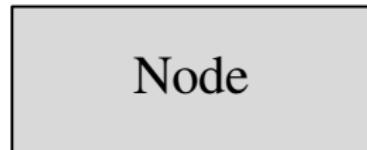
Nodes represent an individual or program.

They know:

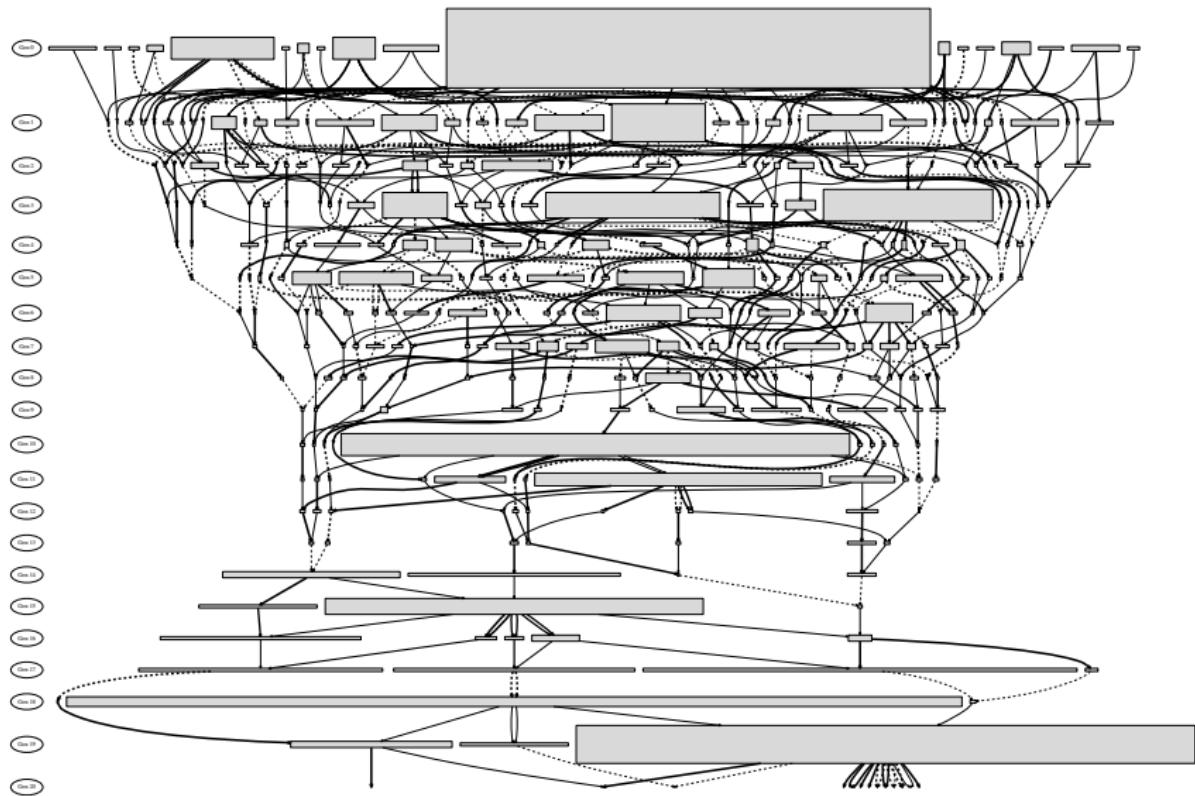
- Number of selections: Width
- Number of children: Height

← Number of Selections →

↑ Number of Children ↓

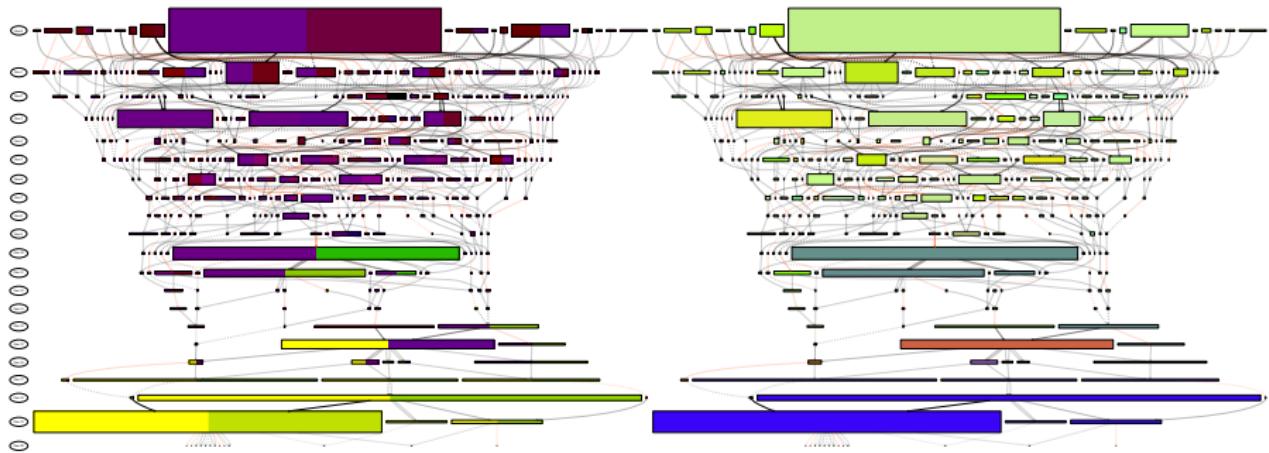


Run 0: No color



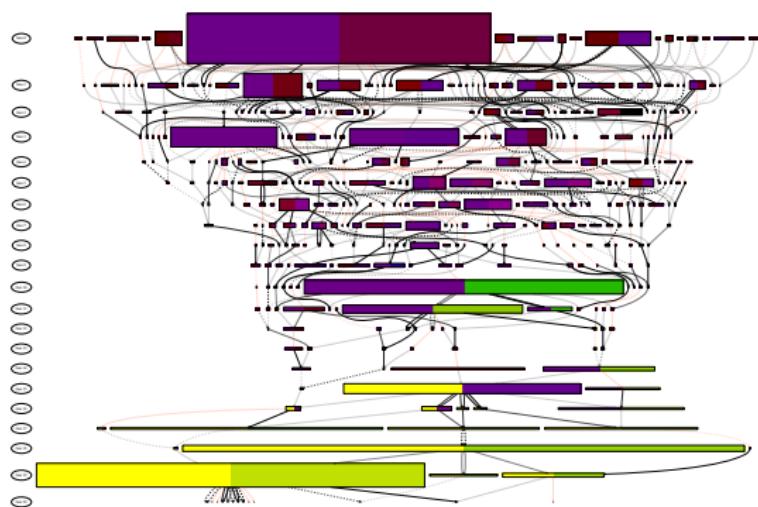
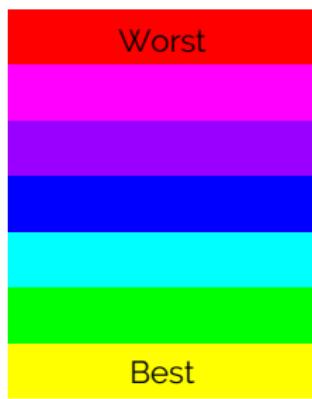
Node Coloring

There are two different techniques we use:
Error Based & RBM



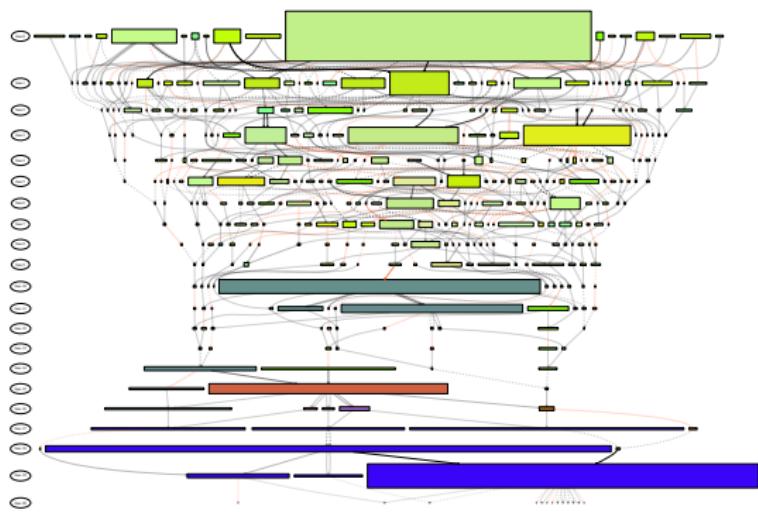
Error Based Node Coloring

- Two part problem
- Based on an individual's total error.
- Additional shading for very high errors.



RBM Node Coloring

- RBM = Restricted Boltzmann Machines
- Dimensionality Reduction Problem
- 200 tests → 3 values



Edges

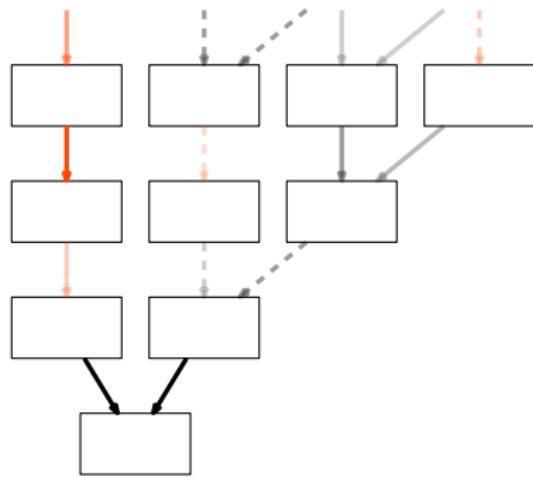
Edges represent how an individual was created.

There are two main techniques: crossover and mutation.

- Crossover
Sections of both parents are combined together.
- Mutation
One parent is copied and some instructions are mutated.

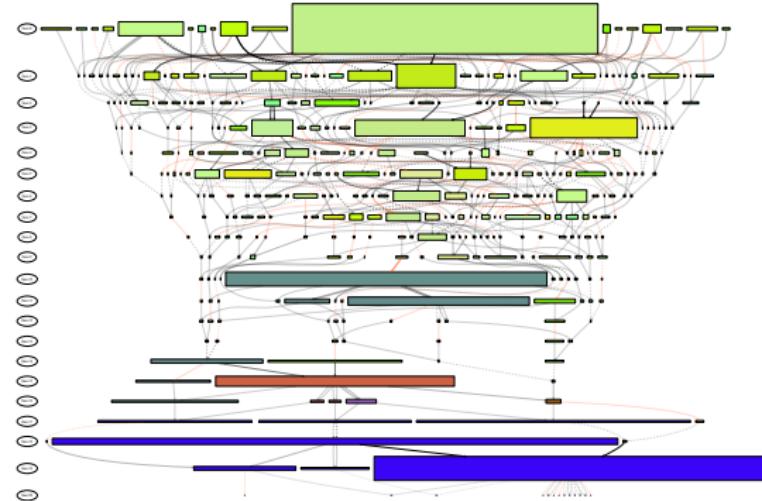
Edge Coloring & Style

- Crossover is indicated by black edges.
- Mutation is indicated by orange edges.
- Solid versus dashed is for variations of crossover and mutation.
- Transparency is based on the number of children of the child.
 - Transparent: Few Children
 - Opaque: Many Children



Full Examples in RBM Coloring

- Run 0: Ends in 20 Generations
- Run 1: Ends in 129 Generations
 - Notice waist with large nodes



Outline

1 Background

2 Graph Structure

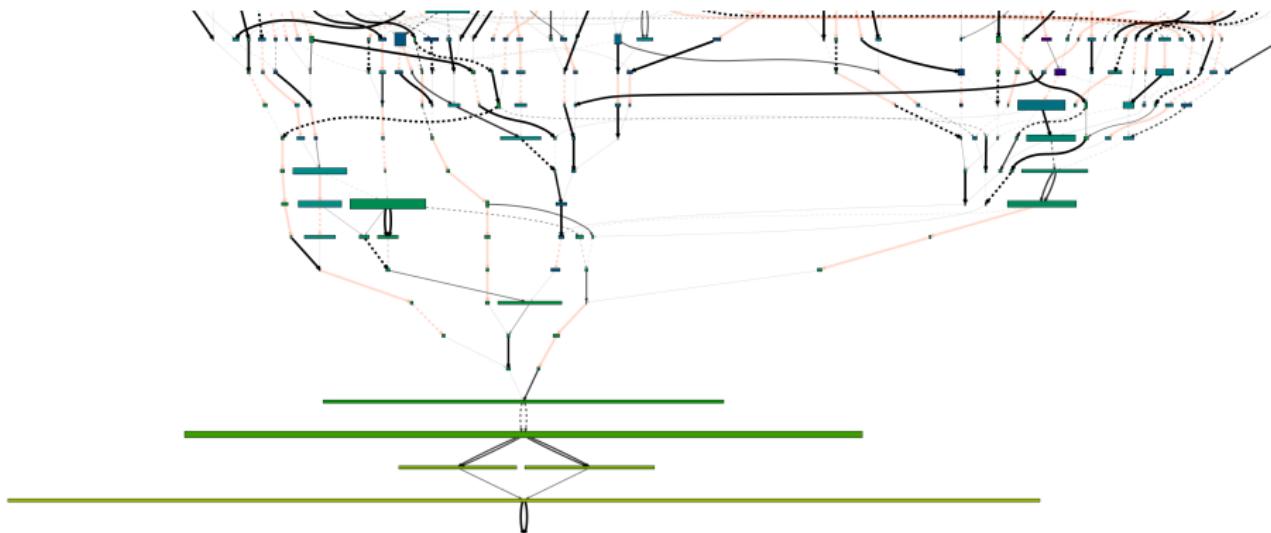
3 What did I learn?

- Hyperselection
- Filtering Ancestry Trees

4 Conclusions

Hyperselection

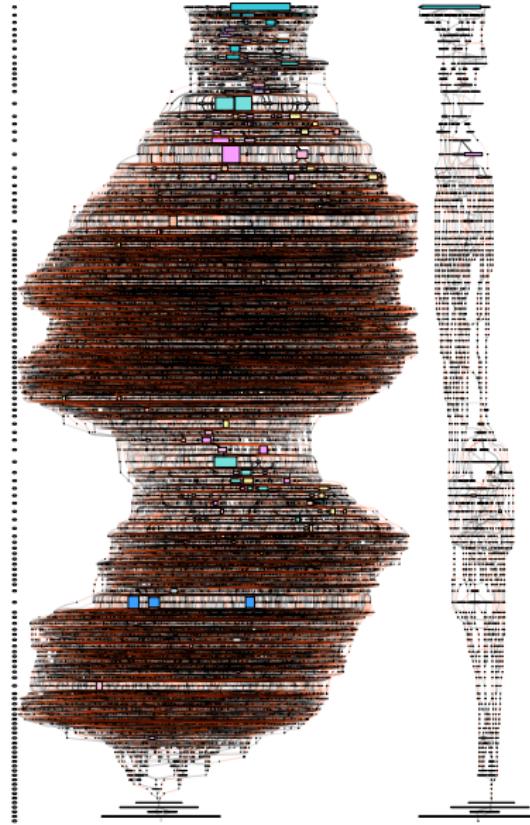
One focus of this research was to examine *hyperselection*. This occurs when an individual receives a noticeably high number selections in a population.



In our graphs, we can see this as a very wide node.

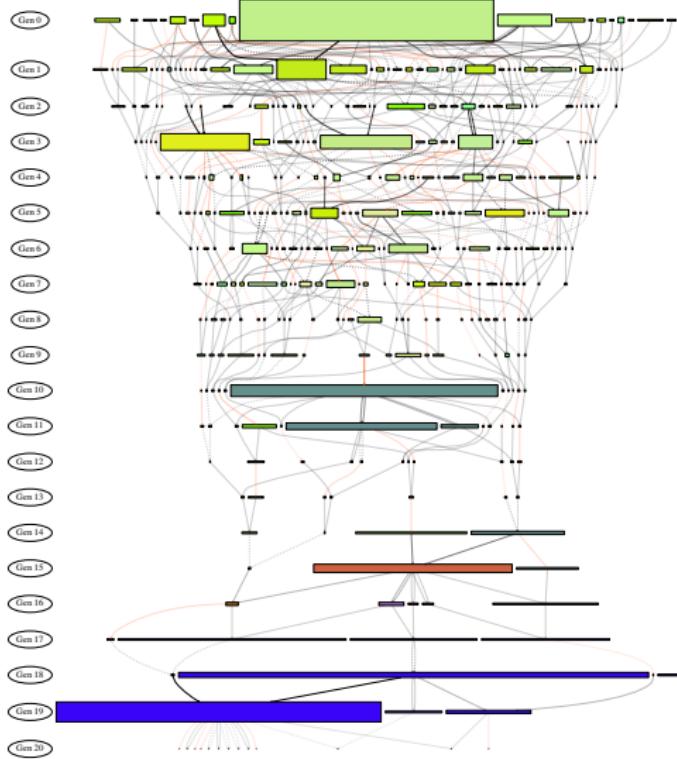
Filtering Ancestry Trees

- This is still a lot of data!
 - 22,435 Individuals
 - 1,597 Individuals
- We filter out parents who did not contribute much genetic information.
- Numbers on edges indicate similarity.
- The smaller the number is, the more similar they are.

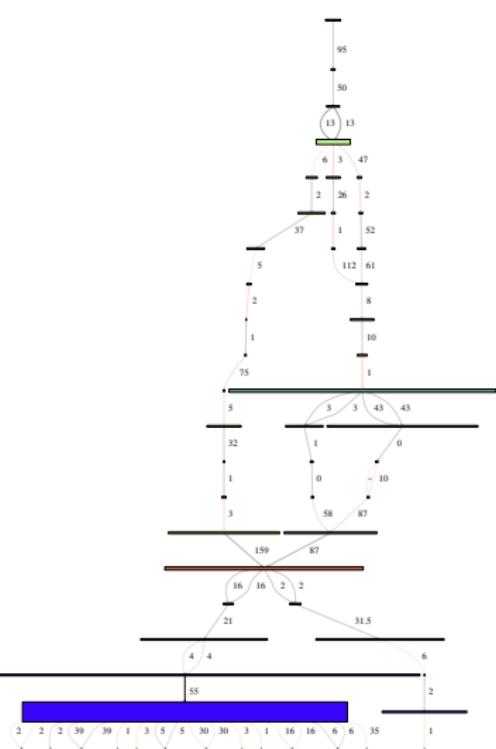


Another Filtered Example

394 Individuals



54 Individuals



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Conclusions

In Our Research We've:

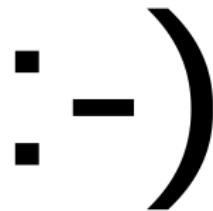
- Turned our complex runs into organized ancestry trees.
- Modified aspects of both nodes and edges to visualize analytics.
- Learned about hyperselection and filtering ancestry trees.

The Next Step:

- Presenting related work in May at GPTP in Ann Arbor, Michigan
- In July we'll be attending GECCO in Denver, Colorado

Thanks!

Thank you for your time & attention!



Special Thanks to Nic McPhee!

References

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