

AUGMENTED POPULATION BASED TRAINING (APBT)

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Problem

Searching and Tuning Neural Nets is a difficult and long process!

- Artificial Neural Nets (ANN) are very powerful function approximators
- To use effectively, ANNs require lots of hyperparameters to tune (6+, too much to exhaustively search!)
- Moreover, finding the right topology is even more important and difficult. Wrong topology invalidates best hyperparameters...



Approach

We propose Augmented Population Based Training!

- Leverage power of Genetic Algorithms to help search both topology and hyperparameters in SINGLE joint training process.
- Produce optimal topology (high accuracy, low size) trained on optimal hyperparameter schedules (adaptive, not just one value)
- Requires fewer hyperparameters (3 at most, 2 usually!)
- Can be parallelize for speed boost.
- Better than manual, grid, and random search

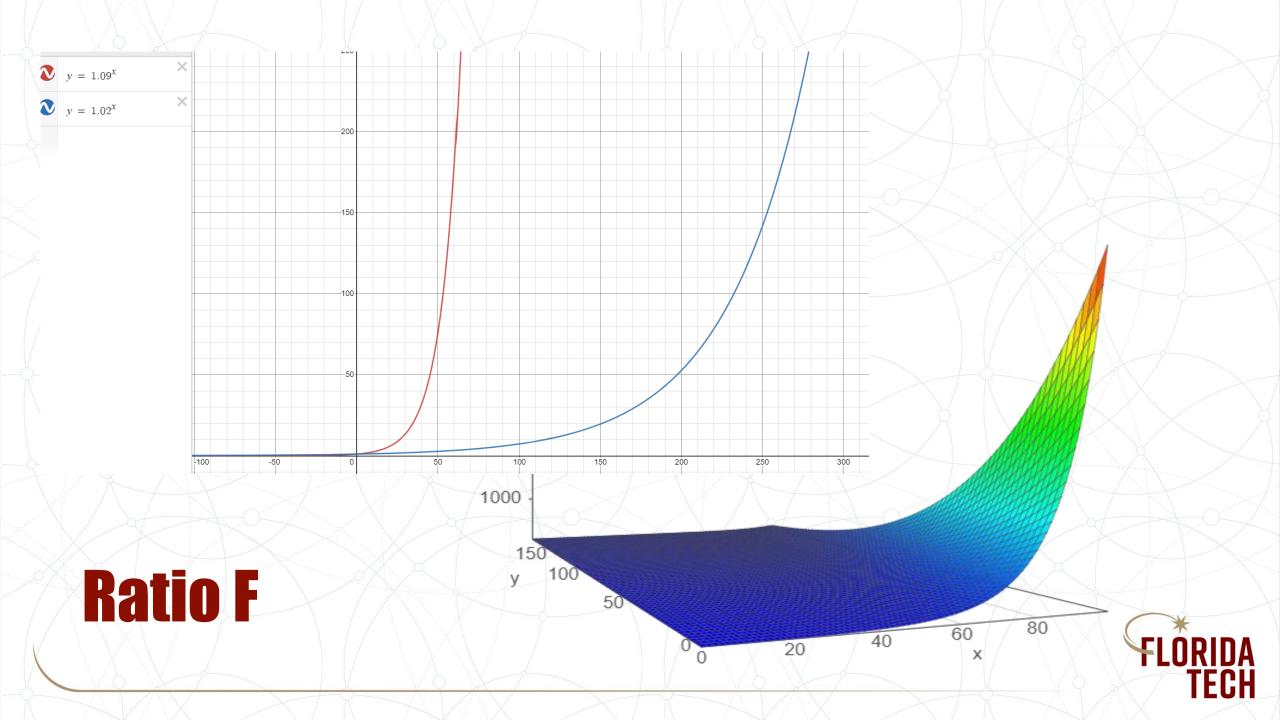


Evaluate

- · It's our Fitness function here.
- Takes a neural net, a validation/test set, and returns a score
- Evaluate the accuracy (1), get net size (2), and apply a special ratio function (f) on the accuracy and size to get performance score.

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$$f(accuracy, size) = \frac{1.09^{accuracy}}{1.02^{size}}$$





Truncation Selection

- It's our selection function based on the readiness threshold.
- A neural net is simply "ready" after a set number of iteration has elapsed since last time it was ready
- Truncation selection works by first ranking the nets in the population by performance ratio. If current net is in the bottom 20% of the population, we sample another net uniformly from the top 20% of the population to exploit.

Exploit

- · It's our crossover function.
- When a net is ready after a certain number of iterations, and is struggling in performance (bottom 20%), it can exploit the rest of the population by copying the topology, weights, and hyperparameters of the top net selected from the population with truncation selection.

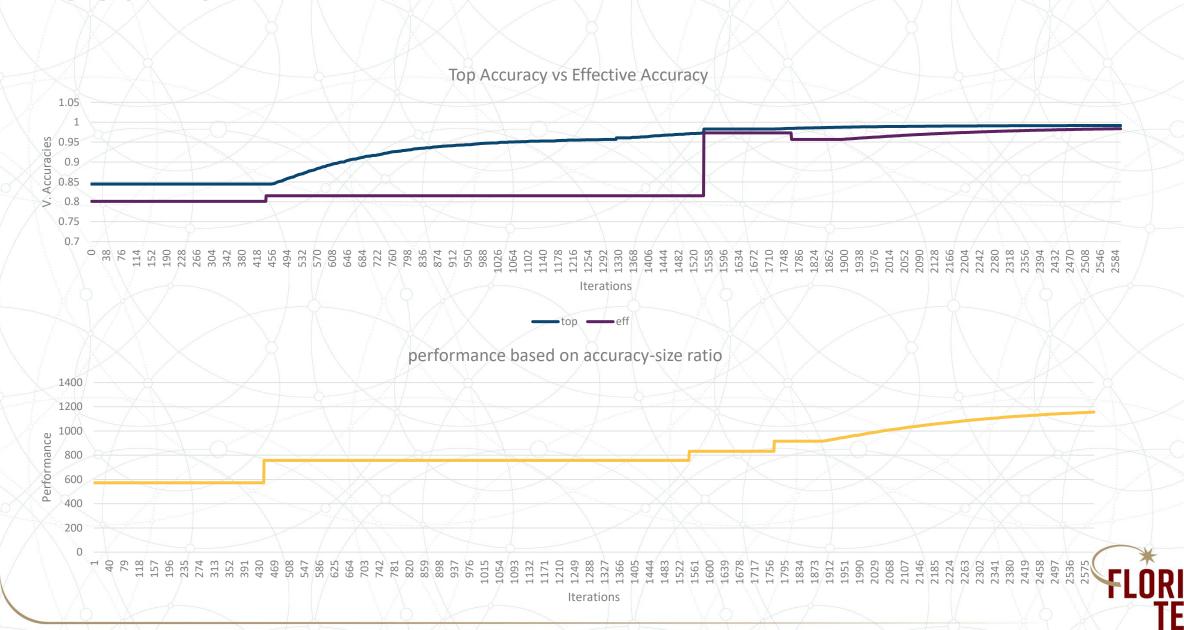


Explore

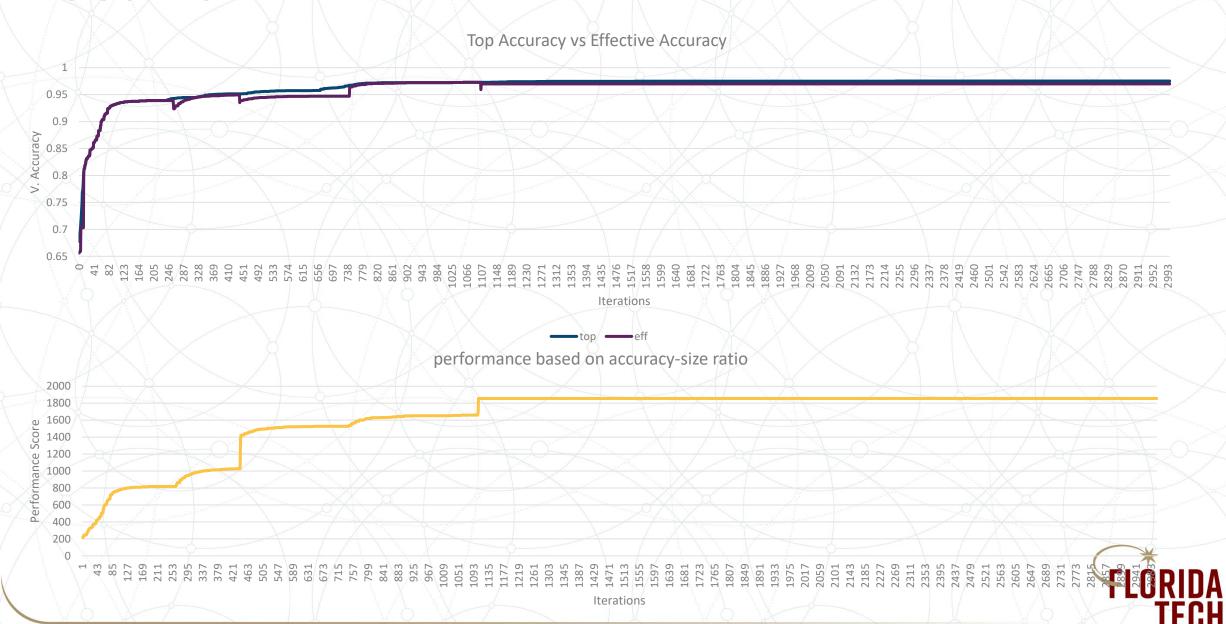
- · It's our mutation function.
- Following exploitation, the net experiences a perturbation in its hyperparameters, and topology before continuing training with the new hypermeters and topology.
- Hyperparameters are multiplied by either .8 or 1.2
- Topology either gains a neuron, loses a neuron, or does nothing.



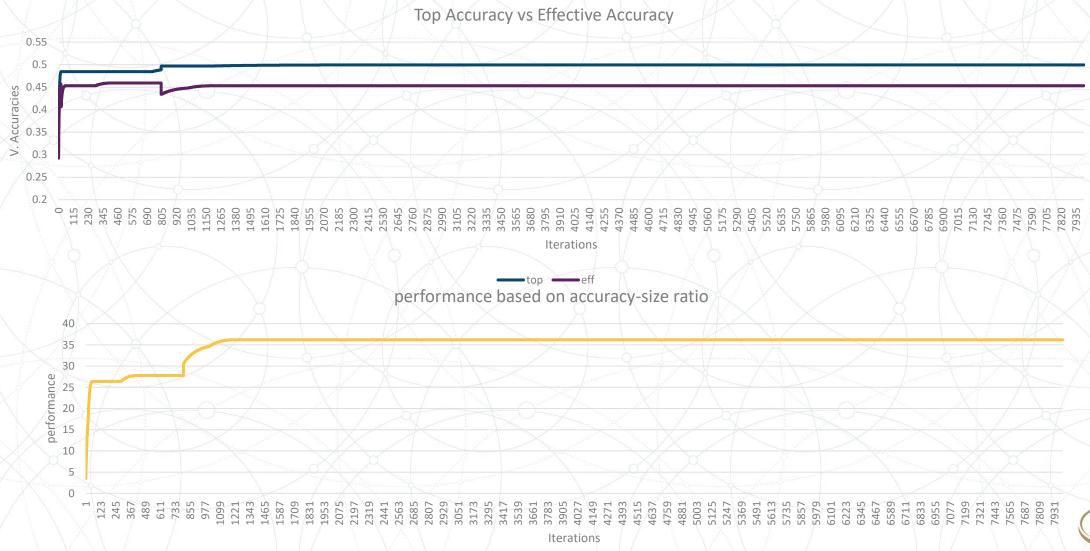
Results Tennis Dataset



Results Iris Dataset



Results Identity Dataset





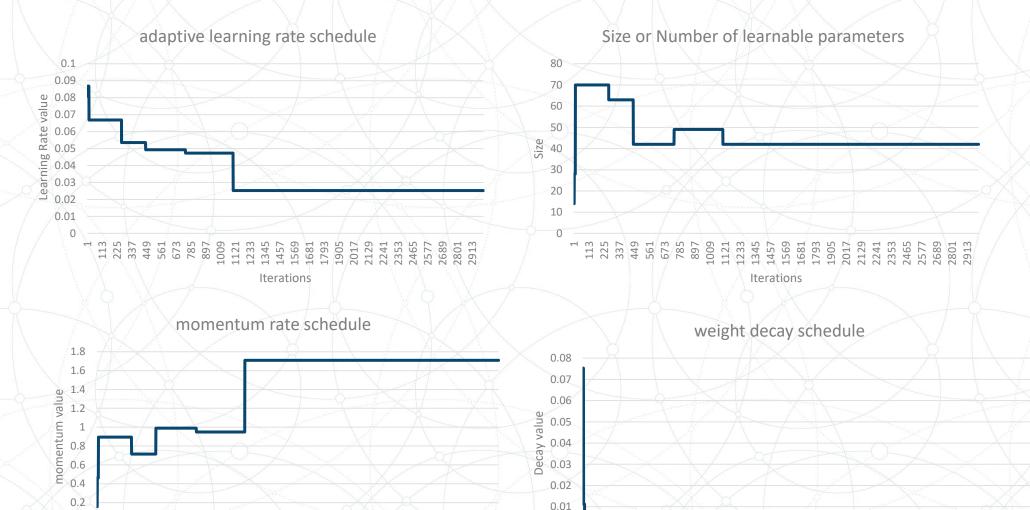
Results

	Iris							
	Trial 0		/	Trial 1			Trial 2	
	Best	Most acc		Best	Most acc		Best	Most acc
Test Acc %	94.24	96.6		91.24	94.88		66.42	68.8
Num Params	42	63		28	49		7	70
Topology	4,6,3	4,9,3		4,4,3	4,7,3		4,1,3	4,10,3
	John John Stranger							
			N. X			X I	9	
\ \X\ \	Tennis							
	Trial 0			Trial 1			Trial 2	
	Best	Most acc		Best	Most acc		Best	Most acc
Test Acc %	96.72	97.14		85.33	90.59		92.49	95.77
Num Params	72	120		12	96		12	24
Topology	10,6,2	10,10,2		10,1,2	10,8,2		10,1,2	10,2,2
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			K X		The state of the s	X		Y-Y-V
1	Identity	(struggles here)	20/	10-1			10/	
	Trial 0			Trial 1			Trial 2	
	Best	Most acc		Best	Most acc		Best	Most acc
Test Acc %	55.2	93.38		54.81	92.39		55.32	89.12
Num Params	16	144		16	160		16	144
Topology	8,1,8	8,9,8		10,1,2	8,10,8		8,1,8	8,9,8



Hyperparameter schedules (think of what they mean)

Iterations



Iterations



Discussion: Advantages

- Works quite well in optimizing topology and hyperparameters!
- Provides both absolute highest accuracy net and highest accuracy with lowest size net.
- Trains ANN adaptively on hyperparameter schedules as opposed to values for efficient training
- Requires much less tunable parameters.
- Found 10.2.2 at 96% on Tennis
- Not sensitive to local minima or overfitting
- Can be parallelized



Discussion: Drawbacks

- APBT uses k (population size) times more space to run than BP, runs k times longer. But it can be parallelized to run much faster!
- APBT is inconsistent with too small k and to small datasets
- APBT struggles to find minimum size net with Identity dataset
- Doesn't support early stopping yet



Some Observations

- Higher k the better, as long as resources allow it.
 Increases diversity in population
- Higher Selection Readiness allows nets to train for longer before sharing. But too high readiness and it's just k-random search.
- Most effectively accurate net is not always the most absolutely accurate
- Longer epochs range allow for rest of population to catch up to the top performers.



Summary

- APBT can work really well and requires far less hyper parameters, providing the most accurate, lowest size neural net, which is trained on the optimal hyperparameter schedule.
- However, APBT takes k (population size) times more space to run than BP, and without parallelism runs k times longer than BP. Can be inconsistent with smaller k and very small datasets. Struggles with Identity
- In all, to solve the problem of ANN topology and hyperparameters optimization, Augmented Population Based Training is a powerful step in the right direction.



Thank you.

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