

The Lottery Ticket Hypothesis

paper by Jonathan Frankle and Michael Carbin @ MIT CSAIL [link]

presentation by **Jack Morris** 12/1/19

Background: Pruning

- To reduce the size of a neural network by removing unwanted parts
- People have been trying to **prune** neural networks for awhile
 - Idea originated into 1990s

The Pruning Process

- 1. Train the network
- 2. Remove superfluous structure
- 3. Fine-tune the network
- 4. [optional] iteratively repeat steps 2 and 3

What structure?

Weights? Neurons? Filters? Channels? Layers?

What does "superfluous" mean?

Magnitudes? Gradients? Activations?

Motivation

As you may imagine, lots of groups have tried something like this before

"Training a pruned model from scratch performs worse than retraining a pruned model, ..., which may be due to the difficulty of training small networks from scratch" – Pruning Filters for Efficient ConvNets

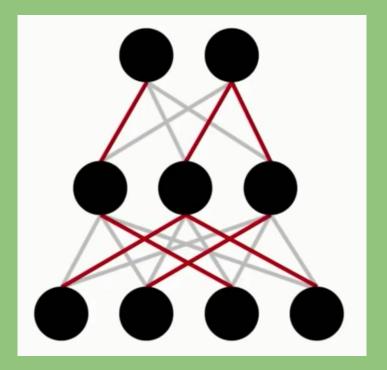
Motivating Questions

Can we train sparsely pruned networks from scratch? Yes

Corollary: Do networks have to be overparameterized to learn? No

There's a catch: **Need to reuse the weight initializations from the original training process**.

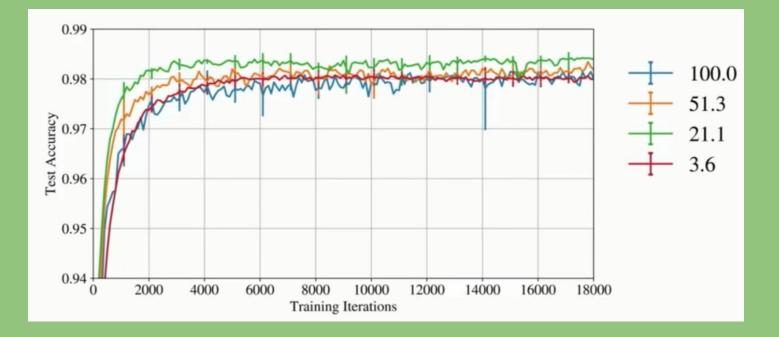
- 1. Randomly initialize the network's weights
- 2. Train it and prune superfluous structure
- 3. Reset each remaining weight to its value from 1
- 4. Repeat and 💰 **PROFIT** 💰



- Giant appendix shows that this works with batch normalization, dropout, convolutional layers, weight decay, residual connections, optimizers, etc. for any hyperparameter choices
- Caveats:
 - If you randomly reinitialize the network, this won't work
 - 2. You still have to train the network first (so it's not a particularly efficient process)

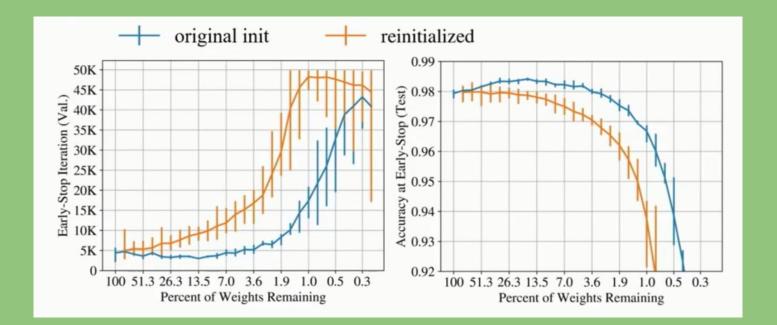
- So to recap, these small subnetworks:
 - 1. Are between 15% and 1% of the original size
 - 2. Learn faster than the original network
 - 3. Reach the same or higher test accuracy

Results



LeNet 300-100-10 for MNIST / fully-connected / 300k parameters

Results



LeNet 300-100-10 for MNIST / fully-connected / 300k parameters

The Lottery Ticket Hypothesis

Plain English:

Dense, trainable networks have sparse trainable subnetworks that are equally capable

Formally:

f(x; W) reaches accuracy **a** in **t** iterations f(x; m°W) reaches accuracy **a'** in **t'** iterations

$\exists m \mid \sum m \ll w$ $a' \ge a$ $t' \le t$

Possible future work

- Finding a way to prune networks early in training
- Examining these subnetworks to see what works– use this info to develop better architectures and initializations
- Make good subnetworks and reuse them on tasks
 - (A good test for overfitting, too)
- Stabilizing the Lottery Ticket Hypothesis (Frankle, 2019)
 - Prune after a few iterations, not at t=0
 - Compress ResNet-50, Inception-v3 in one shot by over 50%!

Questions?