

Learning Compact, General Purpose Neural Network Architectures

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Neural Architecture Search - Why?

Deep Learning:

Data + **Architectures** + Non-architecture hyperparameters \longrightarrow Weights.

NAS:

Data + Non-architecture hyperparameters \longrightarrow **Architectures** + Weights.

Why?

- Current Methods - **Primitive, limited by our biases.**
- Focusing - Optimizing weights θ , correct function form f ?
- Architectures - Extremely **complicated**.

Neural Architecture Search - Current Methods?



Issues:

1. **Computationally intensive & millions of parameters** (Zoph and Le [2016], Zoph et al. [2018], Real et al. [2017, 2019]).
2. Still require **domain engineering** (Zoph and Le [2016], Zoph et al. [2018]).
3. **Restrictive search space** - convolutional layers (Zoph et al. [2018], Real et al. [2019], Baker et al. [2016], Suganuma et al. [2017]).

Image: Adapted from Elsken et al. Neural architecture search: A survey.

Our Work - Compact, General Purpose

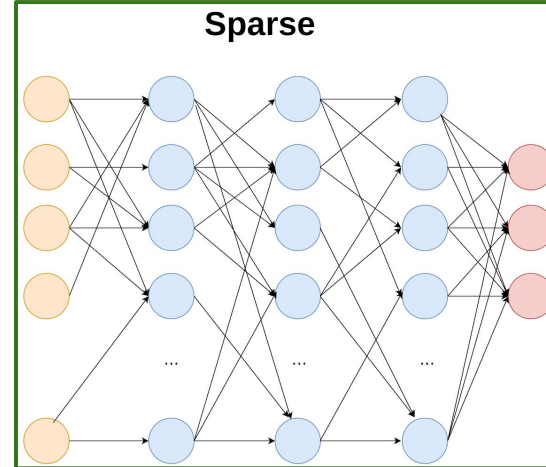
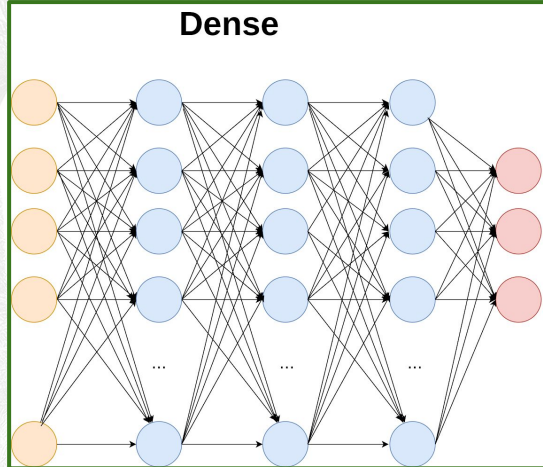
Goal:

Compact
General Purpose



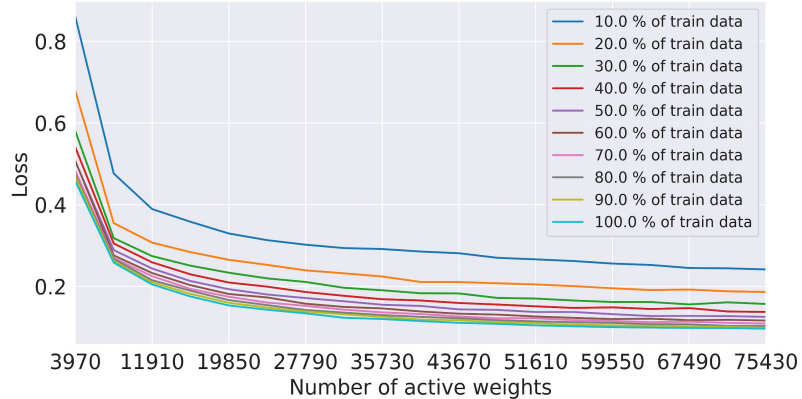
Smaller Models, fewer weights.
Only Neurons, Weights, Hidden layers.

Dense vs Sparse Neural Networks:

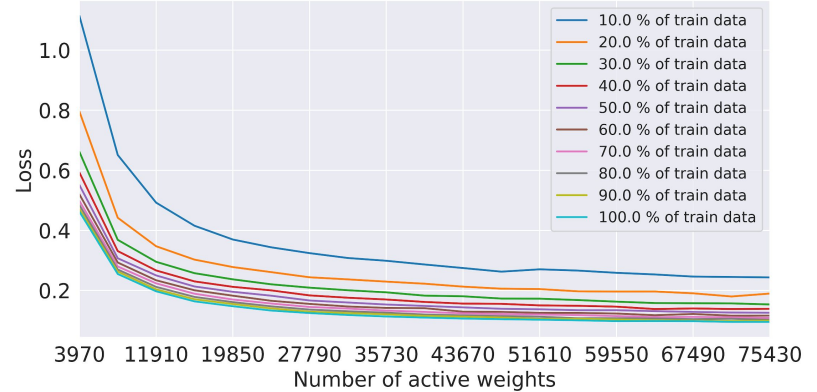


Our Work - Sparse vs Dense NN - 1 Hidden Layer

Dense Neural Network - Average Loss vs Number of Active Weights

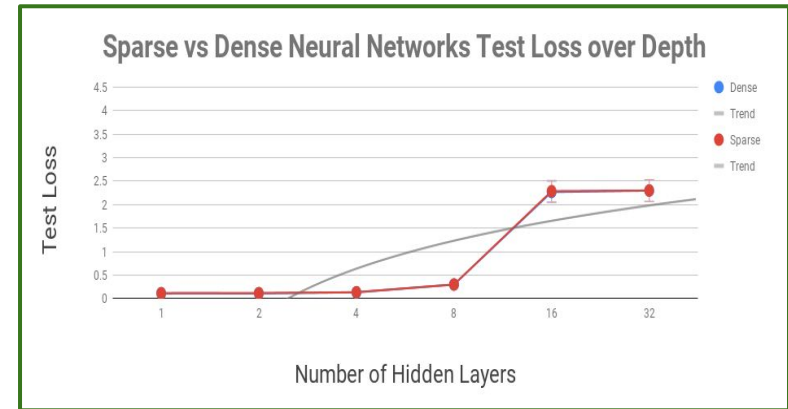
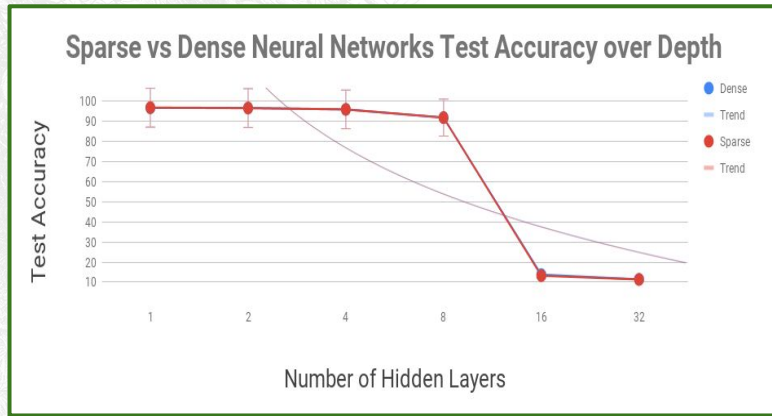


Sparse Neural Network - Average Loss vs Number of Active Weights



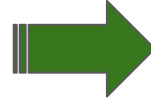
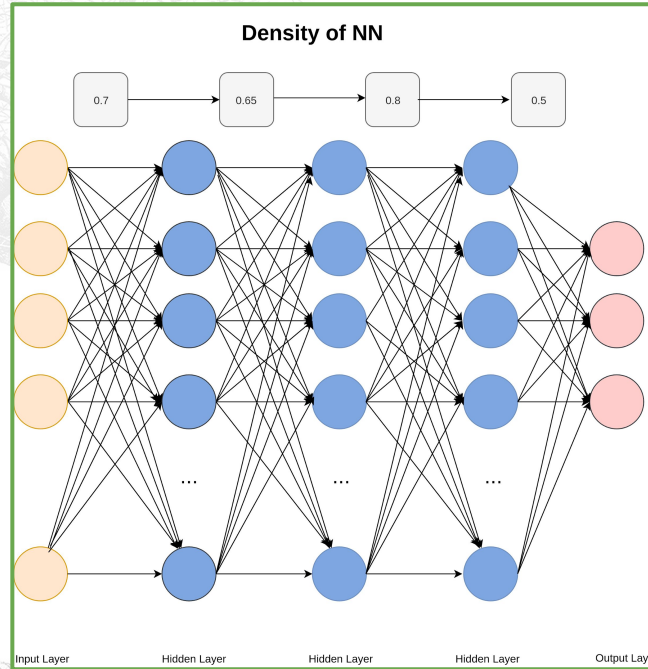
Our Work - Sparse vs Dense NN - Depth

Deeper NN \longrightarrow Behaviour Holds \therefore **Number of weights = Capacity!**



Our Work - Conclusion

Density - % of active weights in a layer.



Preliminary Results - Mnist

- **97.86%** Accuracy.
- **0.072** Test Loss.
- Approx. **60 000** active weights.
- **No tuning of non-architecture hyperparameters!**

Future

- Dynamic Neural Network Architectures.
- Performance Estimation.



References

- Zoph, B., Vasudevan, V., Shlens, J., and Le, Q. V. (2018). Learning transferable architectures for scalable image recognition. In Proceedings of the IEEE conference on computer vision and pattern recognition, pages 8697–8710.
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