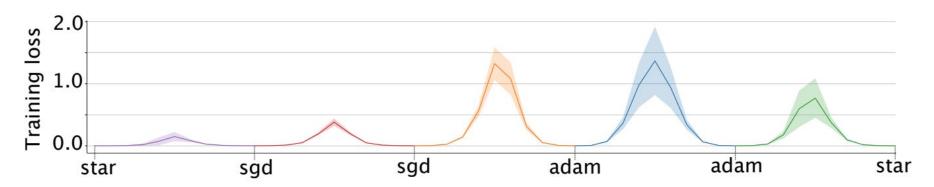








Do Deep Neural Network Solutions form a Star Domain?



Ankit Sonthalia¹, Alexander Rubinstein¹, Ehsan Abbasnejad², Seong Joon Oh¹

¹ Scalable Trustworthy AI at the Tübingen AI Center, University of Tübingen

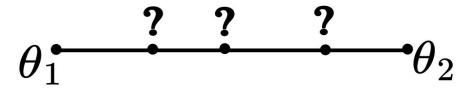
² Australian Institute for Machine Learning

Different seeds lead to different minima

Dataset + architecture + hyperparameters → Model

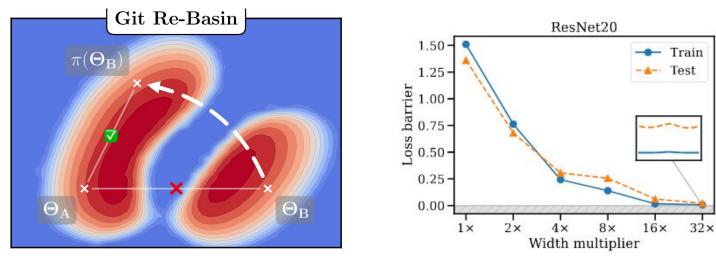
• θ_1 , θ_2 are solutions.

- What about 0.3 x θ_1 + 0.7 x θ_2 ?
- What about t x θ_1 + (1-t) x θ_2 ?



What do NN solution sets look like?

Prior work: Given *sufficient width*, all solutions are linearly connected, modulo permutations. [1, 2]



(Figures from [2])

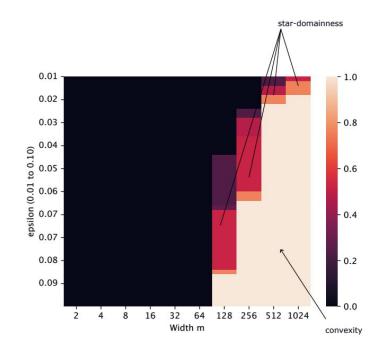
[1] Entezari, R., Sedghi, H., Saukh, O., & Neyshabur, B. The role of permutation invariance in linear mode connectivity of neural network. ICLR 2022. [2] Ainsworth, S. K., Hayase, J., & Srinivasa, S. Git re-basin: Merging models modulo permutation symmetries. ICLR 2023.

Our claim: at *smaller* widths, star-domainness holds.

Claim: Width increases → star-domainness increases until solution set becomes convex.

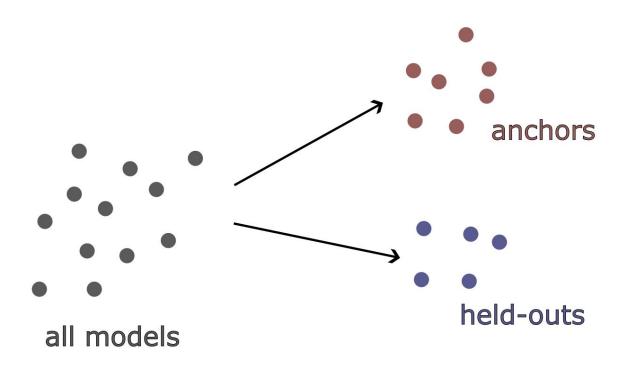
Toy experiment: 2-layer linear networks (minimal non-convex NN).

Main paper: real-world datasets.



The Starlight Algorithm: explicitly finding θ^*

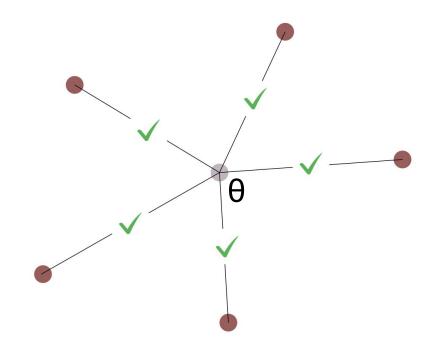
Train several solutions identically, except for the random seed.



How to find θ^* ?

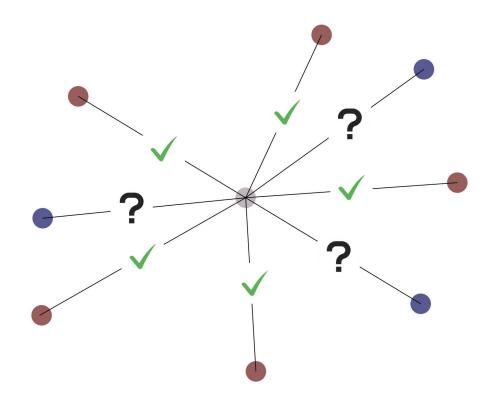
Use the anchors to train a "star model", θ .

- Force θ to be connected to all anchors.
- Use Monte-Carlo optimization.

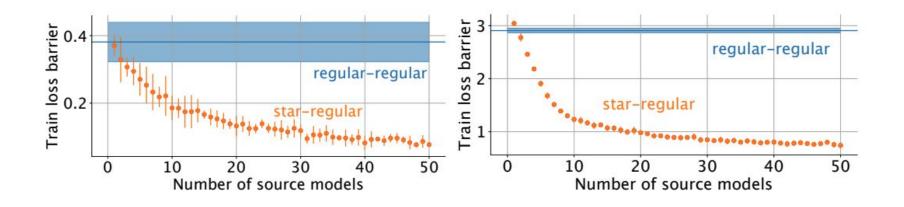


Is θ the θ * that we were seeking?

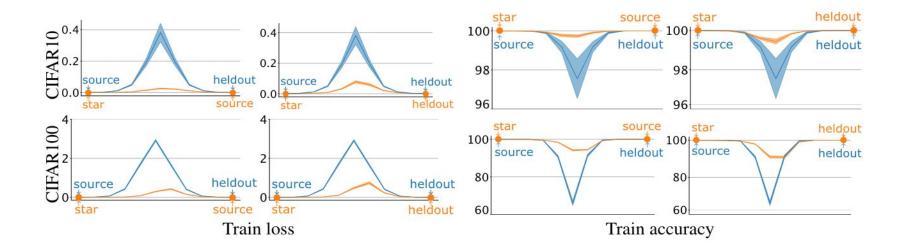
Use the held-outs to test this.



Result: Barriers with held-out models decrease #anchors increases.



Result: Barriers with held-out models decrease #anchors increases.



Code is available at https://github.com/aktsonthalia/starlight.