



A statistical test of eigenvector angles to evaluate the similarity of neural network simulations

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Theory Null hypothesis: Given two independent matrices A and B of type $\mathbf{Y}_N = \mathbf{X}\mathbf{X}^T$, where \mathbf{X} is a Distribution of angles $(\alpha N) \times N$ random matrix whose entries are independent identically between random vectors in \mathbb{R}^N [1] distributed random variables with mean 0, variance $\sigma^2 < \infty$ and $N \to \infty$. $f(\phi) \propto \sin^{N-2}(\phi) \qquad \phi \in [0,\pi]$ 15D angle distribution 1.5-Weighted angle smallness Define angle smallness as



Random eigenvalue distribution [2]

$$\int_{\lambda_{-}}^{\lambda_{+}} \tilde{f}_{N}(\frac{\Delta}{\lambda}) \cdot h_{\alpha}(\lambda) \cdot \frac{d\lambda}{\lambda}$$
-smallness (15D, $\alpha = 500$)
$$\int_{N}^{\infty} \int_{0}^{\infty} Define \text{ score as}} \int_{0}^{N} ple \text{ distribution } g_{N,\alpha}$$

$$\eta = \frac{1}{N} \sum_{i}^{N} w_{i} \cdot \Delta$$

1.0

 $p(\eta) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{\eta^2}{2\sigma^2}\right) \qquad \sigma^2 = \frac{1}{N} \int x^2 \cdot g_{N,\alpha}(x) \, dx$

0.5

1.ÖO

1.05

sample distribution

0.95

2.

0.90

Are two networks similar?

A novel statistical test based on random angles and eigenvalues can answer that.

 Δ_i

interactive notebook

time [s]

neurons

The test can detect (p < 0.05) isolated hubs in 100-neuron networks of size > 4 and correlation > 0.2.

The test doesn't necessarily rely on a prevalent structure to detect similarity.

References

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