

# reading list

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<https://arxiv.org/pdf/1605.08361>

Du, S.S., Zhai, X., Póczos, B. and Singh, A., 2018. Gradient descent provably optimizes over-parameterized neural networks. ICLR 2019

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<http://proceedings.mlr.press/v97/allen-zhu19a/allen-zhu19a.pdf>

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Liu, C., Zhu, L. and Belkin, M., 2022. Loss landscapes and optimization in over-parameterized non-linear systems and neural networks. Applied and Computational Harmonic Analysis.

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<https://arxiv.org/abs/2003.00307>