

# **Favourable Landscape of Phase Retrieval Problem with Optimal Sampling Complexity**

**(27 June 2018)**

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There are many efficient numerical solvers based on non-convex optimizations for the phase retrieval problem. Despite possible local minima of nonconvex objective functions, these algorithms often work remarkably well to find the global minimum. To explain this phenomenon theoretically, there are two frameworks. One theoretical framework is based on the analysis of the nonconvex objective functions locally in a small neighbourhood of global minimizers. It first constructs a special initialization that is close enough to a global minimizer, and then proves convergence of nonconvex algorithms to the global minimizer. This explanation usually needs only  $O(n)$  samples, the optimal sampling complexity. However, it does not explain why nonconvex algorithm with arbitrary initializations still works well. The other theoretical framework analyzes the nonconvex optimization more globally. It proves that nonconvex objective functions in phase retrieval have a favourable landscape - any local minimum is global. Therefore, it is not an issue to get trapped into a local minimum. To have such a favourable landscape, the best existing result needs  $O(n \log^3 n)$  samples, which is not optimal. In this talk, the speaker proves that, with  $O(n)$  samples, some nonconvex objective functions for phase retrieval can still have the favourable landscape.