

Stat 244 Project Proposal

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We are interested in studying the statistical and computational techniques in GLMs with high-dimensional data. We build upon the well-known lasso method for OLS, understanding the adaptations relevant to GLMS, their computational complexities, and their asymptotic guarantees. To this end, we compiled a list of references which we find interesting.

References

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