



Spectral analysis of large reflexive generalized inverse and Moore-Penrose inverse matrices

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Abstract

A reflexive generalized inverse and the Moore-Penrose inverse are often confused in statistical literature but in fact they have completely different behaviour in case the population covariance matrix is not a multiple of identity. In this paper, we study the spectral properties of a reflexive generalized inverse and of the Moore-Penrose inverse of the sample covariance matrix. The obtained results are used to assess the difference in the asymptotic behaviour of their eigenvalues.

Keywords: reflexive generalized inverse; Moore-Penrose inverse; random matrix theory; free probability.

1 Introduction

Let $\mathbf{Y}_n = (\mathbf{y}_1, \mathbf{y}_2, \dots, \mathbf{y}_n)$ be the $p \times n$ data matrix which consists of n column vectors of dimension p with $E(\mathbf{y}_i) = \mathbf{0}$ and $Cov(\mathbf{y}_i) = \Sigma$ for $i \in 1, \dots, n$. We assume that $p/n \rightarrow c \in (1, +\infty)$ as $n \rightarrow \infty$. This type of limiting behavior is also referred to a "large dimensional asymptotics" or "the Kolmogorov asymptotics". In this case, the traditional estimators perform very poorly and tend to over/underestimate the unknown parameters of the asset returns, i.e., the mean vector and the covariance matrix.

Throughout this paper it is assumed that there exists a $p \times n$ random matrix \mathbf{X}_n which consists of independent and identically distributed (i.i.d.) real random variables with zero mean and unit variance such that

$$\mathbf{Y}_n = \Sigma^{\frac{1}{2}} \mathbf{X}_n, \quad (1)$$