

# Estimating the Log-Partition Function of the Trace Norm Distribution\*

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## Abstract

The partition function of the Trace Norm Distribution quickly becomes too large a number for floating point standard. So, we must instead calculate the logarithm of the partition function. We discuss methods for estimating the logarithm of the partition function.

## 1 Introduction

An important quantity in the trace norm distribution partition function is [1],

$$J \equiv \frac{1}{m!} \int_0^\infty \cdots \int_0^\infty \prod_{i=1}^m e^{-\lambda \sigma_i} \sigma_i^{n-m} \prod_{i < j} |\sigma_i^2 - \sigma_j^2| d\sigma_m \dots d\sigma_2 d\sigma_1. \quad (1)$$

We have discussed a method for estimating this quantity via sampling. In empirical tests, this sampling method seems to be highly effective for tall, skinny matrices (rows  $\gg$  columns, or vice versa). But, this quantity quickly grows too large for standard computer floating point types. So, we would like to estimate  $\log J$  instead. Also, if we can estimate  $\log J$ , we can factor a constant out of  $J$  so that the computations do not overload floating point types.

The gamma distribution pdf is  $g(x|\alpha, \theta) = \frac{x^{\alpha-1} e^{-x/\theta}}{\Gamma(\alpha)\theta^\alpha}$ . Let  $\alpha = n - m + 1$  and  $\theta \equiv 1/\lambda$ . We can rewrite the above integral to include  $g$ ,

$$J \equiv \frac{1}{m!} \int_0^\infty \cdots \int_0^\infty \Gamma^m(\alpha) \theta^{m\alpha} \prod_{i=1}^m g(\sigma_i|\alpha, \theta) \prod_{i < j} |\sigma_i^2 - \sigma_j^2| d\sigma_m \dots d\sigma_2 d\sigma_1. \quad (2)$$

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where  $\Gamma^m(\alpha) \equiv [\Gamma(\alpha)]^m$ . Since  $\log(x)$  is a convex function,

$$\log J \geq \int_0^\infty \cdots \int_0^\infty \prod_{i=1}^m g(\sigma_i|\alpha, \theta) \left[ m \log \Gamma(\alpha) + m\alpha \log \theta + \sum_{i<j} \log |\sigma_i^2 - \sigma_j^2| \right] - \log m!. \quad (3)$$

Thus, we can use the sampling scheme from [1] to estimate a lower bound on  $\log J$ . We can either use this as our estimate of  $\log J$ , or use this to scale values in the original sampling procedure to avoid numerical difficulties.

## References

- [1] J. D. M. Rennie. Computing the trace norm distribution via sampling. <http://people.csail.mit.edu/~jrennie/writing>, January 2006.