

A Simple Formula for the Moments of Unitarily Invariant Matrix Distributions

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Complex Unitarily Invariant Distributions

- **Complex multivariate Gaussian:** Modeling baseband signal measurements.
- **Complex Wishart and complex inverse Wishart:** Sample covariances and inverse sample covariances

$$R = XX^H \text{ and } R^{-1} = (XX^H)^{-1}$$

- **Complex beta:** Likelihood ratios for detection

$$y = \mu Hx + Jz + n \sim \begin{cases} \mathcal{CN}[Jz, \sigma^2 I], & \mathcal{H}_0 (\mu = 0) \\ \mathcal{CN}[\mu Hx + Jz, \sigma^2 I], & \mathcal{H}_1 (\mu > 0) \end{cases}$$

$$\ell = \frac{y^H P_J^\perp E_{JH} P_J^\perp y}{y^H P_J^\perp (I - E_{JH}) P_J^\perp y}$$

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Complex Unitarily Invariant Distributions

- Complex matrix beta:
 - Fisher Information in compressed sensing (Pakrooh et al. 2015)

$$y = x(\theta) + n \quad \text{vs} \quad z = \Phi y$$

$$G = \left[\frac{\partial}{\partial \theta_1} x(\theta) \mid \cdots \mid \frac{\partial}{\partial \theta_p} x(\theta) \right]$$

$$F = \frac{1}{\sigma^2} G^H P_{\Phi^H} G \quad \text{and} \quad \text{CRLB} = F^{-1}.$$

- Detection statistics in passive radar (Howard et al. 2015)

$$\gamma = \frac{|G|}{\prod_{i=1}^M |G_{ii}|}; \quad G = X^H X$$

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Calculating Moments

- **Select prior work:**
 - Multivariate normal: Isserlis (1918)
 - Complex multivariate normal: James (1953); Muirhead (book)
 - Real Wishart and matrix beta: Muirhead (book)
 - Complex Wishart: Maiwald & Kraus (2000); Graczyk *et al.* (2003); Nagar & Gupta (2011).
 - Complex matrix beta: Gupta *et al.* (2009), Pakrooh *et al.* (2015)
- **Goal:** Derive a formula for moments of *all* complex matrix distributions that can be transformed to a unitarily invariant distribution via conjugation by a fixed matrix.

Overview

- **Methodology:** Exploiting the relationship between the moments of unitarily invariant distributions and the joint action of the unitary and symmetric groups on $(\mathbb{C}^N)^{\otimes d}$.
- **Key:** *Schur-Weyl duality*, which relates the irreducible representations of unitary and symmetric groups.
- **Advantage:** A moment formula for all complex multivariate unitarily invariant distribution that separates the combinatorial aspect of moment computation from the calculation of a small number of specific distribution dependent moments.
- **Closest prior work:** Graczyk *et al.* (2003); Exploits the relationship between moments of the Wishart distribution and the symmetric group.

The Moment Problem

- Q is a complex $N \times N$ matrix-valued random variable with probability distribution $F(Q)$.
- The distribution is unitarily invariant, if for all $V \in \mathcal{U}(N)$, the group of $N \times N$ unitary matrices, VQV^\dagger has the same distribution as Q .
- The notation $Q^{\otimes d}$ denotes the tensor power $Q \otimes Q \otimes \cdots \otimes Q$, with d factors.
- View Q as a random linear operator on \mathbb{C}^N and $Q^{\otimes d}$ as a random linear operator on $(\mathbb{C}^N)^{\otimes d}$.
- $E(Q^{\otimes d})$ is a fixed linear operator on $(\mathbb{C}^N)^{\otimes d}$. Its matrix elements include the moments corresponding to all products of matrix elements of Q with d factors.
- **Problem:** Compute $E(Q^{\otimes d})$.

Group Actions

- Two important group actions can be defined on $(\mathbb{C}^N)^{\otimes d}$.
 - The first is the action of the symmetric group \mathcal{S}_d (group of permutations on d letters) on $(\mathbb{C}^N)^{\otimes d}$ defined by

$$L_\sigma(\mathbf{v}_1 \otimes \cdots \otimes \mathbf{v}_d) = \mathbf{v}_{\sigma(1)} \otimes \cdots \otimes \mathbf{v}_{\sigma(d)}.$$

- The other is the action of the unitary group $\mathcal{U}(N)$ on $(\mathbb{C}^N)^{\otimes d}$ defined by

$$U^{\otimes d}(\mathbf{v}_1 \otimes \cdots \otimes \mathbf{v}_d) = U\mathbf{v}_1 \otimes \cdots \otimes U\mathbf{v}_d.$$

- Under these two group actions $(\mathbb{C}^N)^{\otimes d}$ becomes a unitary representation of the product group $\mathcal{U}(N) \times \mathcal{S}_d$.

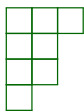
Decomposition into Irreducible Representations

- This representation decomposes into **irreducible representations** $\{W_\lambda\}$:

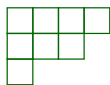
$$(\mathbb{C}^N)^{\otimes d} = \bigoplus_{\lambda} W_\lambda$$

where the sum is over all partitions λ of d with no more than N parts. That is, $\lambda = (\lambda_1, \dots, \lambda_\ell)$, $\ell \leq N$, with the λ_i integers satisfying $\lambda_i \geq \lambda_{i+1}$ and $\lambda_1 + \dots + \lambda_\ell = d$.

- Each λ can be associated with a **Young diagram**:



$$\lambda = (3, 2^2, 1)$$



$$\lambda' = (4, 3, 1)$$

Decomposition into Irreducible Representations

- The decomposition $(\mathbb{C}^N)^{\otimes d} = \bigoplus_{\lambda} W_{\lambda}$ means that linear operators in $\mathcal{U}(N) \times \mathcal{S}_d$ can be **simultaneously block diagonalized** with the blocks labeled by λ .
- **Schur-Weyl duality** states that

$$W_{\lambda} = V_{\lambda} \otimes S_{\lambda}$$

where V_{λ} are non-equivalent unitary irreducible representations of $\mathcal{U}(N)$ and S_{λ} are non-equivalent unitary irreducible representations of \mathcal{S}_d .

Decomposition into Irreducible Representations

- Associated with each W_λ there is a Young orthogonal projection operator Y_λ onto that subspace. These
 - satisfy

$$\sum_{\lambda} Y_{\lambda} = I, \quad Y_{\lambda_1} Y_{\lambda_2} = \delta_{\lambda_1, \lambda_2} Y_{\lambda_1}$$

- and take the form

$$Y_{\lambda} = \frac{\chi_{\lambda}(e)}{d!} \sum_{\sigma \in \mathcal{S}_d} \chi_{\lambda}(\sigma) L_{\sigma}.$$

The $\chi_{\lambda}(\sigma)$ are real integer coefficients which comprise the irreducible *characters* of \mathcal{S}_d . These can be constructed by standard methods (Hamermesh (1962)).

A Key Observation

- The linear operator $E(Q^{\otimes d})$ commutes with the action of $U(N) \times \mathcal{S}_d$ on $(\mathbb{C}^N)^{\otimes d}$.
- So if $E(Q^{\otimes d})$ is restricted to one of the irreducible representations W_λ , then **Schur's Lemma** implies that $E(Q^{\otimes d})$ acts as a multiple, $m_\lambda(Q)$, of the identity.
- Then we simply need to calculate this $m_\lambda(Q)$, which is a coefficient moment of Q .
- Choose a **convenient** unit vector $w_\lambda \in W_\lambda$ in each irreducible representation and compute

$$m_\lambda = \langle w_\lambda, E(Q^{\otimes d}) w_\lambda \rangle.$$

The Main Result

- The d^{th} order moments of Q are given by

$$\mathbb{E}(Q^{\otimes d}) = \sum_{\lambda} m_{\lambda} Y_{\lambda}$$

where the coefficient moments $m_{\lambda}(Q)$ are given by

$$m_{\lambda} = \mathbb{E} \left(M_{\lambda'_1}^{\alpha'_1} M_{\lambda'_2}^{\alpha'_2} \dots M_{\lambda'_k}^{\alpha'_k} \right)$$

where M_{ℓ} denotes the leading principal minor of order ℓ for Q , i.e., the determinant of its top left $\ell \times \ell$ block:

$$\det[\langle \mathbf{e}_i, Q \mathbf{e}_j \rangle]_{i,j=1,\dots,\ell},$$

and $\lambda' = (\lambda'_1, \lambda'_2, \dots, \lambda'_k)$ is the transpose of λ .

The Main Result (Continued)

- If X is a complex matrix-valued random variable with the property that $Q = G^{-1}X(G^{-1})^\dagger$ has a unitarily invariant probability distribution for some fixed non-singular matrix G , then

$$E(X^{\otimes d}) = \sum_{\lambda} m_{\lambda}(Q) R_{\lambda}$$

where $R_{\lambda} = Y_{\lambda}R^{\otimes d}Y_{\lambda}$, with $R = GG^\dagger$.

Computing Moments

- Calculate a d^{th} order moment of elements of $X = GQG^\dagger$.
- For $R = GG^\dagger$, matrix elements of R_λ are

$$(R_\lambda)_{IJ} = \frac{\chi_\lambda(e)}{d!} \sum_{\sigma \in \mathcal{S}_d} \chi_\lambda(\sigma) (R^{\otimes d})_{I\sigma(J)}$$

where $I = (i_1, \dots, i_d)$ and $J = (j_1, \dots, j_d)$.

- We have

$$\mathbb{E}(x_{i_1 j_1} \cdots x_{i_d j_d}) = \mathbb{E}(X^{\otimes d})_{IJ} = \sum_{\lambda} m_\lambda(Q) (R_\lambda)_{IJ}$$

- For $R = I$,

$$\mathbb{E}(q_{i_1 j_1} \cdots q_{i_d j_d}) = \mathbb{E}(Q^{\otimes d})_{IJ} = \sum_{\lambda} m_\lambda(Q) \frac{\chi_\lambda(e)}{d!} \sum_{\sigma: I=\sigma(J)} \chi_\lambda(\sigma)$$

Minor Moments for Some Distributions

- For some complex matrix distributions, the coefficient moments $m_{\lambda}(Q)$ can be computed from just a knowledge of the normalization factors of the distributions.
- Two important examples:

Distribution	m_{λ}
Beta	$\prod_{\ell=1}^k \frac{B_{\lambda'_{\ell}}(a + \sum_{j=1}^{\ell} \alpha'_j, b)}{B_{\lambda'_{\ell}}(a + \sum_{j=1}^{\ell-1} \alpha'_j, b)}$
Wishart/Gamma	$\prod_{\ell=1}^k \frac{\Gamma_{\lambda'_{\ell}}(a + \sum_{j=1}^{\ell} \alpha'_j)}{\Gamma_{\lambda'_{\ell}}(a + \sum_{j=1}^{\ell-1} \alpha'_j)}$

- Also applies to the type II complex beta distribution, the inverse complex matrix beta distribution, and the inverse complex gamma/Wishart distributions.

Generalization of Isserlis' Theorem

- Suppose \mathbf{z} is a unitarily invariant vector-valued random variable.
- If $\mathbf{y} = R^{1/2}\mathbf{z}$, with R a positive-definite hermitian matrix, and $Q = \mathbf{z}\mathbf{z}^\dagger$, then there is only a single term $\boldsymbol{\lambda} = (d)$ in the expansion with $m_{(d)}(Q) = \mathbf{E}(|z_1|^{2d})$:

$$\begin{aligned}\mathbf{E}(y_{i_1}\bar{y}_{j_1} \cdots y_{i_d}\bar{y}_{j_d}) &= R^{\otimes d}\mathbf{E}(Q^{\otimes d}) \\ &= \frac{1}{d!}\mathbf{E}(|z_1|^{2d}) \sum_{\sigma \in \mathcal{S}_d} (R^{\otimes d})_{I\sigma(J)}.\end{aligned}$$

- This is a generalization of Isserlis' Theorem (1918) to unitarily invariant vector distributions.

Concluding Remarks

- The formula for the moments of complex multivariate unitarily invariant distributions derived here provides a powerful tool for
 - dealing with the combinatorial aspect of moment computation
 - and separating them from calculations of a small number of specific distribution dependent moments.
- Our approach could be generalized to compute moments of other classes of distributions.