

# Computations of the Expected Euler Characteristic for the Largest Eigenvalue of a Real Wishart Matrix

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# Largest Eigenvalue of Real Wishart Matrix

Let  $\xi_i \in \mathbb{R}^m$  be distributed as  $N_m(\mu_i, \Sigma)$ .

The Wishart distribution  $W_m(n, \Sigma; \Omega)$  is induced by the random matrix

$$W = \Xi \Xi^\top, \quad \Xi = (\xi_1, \dots, \xi_n) \in \mathbb{R}^{m \times n},$$

where  $\Omega = \Sigma^{-1} \sum_{i=1}^n \mu_i \mu_i^\top$  is the parameter matrix.

We call  $W_m(n, \Sigma; \Omega)$  **non-central** if  $\Omega \neq 0$ .

Let  $\lambda_1(W)$  be the largest eigenvalue of  $W$ . The distribution of  $\lambda_1(W)$  is of particular interest in testing hypothesis.

# Motivation and Previous works

Let  $W_m(n, \Sigma; \Omega)$  be non-central.

**Goal:** Efficient evaluation of  $\Pr(\lambda_1(W) \geq x)$  for many  $x$ .

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- ▶ (James *et al.*, 1954) When  $\Omega = 0$ , express  $\Pr(\lambda_1(W) \geq x)$  as a hypergeometric function  ${}_1F_1$
- ▶ (Hashiguchi *et al.*, 2013) Efficient evaluation of  ${}_1F_1$  using holonomic gradient method
- ▶ (Danufane *et al.*, 2017) In MIMO problem, evaluation of  $\Pr(\lambda_1(W) \geq x)$  if  $W$  is a complex matrix and  $\Omega \neq 0$ .

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**Our contribution:** Efficient evaluation of  $\Pr(\lambda_1(W) \geq x)$  if  $W$  is a **real** matrix and  $\Omega \neq 0$ .

# Euler Characteristic Method

Let  $W_m(n, \Sigma; \Omega)$  be non-central and  $W$  be a real matrix.

**Difficulty:** No explicit formula for  $\Pr(\lambda_1(W) \geq x)$ .

# Euler Characteristic Method

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Adler, Tayler and Takemura (2000, 2005), Kuriki and Takemura (2001, 2008, 2009): Use Euler characteristic heuristic to approximate probabilities of random fields.

**Fact:**  $\lambda_1(W)^{1/2}$  is the maximum of a Gaussian field

$$\{u^T \Xi v \mid \|u\|_{\mathbb{R}^m} = \|v\|_{\mathbb{R}^n} = 1\}.$$

**Idea:** Approximation by the expected Euler characteristic heuristic:

$$\Pr(\lambda_1(W) \geq x) \approx E[\chi(M_x)] \quad \text{when } x \text{ is large,}$$

where  $M_x$  is a manifold induced by  $W$  and  $x$ .

# Outline

- ▶ Explicit formula for the expectation of the Euler characteristic number of a manifold related to a random matrix
- ▶ Numerical evaluation for the integral formula by holonomic gradient method



# Manifold of a Random Matrix

Let  $A$  be a real  $2 \times 2$  random matrix. Define a manifold

$$M = \{hg^T \mid g \in S, h \in S\}.$$

Set

$$f(U) = \text{tr}(UA), \quad U \in M,$$

and

$$M_x = \{U \in M \mid f(U) \geq x\},$$

which is a manifold induced by  $A$  and  $x$ .

# Euler Characteristic Number

Let  $A$  be a real  $2 \times 2$  random matrix and  $M_x$  be the related manifold.

**Recall:** The Euler characteristic is defined for the surfaces of polyhedra by

$$\chi = V - E + F.$$

For convex polyhedron's surface,  $\chi = 2$ .

We can also define the Euler characteristic for  $M_x$  and denote it by  $\chi(M_x)$ .

# Expectation of the Euler Characteristic Number

Let  $A$  be a real  $2 \times 2$  random matrix and  $M_x$  be the related manifold.

**Recall:**  $f(U) = \text{tr}(UA)$ ,  $U \in M_x$ .

Let  $hg^T$  be a critical point of  $f$ . Take  $(g, G) \in SO(2)$  and  $(h, H) \in SO(2)$ . Set

$$\sigma = g^T Ah, \quad b = G^T AH,$$

which are singular values of  $A$ .

**Theorem 1:** Assume  $x > 0$  and  $f(U)$  is a Morse function for almost all  $A$ 's. Then  $E[\chi(M_x)]$  is equal to

$$\frac{1}{2} \int_x^\infty d\sigma \int_{-\infty}^\infty db \int_S G^T dg \int_S H^T dh (\sigma^2 - b^2) p(A).$$

# Expectation of the Euler Characteristic Number

**Recall:** Approximation by the expected Euler characteristic heuristic:

$$\Pr(\lambda_1(W) \geq x) \approx E[\chi(M_x)] \quad \text{when } x \text{ is large,}$$

where  $M_x$  is a manifold induced by  $W = AA^T$  and  $x$ .

**Goal:** Efficient evaluation of the integral in [Theorem 1](#) when  $AA^T$  is a non-central Wishart matrix and  $x$  is large.

# Expectation of the Euler Characteristic Number

Let  $M = \begin{pmatrix} m_{11} & 0 \\ m_{21} & m_{22} \end{pmatrix}$  and  $\Sigma = \begin{pmatrix} 1/s_1 & 0 \\ 0 & 1/s_2 \end{pmatrix}$  such that

$$A = \sqrt{\Sigma}V + M, \text{ where } V = (v_{ij}), v_{ij} \sim \mathcal{N}(0, 1) \text{ i. i. d.}$$

Then the integral in [Theorem 1](#) becomes

$$\int_x^\infty d\sigma \int_{-\infty}^\infty db \int_{-\infty}^\infty ds \int_{-\infty}^\infty f(\sigma, b, s, t) dt, \quad (1)$$

where

$$f = \frac{s_1 s_2 (\sigma^2 - b^2)}{(1 + s^2)(1 + t^2)} \exp\left\{-\frac{1}{2}R\right\}, \quad R \in \mathbb{Q}(\sigma, b, s, t)$$

We denote (1) by  $F(M, \Sigma; x)$ .

# Challenge for Evaluation

Assume  $A = \sqrt{\Sigma}V + M$ , where  $V = (v_{ij})$ ,  $v_{ij} \sim \mathcal{N}(0, 1)$  i. i. d..

- ▶  $F(M, \Sigma; x)$  contains parameters  $M, \Sigma$ .
- ▶ Numerical integration for  $F(M, \Sigma; x)$  is time-consuming and not reliable for many  $x$ .

**Observation:** the integrand of  $F(M, \Sigma; x)$  is holonomic (D-finite).

**Idea:** Use holonomic gradient method to evaluate  $F(M, \Sigma; x)$ .

# Holonomic Gradient Method

$f(\theta, t)$ : unnormalized probability distribution function w.r.t.  $t = (t_1, \dots, t_n)$ , where  $\theta = (\theta_1, \dots, \theta_m)$  is a parameter vector.

$$z(\theta) = \int_{\Omega} f(\theta, t) dt$$

is the **normalizing constant**.  $f(t, \theta)/z(\theta)$  is a probability distribution function on  $\Omega$ . **Evaluation of  $z(\theta)$  is a fundamental problem in statistics.**

**Example:**  $f(\theta, t) = \exp\left(\frac{-t^2}{2\theta^2}\right)$  with  $\Omega = (-\infty, +\infty)$ . Then

$$z(\theta) = \sqrt{2\pi\theta^2}.$$

# Holonomic Gradient Method

An analytic function  $f(x)$  is called **holonomic** or **D-finite** when it satisfies  $n$  linear ODE's (**holonomic system**)

$$\sum_{j=0}^{r_i} a_{ij} \left( \frac{\partial}{\partial x_i} \right)^j f = 0, \quad a_{ij}(x) \in \mathbb{C}[x_1, \dots, x_n], \quad i = 1, \dots, n.$$

**Theorem** (Zeilberger, 1990): If  $f(x)$  is holonomic, then the integral  $\int_{\Omega} f(x) dx_n$  is holonomic in  $(x_1, \dots, x_{n-1})$  (under some conditions on  $\Omega$ ).

**Holonomic Gradient Method** (Nakayama *et al.*, 2011): When  $f(\theta, t)$  is holonomic, the normalizing constant  $z(\theta)$  satisfies a system of linear PDEs, which can be constructed by Gröbner bases. Evaluate  $z(\theta)$  and its derivatives by the system with methods in numerical analysis.



### 3 Steps of Holonomic Gradient Method

1. Construct a Pfaffian system for  $z(\theta)$ .
2. Evaluate numerically  $z(\theta)$  and its derivatives at  $\theta = \theta_0$ .
3. Apply numerical analysis methods for the Pfaffian system.

Example:

$$z(\theta) = \int_{\Omega} \exp(\theta t) t^{1/2} (1-t)^{1/2} dt, \quad \Omega = [0, 1]$$

By creative telescoping,

$$(\theta \partial_{\theta}^2 + (3 - \theta) \partial_{\theta} - 3/2) z = 0, \quad \partial_{\theta} = \frac{\partial}{\partial \theta}$$

Then  $\frac{\partial}{\partial \theta} \mathbf{Z} = P \mathbf{Z}$ , where

$$\mathbf{Z} = \begin{pmatrix} z \\ \frac{\partial}{\partial \theta} z \end{pmatrix}, \quad P = \begin{pmatrix} 0 & 1 \\ \frac{3}{2\theta} & -\frac{3-\theta}{\theta} \end{pmatrix}$$

# Evaluation of the Expected Euler Characteristic

**Recall:**  $E[\chi(M_x)] = F(M, \Sigma; x)$  is equal to

$$\int_x^\infty d\sigma \int_{-\infty}^\infty db \int_{-\infty}^\infty ds \int_{-\infty}^\infty f(\sigma, b, s, t) dt,$$

where  $f$  is hyperexponential over  $\mathbb{Q}(\sigma, b, s, t)$ . Thus,  $-F'(M, \Sigma; x)$  is equal to

$$\int_{-\infty}^\infty db \int_{-\infty}^\infty ds \int_{-\infty}^\infty f(x, b, s, t) dt,$$

**Idea:** Use creative telescoping method to derive an ODE for  $F'(M, \Sigma; x)$

# Creative Telescoping Method

Given a holonomic function  $f(\theta, t)$  with annihilator

$$\text{ann}(f) \subset \mathbb{C}(\theta, t)[\partial_\theta, \partial_t].$$

Find nontrivial

$$P(\theta, \partial_\theta) + \partial_t Q(\theta, t, \partial_\theta, \partial_t) \in \text{ann}(f)$$

Then  $z(\theta) = \int_\Omega f(\theta, t) dt$  satisfies  $P(z) = 0$  (under some conditions on  $\Omega$ ). We call  $P$  a **telescoper** for  $\text{ann}(f)$ .

# Creative Telescoping Method

- ▶ (Zeilberger, 1990): Sylvester's dialytic elimination for **multiple integrals**
- ▶ (Takayama, 1992; Oaku, 1997): D-module theoretical algorithms for **multiple integrals**
- ▶ (Chyzak, 2000): a generalization of Gosper's algorithm for **single integrals of multivariate holonomic functions**
- ▶ (Koutschan, 2010): rational **ansatz** approach for **multiple integrals**
- ▶ (Bostan *et al.*, 2010, 2013; Chen *et al.*, 2015, 2016): reduction-based algorithms for **single integrals of bivariate holonomic functions**

# Chyzak's algorithm

Given a holonomic function  $f(\theta, t)$  with annihilator

$$\text{ann}(f) \subset R = \mathbb{C}(\theta, t)[\partial_\theta, \partial_t].$$

We call  $\dim_{\mathbb{C}}(R/\text{ann}(f))$  the **(holonomic) rank** of  $\text{ann}(f)$ .

**Goal:** Drive an ODE for

$$G(M, \Sigma; x) = \int_{-\infty}^{\infty} db \int_{-\infty}^{\infty} ds \int_{-\infty}^{\infty} f(x, b, s, t) dt,$$

where  $f$  is hyperexponential over  $\mathbb{Q}(\sigma, b, s, t)$ .

Using Chyzak's algorithm, find a holonomic system of rank 2 for

$$f_1(x, b, s) = \int_{-\infty}^{\infty} f(x, b, s, t) dt$$

in 5 seconds using a Linux computer with 15.10 GB RAM.

# Chyzak's algorithm

**Goal:** Drive an ODE for

$$G(M, \Sigma; x) = \int_{-\infty}^{\infty} db \int_{-\infty}^{\infty} f_1(x, b, x) ds,$$

where  $\text{ann}(f_1)$  has holonomic rank 2.

Using Chyzak's algorithm, find a holonomic system of rank 6 for

$$f_2(x, b) = \int_{-\infty}^{\infty} f_1(x, b, s) ds$$

in 16 mins by specifying  $M$  and  $\Sigma$ .

**Question:** Is it possible to compute a holonomic system for  $f_2$  without specifying  $M$  and  $\Sigma$ ?

# Stafford Heuristic

Consider

$$R_n = \mathbb{K}(x_1, \dots, x_n)[\partial_1, \dots, \partial_n],$$

$$T_n = \{\partial_1^{i_1} \cdots \partial_n^{i_n} \mid (i_1, \dots, i_n) \in \mathbb{N}^n\}.$$

**Heuristic:** Given a holonomic system  $H$  in  $R_n$ , compute new holonomic system  $H_1$  in  $R_{n-1}$  s.t.  $H_1 \subset (R_n \cdot H + \partial_n R_n) \cap R_{n-1}$ .

1. Pick  $S_1, S_2 \in T_{n-1}$ .
2. Using rational ansatz method, check existence of telescoper  $P_i$  of  $H$  with support  $S_i$ ,  $i = 1, 2$ . If  $P_i$  exists, go to step 3. Otherwise, go to step 1.
3. Compute Gröbner basis  $H_1$  of  $\{P_1, P_2\}$ . If  $H_1$  is holonomic, then output  $G_1$ . Otherwise, go to step 1.

**Stafford Theorem:** Every left ideal in  $R_n$  can be generated by 2 elements.

# Stafford Heuristic

**Goal:** Drive an ODE for

$$G(M, \Sigma; x) = \int_{-\infty}^{\infty} db \int_{-\infty}^{\infty} f_1(x, b, x) ds,$$

where  $\text{ann}(f_1) = \langle H \rangle$  has holonomic rank 2.

1. Pick

$$S_1 = \{1, \partial_b, \partial_x, \partial_b^2, \partial_b \partial_x, \partial_x^2, \partial_x^3\},$$

$$S_2 = S_1 \cup \{\partial_b^2 \partial_x, \partial_b \partial_x^2, \partial_b^3\}.$$

2. Using rational ansatz method, find telescoper  $P_i$  of  $H$  with support  $S_i$ ,  $i = 1, 2$ .
3. Compute Gröbner basis  $H_1$  of  $\{P_1, P_2\}$ . We find that  $H_1$  has holonomic rank 6.



# Chyzak's algorithm vs Stafford Heuristic

**Goal:** Drive an ODE for

$$G(M, \Sigma; x) = \int_{-\infty}^{\infty} db \int_{-\infty}^{\infty} f_1(x, b, x) ds,$$

where  $\text{ann}(f_1)$  has holonomic rank 2.

Below is a table of time (seconds) for deriving holonomic systems of

$$f_2(x, b) = \int_{-\infty}^{\infty} f_1(x, b, s) ds.$$

# pars	0	1	2	3	4	5
Chyzak	976	$9.8 \times 10^4$	-	-	-	-
Heuristic	43.49	394.4	8527	$4.3957 \times 10^5$	-	$1.5 \times 10^6$

# Conclusion

Let  $W_m(n, \Sigma; \Omega)$  be non-central and  $W$  be a real matrix.

- ▶ Approximate formula of  $\Pr(\lambda_1(W) \geq x)$  by Euler characteristic method
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Thanks!