

# The maximum likelihood degree of rank 2 matrices via Euler characteristics

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# A mixture of independence models

- 1 Consider a pair of four sided dice: one red die and one blue die  $R_1, B_1$ .
  - 2 Consider a second pair of four sided dice: one red die and one blue die  $R_2, B_2$ .
  - 3 Consider a biased coin  $C = [c_1, c_2]$
- The following map induces a set of probability distributions denoted  $\mathcal{M}_{44} \subset \Delta_{15} \subset \mathbb{R}^{16}$  and is called the model.

$$\Delta_1 \times (\Delta_3 \times \Delta_3) \times (\Delta_3 \times \Delta_3) \rightarrow \mathcal{M}_{44} \subset \Delta_{15} \subset \mathbb{R}^{16}$$

$$c_1 R_1 B_1^T + c_2 R_2 B_2^T = [p_{ij}]$$

- $\mathcal{M}_{44}$  is the set of  $4 \times 4$  nonnegative rank at most 2 matrices.
- $\mathcal{M}_{44}$  is a mixture of two independence models.

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# Collecting data and the likelihood function

## Roll the dice

- Rolling the dice we may observe the following **data**:

$$u = [u_{ij}] = \begin{bmatrix} 160 & 8 & 16 & 24 \\ 32 & 200 & 16 & 8 \\ 8 & 24 & 176 & 32 \\ 16 & 40 & 8 & 232 \end{bmatrix}$$

- To each  $p$  in the set of probability distributions  $\mathcal{M}_{44}$  we assign the **likelihood** of  $p$  with respect to  $u$  by the **likelihood function**:

$$\ell_u(p) = \binom{\sum u_{ij}}{u_{11}, \dots, u_{44}}^{-1} \prod_{ij} p_{ij}^{u_{ij}}.$$

- The probability distribution maximizing  $\ell_u(p)$  on the set of distributions  $\mathcal{M}_{44}$  is called the **maximum likelihood estimate (mle)**.
- The mle is the best point of  $\mathcal{M}_{44}$  to describe the observed data.
- The **statistics** problem is to determine mle's.

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# Applied Algebraic Geometry

The mle can be determined by solving the likelihood equations.

- Instead of  $\mathcal{M}_{44}$ , we consider its **Zariski closure**  $X_{44}$ .
- The **Zariski closure** is described by zero sets of homogeneous polynomials.
- The defining polynomials of  $X_{44}$  are the  $3 \times 3$  minors of

$$\begin{bmatrix} p_{11} & p_{12} & p_{13} & p_{14} \\ p_{21} & p_{22} & p_{23} & p_{24} \\ p_{31} & p_{32} & p_{33} & p_{34} \\ p_{41} & p_{42} & p_{43} & p_{44} \end{bmatrix}$$

and the linear constraint  $p_{11} + p_{12} + \cdots + p_{44} - p_s = 0$ .

- The equations define a projective variety of  $\mathbb{P}^{16}$ : rank at  $\leq 2$  matrices
- We consider the **homogenized likelihood function**  
 $\ell_u(p) = \prod_{ij} (p_{ij}/p_s)^{u_{ij}}$  on  $X_{44}$ .

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# Geometric definition of critical points

Critical points can be determined by solving a system of polynomial equations.

- For the models in this talk, the mle is a **critical point** of the homogenized likelihood function.
- The solutions to the **likelihood equations** are critical points.
- One way to formulate the likelihood equations is to use Lagrange multipliers.
  - ▶ We omit a formal description of the likelihood equations, but instead give a geometric description of critical points.

## Geometric definition of critical points (cont.)

Critical points can be determined by solving a system of polynomial equations.

- Let  $X^\circ$  denote the open variety  $X \setminus \{\text{coordinate hyperplanes}\}$ .
  - ▶  $X^\circ$  is the set of points in  $X$  which have nonzero coordinates.
- The **gradient** of the likelihood function up to scaling equals

$$\nabla \ell_u(p) = \left[ \frac{u_{11}}{p_{11}} \quad \frac{u_{12}}{p_{12}} \quad \cdots \quad \frac{u_{44}}{p_{44}} \quad \frac{u_s}{p_s} \right], \quad u_s := - \sum_{ij} u_{ij}.$$

- ▶ The gradient is defined on  $X^\circ$ .
- We say  $p \in X^\circ$  is a **complex critical point**, whenever  $\nabla \ell_u(p)$  is orthogonal to the tangent space of  $X$  at  $p$  and  $p \in X_{reg}^\circ$ .
- The **mle is a critical point** (in the cases we consider).

# Two experiments and ML degree

## Two experiments

- Consider *vectorized* datasets  $u$  for likelihood function  $\ell_u(p)$  on  $X_{44}$ .
  - ▶  $u = \{160, 8, 16, 24, 32, 200, 16, 8, 8, 24, 176, 32, 16, 40, 8, 232\}$ 
    - ★ 191 complex: 25 real and 166 nonreal
  - ▶  $u = \{292, 45, 62, 41, 142, 51, 44, 42, 213, 75, 67, 63, 119, 85, 58, 70\}$ 
    - ★ 191 complex : 3 real and 188 nonreal
- The # of complex solutions was always 191 (this is the ML degree).
- For *general choices* of  $u$  we get the same number of complex critical points.
  - ▶ This number is called the *ML degree* of a variety.

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## Previous Computational Results

- Consider the **mixture model**  $\mathcal{M}_{mn}$  for  $m$ -sided red dice and  $n$ -sided blue dice. Denote its Zariski closure by  $X_{mn}$ .

### Theorem

*The ML-degrees of  $X_{mn}$  include the following:*

$(m, n)$	3	4	5	6	7	8	9	10	11	12
3	10	26	<b>58</b>	<b>122</b>	<b>250</b>	<b>506</b>	<b>1018</b>	<b>2042</b>	4090	8186
4	26	<b>191</b>	<b>843</b>	<b>3119</b>	<b>6776</b>	?	?	?	?	?

- Reference: “Maximum likelihood for matrices with rank constraints”
  - ▶ J. Hauenstein, [], and B. Sturmfels using Bertini.
- Any conjectures for the first row? (Hint add 6.)
- “Maximum likelihood geometry in the presence of sampling and model zeros” gave supporting evidence for up to  $n = 15$ .
  - ▶ E. Gross and [] using Macaulay2.

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# Euler characteristics and ML degrees

Huh proves that the ML degrees are an Euler characteristic in the smooth case.

- Let  $X$  be a **smooth** variety of  $\mathbb{P}^{n+1}$  defined by homogeneous polynomials and the linear constraint

$$p_0 + p_1 + \cdots + p_n - p_s = 0.$$

- Let  $X^\circ$  denote the open variety  $X \setminus \{\text{coordinate hyperplanes}\}$ .

## Theorem [Huh]

The ML degree of the *smooth* variety  $X$  equals the signed Euler characteristic of  $X^\circ$ , i.e.

$$\chi(X^\circ) = (-1)^{\dim X} \text{MLdegree}(X).$$

- The **independence model** (one sided coin) is **smooth**, but the **mixture model** is not.



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# Independence model ML degree

Use Huh's result to give a topological proof.

- Let  $Z$  denote the Zariski closure of the **independence model**, a variety of  $\mathbb{P}^{16}$ .
- The following map gives an algebraic geometry parameterization of  $Z$ .

$$\mathbb{P}^3 \times \mathbb{P}^3 \rightarrow Z$$

$$([r_1, r_2, r_3, r_4], [b_1, b_2, b_3, b_4]) \rightarrow \left[ r_i b_j, \sum_{ij} r_i b_j \right] \text{ where } i, j \in \{1, 2, 3, 4\}.$$

- Let  $\mathcal{O}$  denote  $\mathbb{P}^3 \setminus \mathbf{V}(x_0 x_1 x_2 x_3 (x_0 + x_1 + x_2 + x_3))$ . Then we have a parameterization of  $X^\circ$  given by

$$\mathcal{O} \times \mathcal{O} \rightarrow X^\circ$$

because  $\sum_{ij} r_i b_j = (\sum_i r_i) (\sum_j b_j)$ .

- Using **inclusion-exclusion** and the **additive properties** of Euler characteristics we see that  $\chi(\mathcal{O}) = -1$ .
- By the **product property**  $\chi(\mathcal{O} \times \mathcal{O}) = 1$ .
- This parameterization is a **homeomorphism** thus  $\chi(\mathcal{O} \times \mathcal{O}) = \chi(X^\circ)$ .

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# ML degrees of singular models

The ML degree is a stratified topological invariant.

- Let  $(S_1, S_2, \dots, S_k)$  denote a **Whitney stratification** of  $X^\circ$ .
  - ▶ When  $X^\circ$  is smooth the Whitney stratification is  $(X^\circ)$ .
  - ▶ When  $k = 2$ ,  $S_1 = X_{reg}^\circ$  and  $S_2 = X_{sing}^\circ$ .

## Theorem

Given reduced irreducible  $X^\circ$  with Whitney stratification  $(S_1, \dots, S_k)$ , we have

$$\chi(X_{reg}^\circ) = e_{11} \text{MLdegree}(\bar{S}_1) + e_{21} \text{MLdegree}(\bar{S}_2) + \dots + e_{k1} \text{MLdegree}(\bar{S}_k).$$

- The  $e_{ij}$  are topological invariants called **Euler obstructions**, which can be considered as the topological multiplicity of the singularities.
- This theorem is a corollary of Botong Wang and Nero Budur's result that relates ML degrees to Gaussian degrees.
- The Euler obstruction  $e_{11}$  always equals  $(-1)^{\dim X^\circ}$ .

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# Ternary Cubic Example for Singular Case

We determine the ML degree of a singular  $X$  using the previous theorem.

- Let  $X$  be defined by

$$p_2(p_1 - p_2)^2 - (p_0 - p_2)^3 = p_0 + p_1 + p_2 - p_s = 0.$$

- The Whitney stratification of  $X^\circ$  consists of  $S_1$  the regular points (so  $\bar{S}_1 = X$ ) and  $S_2$  the singular point which is  $[1 : 1 : 1 : 3]$ ,

$$\chi(S_1) = e_{11} \text{MLdegree}(X) + e_{21} \text{MLdegree}(\bar{S}_2).$$

- $S_2$  is a point so  $S_2 = \bar{S}_2$  and  $\text{MLdegree}(\bar{S}_2) = 1$ .
- The Euler obstruction  $e_{21}$  is the signed multiplicity of the singular point, i.e.  $e_{21} = -2$ .
  - In general, the sign depends on the dimension of  $S_2$  and the multiplicity is actually the Euler characteristic of a link [Kashiwara].
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# Returning to the mixture model

We apply the Whitney stratification-ML degree theorem to  $X_{mn}^o$ .

- The Whitney stratification of  $X^o = X_{mn}^o$  is given by  $(S_1, S_2)$  where  $S_1$  are the regular points  $X_{mn}^o \setminus Z_{mn}^o$  and  $S_2$  are the singular points  $Z_{mn}^o$ .
  - ▶ Denote the singular points of  $X_{mn}^o$  by  $Z_{mn}^o$ .
  - ▶  $Z_{mn}^o$  should be thought of as the set of rank 1 matrices ( $Z_{mn}$  is the Zariski closure of the independence model)
- By the theorem we have

$$\chi(X_{mn}^o \setminus Z_{mn}^o) = e_{11} \text{MLdegree}(X_{mn}) + e_{21} \text{MLdegree}(Z_{mn}).$$

- It is already well known  $e_{11} = -1$  and  $\text{MLdegree}(Z_{mn}) = 1$ .
- The first lemma we would prove determines  $e_{21}$ :

$$e_{21} = (-1)^{m+n-1} (\min\{m, n\} - 1).$$

- If we knew  $\chi(X_{mn}^o \setminus Z_{mn}^o)$ , then we would know  $\text{MLdegree}(X_{mn})$ .

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# Returning to the mixture model

We apply the Whitney stratification-ML degree theorem to  $X_{mn}^o$ .

- The Whitney stratification of  $X^o = X_{mn}^o$  is given by  $(S_1, S_2)$  where  $S_1$  are the regular points  $X_{mn}^o \setminus Z_{mn}^o$  and  $S_2$  are the singular points  $Z_{mn}^o$ .
  - ▶ Denote the singular points of  $X_{mn}^o$  by  $Z_{mn}^o$ .
  - ▶  $Z_{mn}^o$  should be thought of as the set of rank 1 matrices ( $Z_{mn}$  is the Zariski closure of the independence model)
- By the theorem we have

$$\chi(X_{mn}^o \setminus Z_{mn}^o) = e_{11} \text{MLdegree}(X_{mn}) + e_{21} \text{MLdegree}(Z_{mn}).$$

- It is already well known  $e_{11} = -1$  and  $\text{MLdegree}(Z_{mn}) = 1$ .
- The first lemma we would prove determines  $e_{21}$ :

$$e_{21} = (-1)^{m+n-1} (\min\{m, n\} - 1).$$

- If we knew  $\chi(X_{mn}^o \setminus Z_{mn}^o)$ , then we would know  $\text{MLdegree}(X_{mn})$ .

# Determining the Euler characteristic $\chi(X_{mn}^o \setminus Z_{mn}^o)$

This is our main theorem.

- If we knew  $\chi(X_{mn}^o \setminus Z_{mn}^o)$ , then we would know  $\text{MLdegree}(X_{mn})$ .
- Let  $\Lambda_m$  be a sequence of  $m - 1$  integers  $(\lambda_1, \lambda_2, \dots, \lambda_{m-1})$ .

## Theorem [ - and B. Wang]

Fix  $m$  greater than or equal to 2. Then, there exists  $\Lambda_m$  such that

$$\chi(X_{mn}^o \setminus Z_{mn}^o) = (-1)^{n-1} \sum_{1 \leq i \leq m-1} \frac{\lambda_i}{i+1} - \sum_{1 \leq i \leq m-1} \frac{\lambda_i}{i+1} \cdot i^{n-1}.$$

- Now we prove the conjecture of Hauenstein, [], Sturmfels.

## Using the main theorem

Fix  $m = 3$ .

$$\chi(X_{3n}^{\circ} \setminus Z_{3n}^{\circ}) = (-1)^{n-1} \left( \frac{\lambda_1}{2} + \frac{\lambda_2}{3} \right) - \left( \frac{\lambda_1}{2} \cdot 1^{n-1} + \frac{\lambda_2}{3} \cdot 2^{n-1} \right).$$

$$\chi(X_{mn}^{\circ} \setminus Z_{mn}^{\circ}) = -\text{MLdegree}(X_{3n}) + (-1)^{3+n-1} (\min\{3, n\} - 1).$$

- $\text{MLdegree}(X_{32}) = 1$  yields the relation  $-\lambda_1 - \lambda_2 = 0$ .
- $\text{MLdegree}(X_{33}) = 10$  yields the relation  $-\lambda_2 = -12$ .

$$\text{MLdegree}(X_{3n}) = (2^{n+1} - 6) + (-1)^n ((\min\{3, n\} - 3))$$

- **Main idea:** For fixed  $m$ , if we knew

$$\text{MLdegree}(X_{m2}), \text{MLdegree}(X_{m3}), \dots, \text{MLdegree}(X_{mm})$$

then we can solve for  $\Lambda_m = (\lambda_1, \dots, \lambda_{m-1})$  thereby giving a closed form expression for  $\text{MLdegree}(X_{mn})$  for all  $n$ .

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- **Main idea** (this slide): **Recursively** determine  $\Lambda_m$  thereby giving a closed form formula for  $\text{MLdegree}(X_{mn})$  for fixed  $m$  but *any*  $n$ .
  - ▶ Note  $\text{MLdegree}(X_{mn}) = \text{MLdegree}(X_{nm})$ .
  - ▶ Prove  $\lambda_{m-1}$  of  $\Lambda_m$  is  $(m-1)m!$ .
- Closed form expressions for fixed  $m$  and  $n \geq m$ :

$$\text{MLdeg}X_{4n} = 25 \cdot 1^{n-1} - 40 \cdot 2^{n-1} + 23 \cdot 3^{n-1}$$

$$\text{MLdeg}X_{5n} = -90 \cdot 1^{n-1} + 260 \cdot 2^{n-1} - 270 \cdot 3^{n-1} + 96 \cdot 4^{n-1}$$

$$\text{MLdeg}X_{6n} = 301 \cdot 1^{n-1} - 1400 \cdot 2^{n-1} + 2520 \cdot 3^{n-1} - 2016 \cdot 4^{n-1} + 600 \cdot 5^{n-1}$$

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# Using Numerical Algebraic Geometry

Witness sets allow us to use parallelizable algorithms.

- Treat the  $u_{ij}$  as parameter values that we can adjust,
- If we have a set of critical points for generic data, then we can solve any specific instance of data quickly using a **parameter homotopy**.
- Critical points of  $\ell_U$  for  $u_{\text{general}}$  are taken to
  - ▶ critical points of  $\ell_U$  for  $u_{\text{specific}}$
  - ▶ by a parameter homotopy

$$\begin{bmatrix} \square & \square & \square & \square \\ \square & \square & \square & \square \\ \square & \square & \square & \square \\ \square & \square & \square & \square \end{bmatrix} \text{ --- } > \begin{bmatrix} 160 & 8 & 16 & 24 \\ 32 & 200 & 16 & 8 \\ 8 & 24 & 176 & 32 \\ 16 & 40 & 8 & 232 \end{bmatrix}$$

191 points --- > 191 points

- ▶  $\square$  denotes a random complex number.

# Thank You

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# Outline

- Statistics
  - ▶ Mixture model
- Applied algebraic geometry
  - ▶ Critical points
- Topology
  - ▶ ML degree
  - ▶ Euler obstructions