

# The Maximum Likelihood Degree of Linear Spaces of Symmetric Matrices

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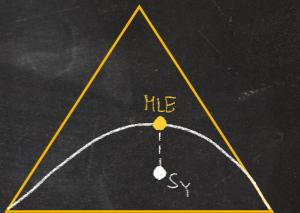
## Maximum likelihood Estimation

- Given:
- $M$ : a statistical model = a set of probability distributions
  - $Y = (Y_1, \dots, Y_n)$ :  $n$  samples of observed data

Goal: Find a distribution in the model  $M$  that best fits the empirical data  $Y$ .

Approach: maximize the likelihood function

$$L_p(Y) := p(Y_1) \cdots p(Y_n) \text{ where } p \in M.$$



A maximum likelihood estimate (MLE) is a distribution in the model  $M$  that maximizes the likelihood.

## Gaussian Models

Density function of  $m$ -dimensional Gaussian with mean zero:

$$P_{\Sigma}(Y) = \frac{1}{\sqrt{\det(2\pi\Sigma)}} \exp\left(-\frac{1}{2} Y^T \Sigma^{-1} Y\right) \text{ where } Y \in \mathbb{R}^m.$$

$\Sigma$  = covariance matrix  
symmetric, positive definite  
matrix in  $\mathbb{R}^{m \times m}$

$K = \Sigma^{-1}$  = concentration matrix  
symmetric, positive definite  
matrix in  $\mathbb{R}^{m \times m}$

A Gaussian model  $M$  is a set of concentration matrices,  
i.e. a subset of the cone of  $m \times m$  symmetric positive definite matrices.

Given data  $Y = (Y_1, \dots, Y_n)$ , the log-likelihood is

$$l_S(K) = \log \det(K) - \text{tr}(KS)$$

where  $S = \frac{1}{n} \sum_{i=1}^n Y_i Y_i^T$  is the sample covariance matrix.

## Linear Concentration Models

Consider a linear Gaussian model  $M$ .

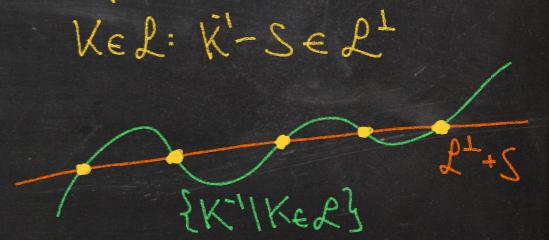
$\Rightarrow$  its Zariski closure in  $\mathbb{S}^m = \{K \in \mathbb{C}^{m \times m} \mid K \text{ symmetric}\}$   
is a linear subspace  $L \subseteq \mathbb{S}^m$ .

trace inner product:  
 $\langle A, B \rangle = \text{tr}(AB)$

The maximum likelihood degree (ML degree) of a linear subspace  $L \subseteq \mathbb{S}^m$   
is the number of complex critical points of  $l_S|_L$  for generic  $S \in \mathbb{S}^m$ .

$$\begin{aligned} l_S(K) &= \log \det(K) - \text{tr}(KS) \\ \Rightarrow \nabla_K l_S &= K^{-1} - S \end{aligned}$$

Assume:  $L$  is regular, i.e.  
 $\exists K \in L : \det(K) \neq 0$ .



## Projective Geometry

The reciprocal variety  $\tilde{L}^*$  of  $L \subseteq \mathbb{P}^m$  is the Zariski closure of  $\{\tilde{K}^* | K \in L\}$  in  $\mathbb{P}^m$ .



For  $L \subseteq \mathbb{P}^m$ , consider projectivizations in  $\mathbb{P}\mathbb{P}^m$ :  $L := \mathbb{P}L$ ,

$$\tilde{L} := \mathbb{P}L^*$$

$$L^\perp := \mathbb{P}L^\perp$$

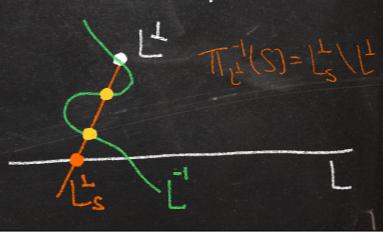
$$L_s^\perp := \mathbb{P}\text{span}\{L^\perp, S\}$$

Prop: The ML degree of a linear space  $L \subseteq \mathbb{P}^m$  is  $|(\tilde{L} \cap L_s^\perp) \setminus L^\perp|$  for generic  $S \in \mathbb{P}\mathbb{P}^m$ .

In other words, the ML degree of  $L$  is the degree of  $\pi_{L^\perp}|_{\tilde{L}}$  where

$$\pi_{L^\perp}: \mathbb{P}\mathbb{P}^m \dashrightarrow \tilde{L}$$

is the projection away from  $L^\perp$ .



## Line Geometry

Consider:

$$\bullet \gamma: \overline{\{(\Sigma, \Gamma, l) \in \tilde{L} \times L^\perp \times \text{Gr}(1, \mathbb{P}\mathbb{P}^m) \mid \Sigma \neq \Gamma, \Sigma \in l, \Gamma \in l\}} \xrightarrow{j \in \text{Gr}(1, \mathbb{P}\mathbb{P}^m)} (\Sigma, \Gamma, l) \mapsto l$$

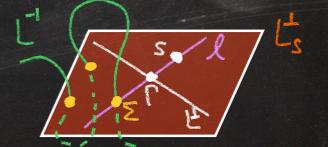
$$\bullet G_S := \{l \in \text{Gr}(1, \mathbb{P}\mathbb{P}^m) \mid S \in l\}$$

Thm [AGKMS]: The ML degree of a linear space  $L \subseteq \mathbb{P}^m$  is  $|\gamma \cap G_S| \cdot \deg(\gamma)$  for generic  $S \in \mathbb{P}\mathbb{P}^m$ .

Otherwise:

$\text{ML degree}(L) = 0$   
 $\Leftrightarrow \gamma \cap G_S = \emptyset$   
 for generic  $S \in \mathbb{P}\mathbb{P}^m$

$= 1$  if  $\text{codim}(L) > 1$   
 point  
 2 ways to span line  
 $\Leftrightarrow \gamma \cap G_S = \{L_s^\perp\}$   
 $\Leftrightarrow L_s^\perp \text{ is a hyperplane}$



## Intersection Theory



Recall:  $\text{ML degree}(L) = |(\tilde{L} \cap L_s^\perp) \setminus L^\perp|$  for generic  $S \in \mathbb{P}\mathbb{P}^m$

(Cor: a)  $\text{ML degree}(L) \leq \deg \tilde{L}$

b) If  $\tilde{L} \cap L^\perp$  is finite & consists only of smooth points of  $\tilde{L}$ , then  $\text{ML degree}(L) = \deg(L) - \deg(\tilde{L} \cap L^\perp)$ .

↪ How to generalize this?

## Intersection Theory

Benedetto Segre  
(1903-1977)

- $\tilde{L} \cap L^\perp \cong \Delta \cap (\tilde{L} \times L^\perp)$  where  $\Delta$  is the diagonal in  $\mathbb{P}\mathbb{P}^m \times \mathbb{P}\mathbb{P}^m$ .
- For  $j \in \{0, \dots, \dim(\tilde{L} \cap L^\perp)\}$ , consider the  $j$ -th Segre class  
 $s^j(\Delta \cap (\tilde{L} \times L^\perp), \tilde{L} \times L^\perp) \in CH_j(\Delta \cap (\tilde{L} \times L^\perp))$ .  
 Chow group of  $j$ -dimensional cycles
- $\sigma^j(\Delta \cap (\tilde{L} \times L^\perp), \tilde{L} \times L^\perp) = \text{degree of } s^j(\Delta \cap (\tilde{L} \times L^\perp), \tilde{L} \times L^\perp)$  under the inclusion  
 $\Delta \cap (\tilde{L} \times L^\perp) \hookrightarrow \Delta \cong \mathbb{P}\mathbb{P}^m$ .

Thm [AGKMS]: The ML degree of a linear space  $L \subseteq \mathbb{P}^m$  is

$$\deg \tilde{L} - \sum_{j=0}^S \binom{N}{j} \sigma^j(\Delta \cap (\tilde{L} \times L^\perp), \tilde{L} \times L^\perp)$$

where  $S := \dim(\tilde{L} \cap L^\perp)$  &  $N := \dim \mathbb{P}\mathbb{P}^m$ .

## Extreme Dimensions

①  $L = \text{point}$

$$\Rightarrow \text{MLdegree}(L) = 1$$

③  $L = \text{hyperplane}$

**Thm [AGKMS]:**  $\{A \in \mathbb{S}^m \mid \text{tr}(AK) = 0\}$   
 let  $\mathcal{L} = \{K \in \mathbb{S}^m \mid \text{tr}(AK) = 0\}$ .  
 $\Rightarrow \text{MLdegree}(\mathcal{L}) = \text{rk}(A) - 1$ .

②  $L = \text{line}$

$GL(m) \curvearrowright \mathbb{S}^m$  congruence action  $\rightsquigarrow GL(m) \curvearrowright \text{Gr}(1, \mathbb{P}\mathbb{S}^m)$

- Weierstraß/Segre: geometric classification of  $GL(m)$ -orbits of lines by Segre symbols Corrado Segre (1863-1924)
- Ferrara/Mandelstam/Sturmfels: formula for MLdegree in terms of Segre symbols

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3x3 matrices

$$\begin{bmatrix} a & x & y \\ x & b & z \\ y & z & c \end{bmatrix} \quad \dim \mathbb{P}\mathbb{S}^3 = 5$$

①  $L = \text{point} \Rightarrow \text{MLdegree}(\mathcal{L}) = 1$

②  $L = \text{line}$

	[1 1 1]	[2 1]	[(1 1) 1]	[3]	[(2 1)]
$\deg L^{-1}$	2	2	1	2	1
$\text{mld}(\mathcal{L})$	2	1	1	0	0

Segre symbols:  
5 congruence classes  
of regular lines

③  $L = 2\text{-plane}$

13 geometric types by CTC Wall

	congruence classes												
	A	B	$B^*$	C	D	$D^*$	E	$E^*$	F	$F^*$	G	$G^*$	H
$\deg L^{-1}$	4	3	4	3	2	4	1	4	2	2	1	2	1
$\text{mld}(\mathcal{L})$	4	3	3	2	2	2	1	1	0	1	0	0	0

DKRS

④  $L = 3\text{-plane}$

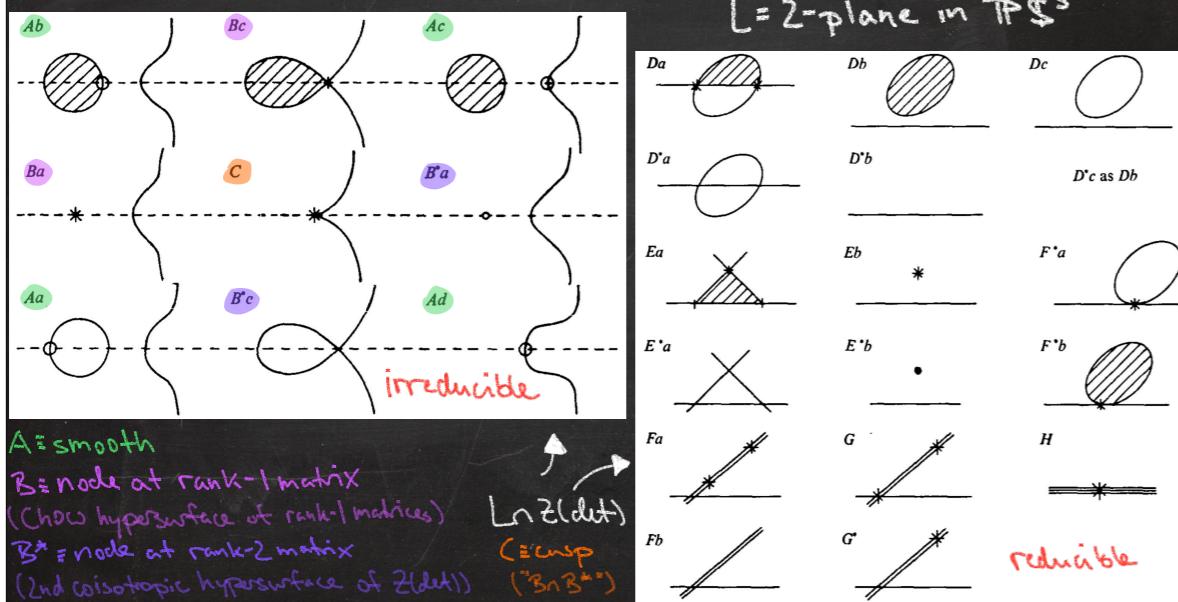
singular  $L^\perp$

	[1 1 1]	[2 1]	[(1 1) 1]	[3]	[(2 1)]	[1; 1]	[1 1; 1]	[2; 1]
$\deg L^{-1}$	4	4	4	4	4	1	4	1
$\text{mld}(\mathcal{L})$	4	3	2	2	1	1	1	0

Segre symbol of  $L^\perp$ :  
8 congruence classes

⑤  $L = \text{hyperplane} \Rightarrow \text{MLdegree}(\{K \in \mathbb{S}^3 \mid \text{tr}(AK) = 0\}) = \text{rk}(A) - 1 \in \{0, 1, 2\}$

## CTC Wall: Nets of Conics



## ML estimation for nets of conics [DKRS]

	A	B	$B^*$	C	D	$D^*$	E	$E^*$	F	$F^*$	G	$G^*$	H
$\deg L^{-1}$	4	3	4	3	2	2	4	1	4	2	0	1	1
$\text{mld}(\mathcal{L})$	4	3	3	2	2	2	1	1	1	1	0	0	0

$L^\perp = \text{Veronese surface in } \mathbb{P}^5$   
 $L^\perp = \text{projection of Veronese surface from plane}$   
 $= \text{plane}$

$L^\perp \cap L^\perp = \emptyset$  1 pt. 1 pt.  $\emptyset$  2 pts. 3 pts. double pt. 1 pt. double line  
 $\emptyset$  1 pt. 1 pt.  $\emptyset$  2 pts. 3 pts. double pt. (smooth) 1 pt. double pt. (smooth)  
 $L^\perp = \text{projection of Veronese surface from point on it}$   
 $= \text{rational normal scroll in } \mathbb{P}^4$

$L^\perp = \text{projection of Veronese surface from tangent line}$   
 $= \text{cone over conic in } \mathbb{P}^3$   
 $L^\perp = \text{projection of Veronese surface from secant line}$   
 $= \mathbb{P}^1 \times \mathbb{P}^1 \subseteq \mathbb{P}^3$   
 Only singular  $L^\perp$

## Extreme ML Degrees

$0 \leq \text{MLdegree}(\mathcal{L}) \leq \deg(\mathcal{L}')$

- $\text{MLdegree}(\mathcal{L}) = \deg(\mathcal{L}') \iff \mathcal{L}' \cap \mathcal{L}^\perp = \emptyset$
- The ML degree of a generic  $L \in \text{Gr}(d, \mathbb{S}^m)$  is  $\deg(\mathcal{L}')$ .
- $NM_{d,m} := \overline{\{L \mid \text{MLdegree}(L) < \deg(\mathcal{L}')\}} \subseteq \text{Gr}(d, \mathbb{S}^m)$

**Thm [JKW]:**  $NM_{d,m} = \text{Bad}_{d,m}$

## Extreme ML Degrees

$0 \leq \text{MLdegree}(\mathcal{L}) \leq \deg(\mathcal{L}')$

**Cor:**  $NM_{d,m} = \text{union of the coisotropic hypersurfaces in } \text{Gr}(d, \mathbb{S}^m)$

associated to the determinantal varieties

$D_s := \{A \in \mathbb{S}^m \mid \text{rk}(A) \leq s\}$  where  $s$  ranges over the integers such that  $\binom{m-s+1}{2} < d \leq \binom{m+1}{2} - \binom{s+1}{2}$ .

Proven for  $\text{Bad}_{d,m}$  by Jiang & Sturmfels.

Chow hypersurface

range [incl.  $\binom{m-s+1}{2}$ ]  
where coisotropic varieties are hypersurfaces

## Extreme ML Degrees

$0 \leq \text{MLdegree}(\mathcal{L}) \leq \deg(\mathcal{L}')$

**Thm [JKW]:**  $NM_{d,m} = \text{Bad}_{d,m}$

- For  $L \in \text{Gr}(d, \mathbb{S}^m_{\mathbb{R}})$ , consider projection  $\beta: \mathbb{S}^m_{\mathbb{R}} \rightarrow \text{Hom}(L, \mathbb{R})$  that is dual to  $L \hookrightarrow \mathbb{S}^m_{\mathbb{R}}$ .  
notion due to Gábor Pataki
- $L$  is called **bad** if the image of the cone of positive semi-definite matrices under  $\beta$  is not closed, equivalently, strong duality in semi-definite programming fails for  $L$ .
- $\text{Bad}_{d,m} := \overline{\{L \in \text{Gr}(d, \mathbb{S}^m_{\mathbb{R}}) \mid L \text{ is bad}\}} \subseteq \text{Gr}(d, \mathbb{S}^m)$

## Extreme ML Degrees

$0 \leq \text{MLdegree}(\mathcal{L}) \leq \deg(\mathcal{L}')$

**Thm [AGKMS]:** The following are equivalent:

- $\text{MLdegree}(\mathcal{L}) = 0$ .
- $\pi|_{\mathcal{L}^\perp}: \mathcal{L}' \dashrightarrow \mathcal{L}$  is not dominant.
- $\text{join}(\mathcal{L}', \mathcal{L}^\perp) \neq \mathbb{P}\mathbb{S}^m$ .
- $(KL^\perp K) \cap \mathcal{L} \neq \emptyset$  for generic  $K \in \mathcal{L}$ .
- For bases  $\{A_1, \dots, A_d\}$  of  $\mathcal{L}^\perp$  &  $\{B_1, \dots, B_d\}$  of  $\mathcal{L}$ ,  $\det(M) \in \mathbb{C}[s_1, \dots, s_d]$  is zero, where  $M_{ij} = \sum_{k,l=1}^d s_k s_l \cdot \text{tr}(A_i B_k A_j B_l)$ .

3x3 matrices  $\begin{bmatrix} a & x & y \\ x & b & z \\ y & z & c \end{bmatrix}$   $\dim \mathbb{P}\mathbb{S}^3 = 5$

①  $L = \text{point} \Rightarrow \text{MLdegree}(L) = 1$

	[1 1 1]	[2 1]	[(1 1) 1]	[3]	[(2 1)]
$\deg L^{-1}$	2	2	1	2	1
$\text{mld}(\mathcal{L})$	2	1	1	0	0

②  $L = \text{line}$

	[1 1 1]	[2 1]	[(1 1) 1]	[3]	[(2 1)]
$\deg L^{-1}$	2	2	1	2	1
$\text{mld}(\mathcal{L})$	2	1	1	0	0

$L^\perp$  &  $L^\perp$  are contained in a common hyperplane

③  $L = 2\text{-plane}$

	A	B	$B^*$	C	D	$D^*$	E	$E^*$	F	$F^*$	G	$G^*$	H
$\deg L^{-1}$	4	3	4	3	2	4	1	4	2	2	1	0	1
$\text{mld}(\mathcal{L})$	4	3	3	2	2	2	1	1	0	1	0	0	0

④  $L = 3\text{-plane}$

	[111]	[21]	[(11)1]	[3]	[(21)]	[;1;]	[11;;1]	[2;;1]
$\deg L^{-1}$	4	4	4	4	4	1	4	1
$\text{mld}(\mathcal{L})$	4	3	2	2	1	1	1	0

here they are not!

⑤  $L = \text{hyperplane} \Rightarrow \text{MLdegree}(\{K \in \mathbb{S}^3 \mid \text{tr}(AK) = 0\}) = \text{rk}(A) - 1 \in \{0, 1, 2\}$

## Final Example

$$L = \{K \in \mathbb{S}^3 \mid \text{tr}(AK) = 0\}, \quad A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \rightsquigarrow L^\perp$$

$$\Rightarrow L^\perp = \mathbb{P} \left\{ \begin{bmatrix} \sigma_{11} & \sigma_{12} & \sigma_{13} \\ \sigma_{21} & \sigma_{22} & \sigma_{23} \\ \sigma_{31} & \sigma_{32} & \sigma_{33} \end{bmatrix} \mid \sigma_{22}\sigma_{33} - \sigma_{23}^2 = 0 \right\} \rightsquigarrow \text{quadric cone}$$

whose vertex set ( $\cong \mathbb{P}^2$ ) contains the point  $L^\perp$

$$\Rightarrow \text{join}(L^\perp, L^\perp) = L^\perp$$

but  $L^\perp$  &  $L^\perp$  are not contained in a common hyperplane