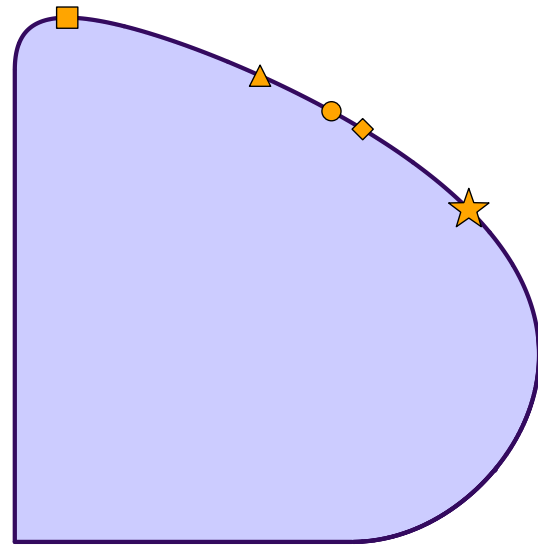
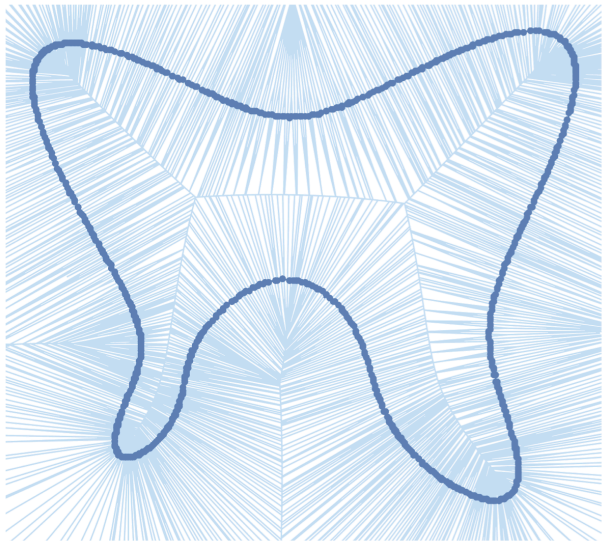


# Distance Optimization

Hannah Friedman

April 15, Metric Algebraic Geometry: Starting Local



# Two Distance Optimization Perspectives

Let  $\mathcal{M} \subseteq \mathbb{C}^n$  be an algebraic variety.

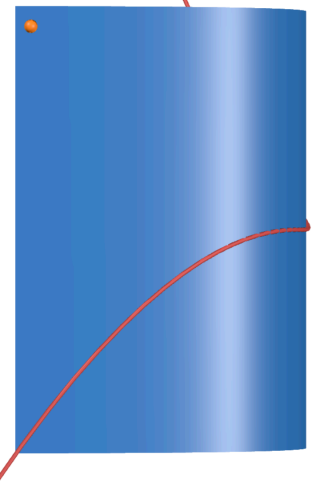
Given a data point  $u \in \mathbb{R}^n$ , what point on  $\mathcal{M}_{\mathbb{R}}$  is **closest** to  $u$ ?

Euclidean distance, Kullback-Leibler divergence, ...

Let  $\mathcal{V}$  be an algebraic variety, and let  $\mathcal{M} \subseteq \mathcal{V}$  be a subvariety.

Given a data point  $u \in \mathcal{V}_{\mathbb{R}}$ , what point on  $\mathcal{M}_{\mathbb{R}}$  is **closest** to  $u$ ?

Euclidean distance, geodesics, ...



# The Euclidean Distance Degree

Let  $\mathcal{M} = V(f_1, \dots, f_r) \subseteq \mathbb{C}^n$  be an algebraic variety of codimension  $c$ .

Given a data point  $u \in \mathbb{R}^n$ , what point on  $\mathcal{M}$  is **closest** to  $u$ ?

$$\min_{x \in \mathcal{M}_{\mathbb{R}}} \|x - u\|^2 = \sum_{i=1}^n (x_i - u_i)^2$$

Critical ideal:

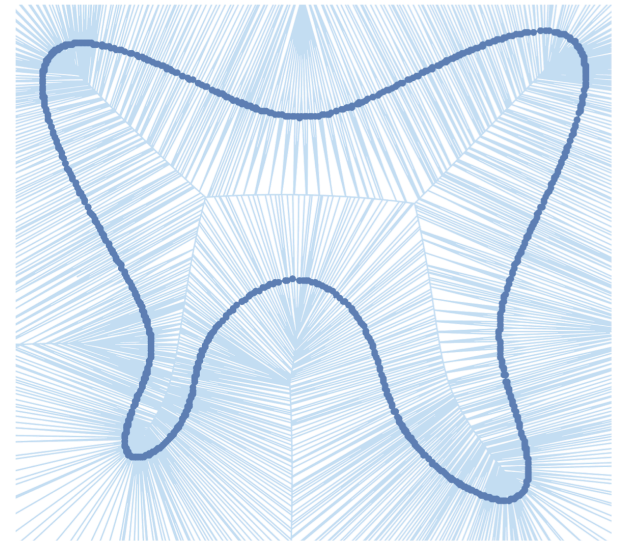
$(\langle f_1, \dots, f_r \rangle + \langle (c+1) \times (c+1) \text{ minors of } \mathcal{J}^u \rangle) : I_{\text{Sing}(\mathcal{M})}^{\infty}$

$$\mathcal{J}^u = \begin{pmatrix} (x - u)^T \\ -\nabla f_1(x) - \\ \vdots \\ -\nabla f_r(x) - \end{pmatrix}$$

For generic data, this ideal is zero dimensional.

The **ED degree** of  $\mathcal{M}$  is its degree.

# Part I: Voronoi Cells



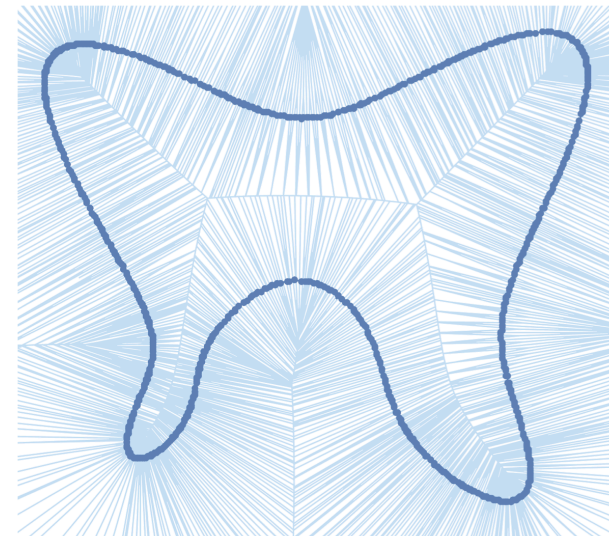
# Voronoi Cells

Let  $\mathcal{M} \subseteq \mathcal{V} \subseteq \mathbb{C}^n$  be an algebraic variety.

Given a data point  $u \in \mathbb{C}^n$ , what point on  $\mathcal{M}$  is **closest** to  $u$ ?

Given a model point  $p \in \mathcal{M}$ , for what points on  $u \in \mathbb{C}^n$  is  $p$  the answer to the above question?

Given a model point  $p \in \mathcal{M}$ , for what points on  $u \in \mathcal{V}$  is  $p$  the answer to the above question?



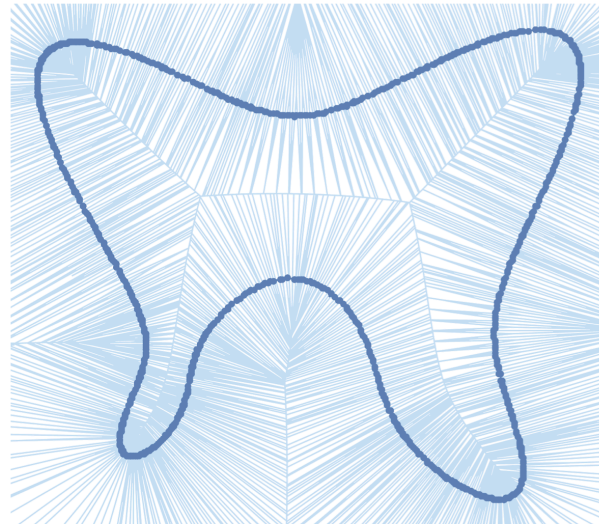
# Voronoi Cells

The Voronoi cells of a point  $p \in \mathcal{M}$  are

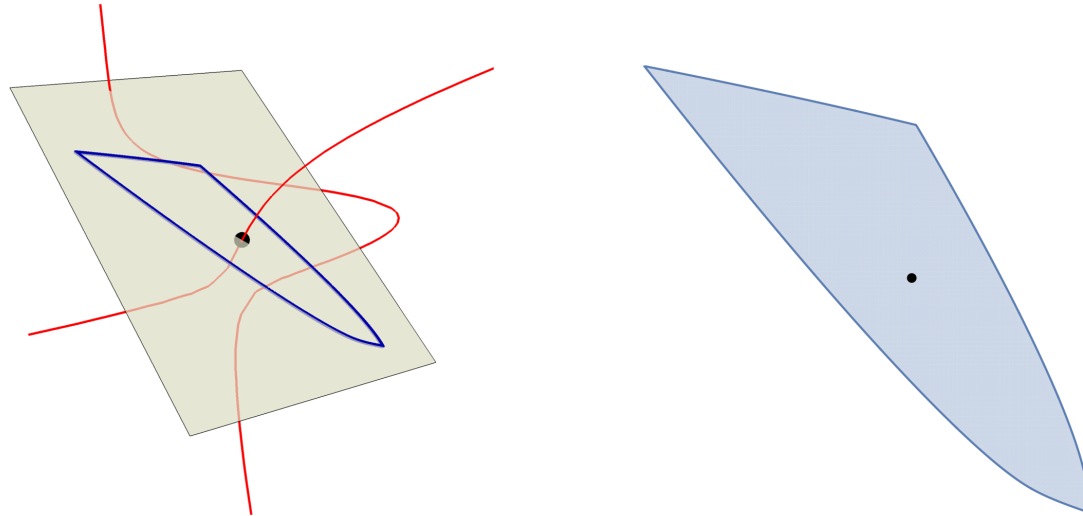
$$\{u \in \mathbb{R}^n : p \in \operatorname{argmin}_{p \in \mathcal{M}} \|u - p\|^2\}$$

$$\{u \in \mathcal{V} : p \in \operatorname{argmin}_{p \in \mathcal{M}} \|u - p\|^2\}$$

The points  $u$  that are equidistant from two points of  $\mathcal{M}$  form the *medial axis* of  $\mathcal{M}$ .



# The Voronoi Boundary of a Point



# Computing the Voronoi Boundary

**Recall:**  $x$  is critical for  $u$  if it is in the vanishing set of this ideal:

$$I_u(x) = (\langle f_1, \dots, f_r \rangle + \langle (c+1) \times (c+1) \text{ minors of } \mathcal{J}^u \rangle) : I_{\text{Sing}(\mathcal{M})}^\infty \quad \mathcal{J}^u = \begin{pmatrix} (x-u)^T \\ -\nabla f_1(x) \\ \vdots \\ -\nabla f_r(x) \end{pmatrix}$$

The Voronoi boundary at the point  $p$  is the zero set of the elimination ideal

$$((I_u(x) + I_u(p) + \langle \|x-u\|^2 - \|p-u\|^2 \rangle) : \langle x-p \rangle^\infty) \cap \mathbb{C}[u_1, \dots, u_n].$$

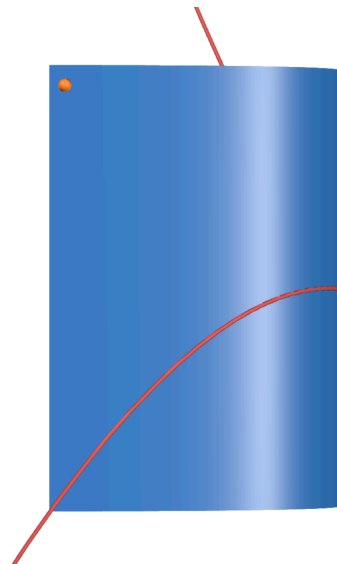
Or one can sample the Voronoi cell by generating data  $u$  such that  $p-u$  is the rowspan of  $\begin{pmatrix} -\nabla f_1(p) \\ \vdots \\ -\nabla f_r(p) \end{pmatrix}$  and then checking if  $p$  is the closest point to  $u$ .

# The Degree of the Voronoi Boundary

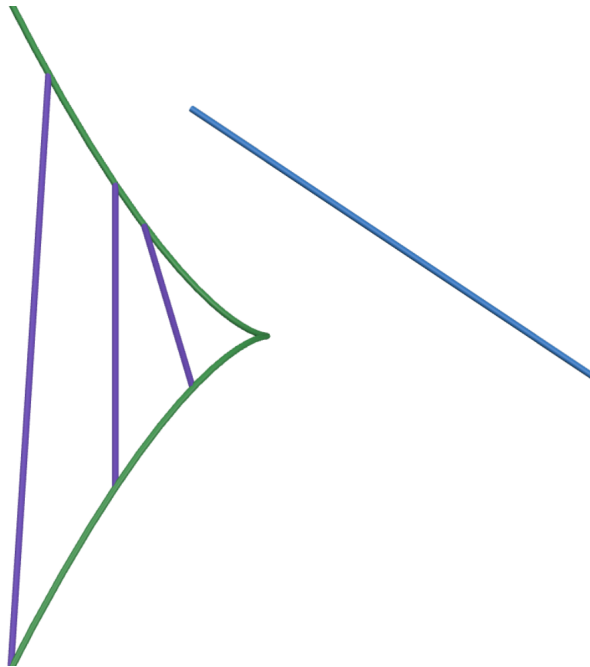
**Theorem (Cifuentes-Ranestad-Sturmfels-Weinstein, 2022).**

Let  $X \subseteq \mathbb{P}^n$  be a curve in general position of degree  $d$  and geometric genus  $g$  with only mild singularities. The degree of boundary of the Voronoi cell at a general point is  $4d + 2g - 6$ .

## Part II: The Grassmann Distance Degree



# Lines in 3-Space



Which secant line is closest to the data line?

# The (Projection) Grassmannian

Today we represent points in the Grassmannian by symmetric projection matrices.

$$\text{pGr}_{\mathbb{R}}(k, n) = \{P \in \text{Sym}^2 \mathbb{R}^n : P^2 = P, \text{trace}(P) = k\}$$

$$\text{pGr}(2, n) = \{P \in \text{Sym}^2 \mathbb{C}^n : P^2 = P, \text{trace}(P) = 2\}$$

This is a smooth, irreducible affine variety of dimension  $2(n - 2)$ .

It is the  $O(\mathbb{C}^n)$  orbit of the matrix  $\begin{pmatrix} \text{Id}_2 & 0 \\ 0 & 0 \end{pmatrix}$ .

# Euclidean Distance Degree

Let  $\mathcal{M} \subseteq \text{pGr}(2, n)$  be a subvariety of the projection Grassmannian.

The **Euclidean distance degree** of  $\mathcal{M}$  is the number of critical points of the optimization problem

$$\min_{P \in \mathcal{M}} \|P - Q\|_F^2 \quad \longleftrightarrow \quad \max_{P \in \mathcal{M}} \text{trace}(PQ)$$

for a **general symmetric matrix**  $Q \in \text{Sym}^2 \mathbb{R}^n$ .

# Grassmann Distance Degree

Let  $\mathcal{M} \subseteq \text{pGr}(2,n)$  be a subvariety of the projection Grassmannian.

The **Grassmann distance degree** of  $\mathcal{M}$  is the number of critical points of the optimization problem

$$\min_{P \in \mathcal{M}} \|P - Q\|_F^2 \quad \longleftrightarrow \quad \max_{P \in \mathcal{M}} \text{trace}(PQ)$$

for a **general projection matrix**  $Q \in \text{pGr}(2,n)$ .

# Extraneous Critical Points

Extraneous critical points are critical for data in the Grassmannian no matter what the model is.

$$\max_{P \in \mathcal{M}} \text{trace}(PQ) \quad Q \in \text{pGr}(2, n)$$

**Proposition (F-Rosana-Sturmfels, 2026).** The  $Q$ -invariant subspaces of dimension 2 are always critical points, regardless of what  $\mathcal{M}$  is. Each of these critical points  $P$  satisfies  $PQ = QP$ .

**Theorem (F-Rosana-Sturmfels, 2026).** For  $n \geq 6$  and a model  $\mathcal{M}$  of dimension at least 4, the ED degree of  $\mathcal{M}$  is strictly larger than the GD degree of  $\mathcal{M}$ .

# Measuring Distance

**Theorem (Ye-Lim, 2016).** Any nice (rotation-invariant) distance between two subspaces in  $\text{Gr}(2,n)$  depends only on the *principal angles* between the subspaces.

## Principal Angles:

If  $P$  and  $Q$  are symmetric projection matrices onto 2-dimensional subspaces of  $\mathbb{R}^n$ , then the matrix  $PQ$  has 2 positive eigenvalues  $\lambda \geq \mu$ .

The principal angles are  $\theta_1 = \cos^{-1}(\sqrt{\lambda})$  and  $\theta_2 = \cos^{-1}(\sqrt{\mu})$ .

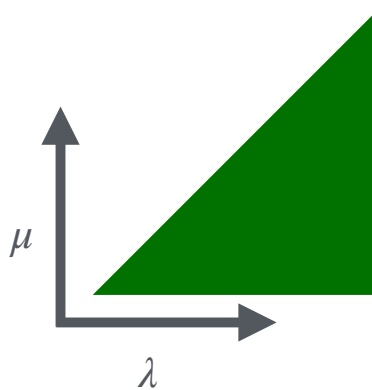
**Upshot:** the goal of any distance minimization problem is to simultaneously minimize the principal angles.

# Measuring Distance

Given a model  $\mathcal{M} \subseteq \text{Gr}(2,n)$  and a data point  $Q \in \text{pGr}(2,n)$ , there is a function

$$\mathcal{M}_{\mathbb{R}} \rightarrow \mathbb{R}^2 \quad P \mapsto (\lambda, \mu)$$

where  $1 \geq \lambda \geq \mu > 0$  are the nonzero eigenvalues of  $PQ$ . The image of this map is the *spectral region*. This is a 2-dimensional subset of the semi algebraic set



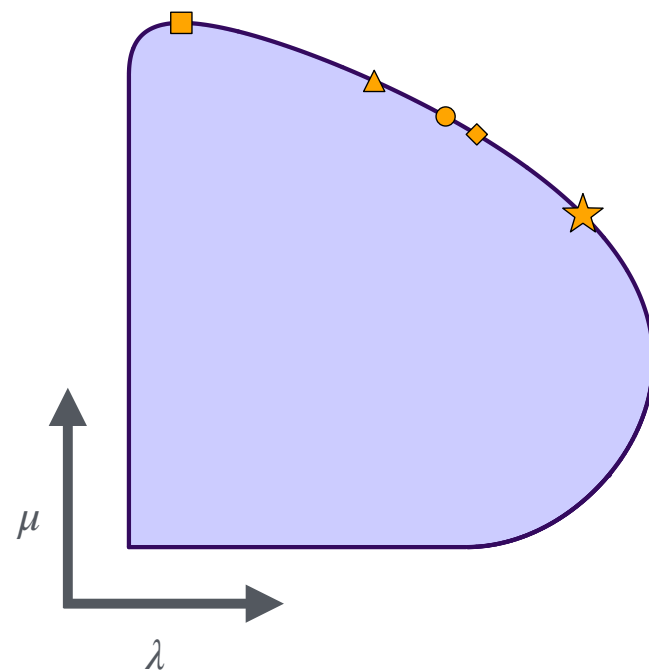
# Example

$$P = \frac{1}{x^2 + y^2 + z^2} \begin{pmatrix} 0 & 0 & 0 & 0 & 0 \\ 0 & x^2 & xy & xz & 0 \\ 0 & xy & y^2 & yz & 0 \\ 0 & xz & yz & z^2 & 0 \\ 0 & 0 & 0 & 0 & x^2 + y^2 + z^2 \end{pmatrix}$$

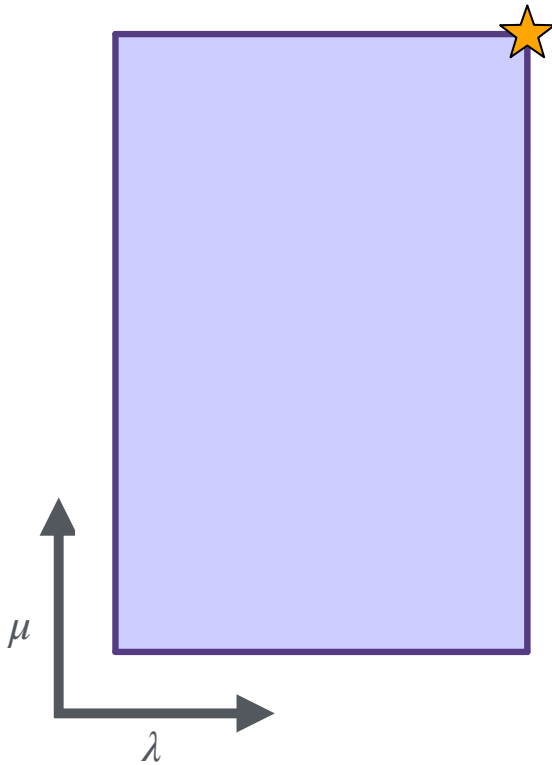
$$Q = \frac{1}{10} \begin{pmatrix} 6 & 4 & 2 & 0 & -2 \\ 4 & 3 & 2 & 1 & 0 \\ 2 & 2 & 2 & 2 & 2 \\ 0 & 1 & 2 & 3 & 4 \\ -2 & 0 & 2 & 3 & 6 \end{pmatrix}$$

# The Spectral Region

	Distance squared in $\lambda, \mu$	Minimizer
Chordal	$4 - 2\lambda - 2\mu$	☆
Geodesic	$\arccos(\sqrt{\lambda})^2 + \arccos(\sqrt{\mu})^2$	○
Procrustes	$4 - 2\sqrt{\lambda} - 2\sqrt{\mu}$	◇
Binet-Cauchy	$1 - \lambda\mu$	△
Fubini-Study	$\arccos(\sqrt{\lambda\mu})^2$	△
Martin	$-\log(\lambda\mu)$	△
Asimov	$\arccos(\sqrt{\mu})^2$	□
Projection	$1 - \mu$	□
Spectral	$2 - 2\sqrt{\mu}$	□



# Snugness



All distances have the same maximal eigenvalues!

Models with this property are called **snug**.

## Theorem (F-Rosana-Sturmfels, 2026).

The following Schubert varieties are snug:

$$\begin{pmatrix} 1 & * & \dots & 0 & \dots & * \\ 0 & 0 & \dots & 1 & \dots & * \end{pmatrix} \quad \begin{pmatrix} 0 & \dots & 1 & 0 & * & \dots & * \\ 0 & \dots & 0 & 1 & * & \dots & * \end{pmatrix}$$

# Questions

- Is the spectral region always convex?
- For a given model and data point, what is the degree of the spectral curve? Is it independent of the data point?
- Are there examples of snug models that are not Schubert varieties?
- What can we say about Voronoi cells of the projection Grassmannian?
- If one intersects a Voronoi cell of the projection Grassmannian with the projection Grassmannian, is the resulting set connected? Convex?
- Is there an analog of the spectral region for other varieties?

# References

- *Metric Algebraic Geometry* (P. Breiding, K. Kohn, B. Sturmfels)
- *The Euclidean Distance Degree* (J. Draisma, E. Horobeț, G. Ottaviani, B. Sturmfels, R. Thomas)
- *Voronoi Cells of Varieties* (D. Cifuentes, K. Ranestad, B. Sturmfels, M. Weinstein)
- *Voronoi Cells in Metric Algebraic Geometry of Plane Curves* (M. Brandt, M. Weinstein)
- *Distance Optimization in the Grassmannian of Lines* (H. Friedman, A. Rosana, B. Sturmfels)