

Minimizing the Condition Number of a Gram Matrix on the Sphere

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Joint work with

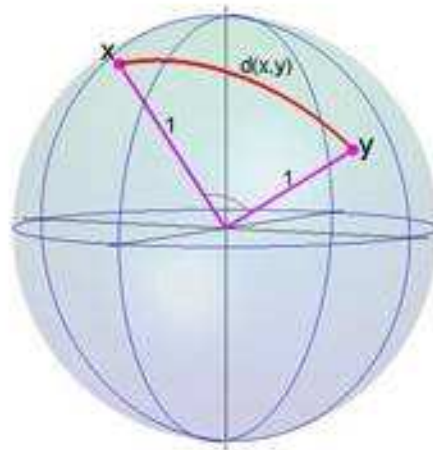
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Least squares approximation on the sphere

$$\mathbb{S}^2 = \{ \mathbf{z} \in \mathbb{R}^3 : \|\mathbf{z}\|_2 = 1 \}, \quad \text{Area } |\mathbb{S}^2| = 4\pi$$



\mathbb{P}_t : the linear space of restrictions of polynomials of degree $\leq t$ in 3 variables to \mathbb{S}^2 .

$$\dim \mathbb{P}_t = (t + 1)^2$$

Gram matrix

\mathbb{P}_t can be spanned by an orthonormal set of **real spherical harmonics** with degree ℓ and order k ,

$$\{Y_{\ell k} \mid k = 1, \dots, 2\ell + 1, \ell = 0, 1, \dots, t\}.$$

Let $X_N = \{\mathbf{z}_1, \dots, \mathbf{z}_N\} \subset S^2$ be **a set of N -points** on the sphere. The Gram matrix is defined as

$$G_t(X_N) = Y(X_N)^T Y(X_N),$$

where $Y(X_N) \in R^{(t+1)^2 \times N}$ and the j -th column of $Y(X_N)$ is given by

$$Y_{\ell k}(\mathbf{z}_j), \quad k = 1, \dots, 2\ell + 1, \quad \ell = 0, 1, \dots, t.$$

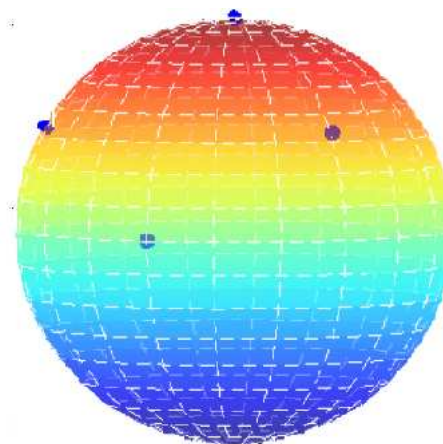
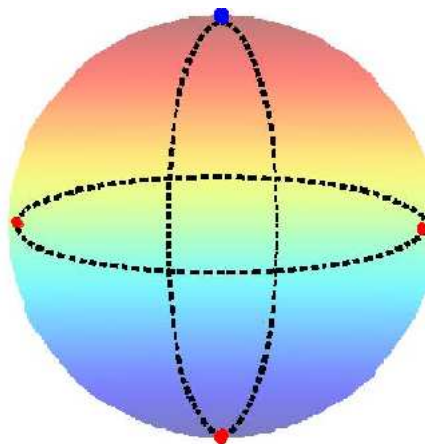
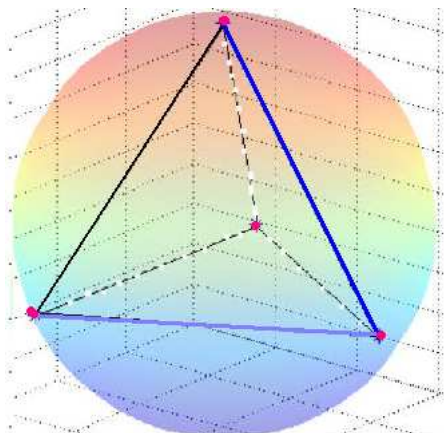
The Gram matrix G_t is a function of a set of N -points X_N .

Distribution of points on the sphere

$$t = 1, \quad \dim P_1 = 4, \quad X_N = \{\mathbf{z}_1, \dots, \mathbf{z}_4\} \subset S^2,$$

$$G_1(X_N) \in R^{4 \times 4}$$

$$\left| \det(G_1(X_N)) = \frac{1}{\pi^4} \right| \quad \left| \det(G_1(X_N)) = 0 \right| \quad \left| \det(G(X_N)) = 0 \right|$$



Regular tetrahedron: $G_1(X_N) = \frac{1}{\pi} I_4$, $\text{cond}(G_1(X_N)) = 1$.

Four sets of points on the sphere

minimum energy system

$$\operatorname{argmin} \sum_{i \neq j}^N \frac{1}{\|\mathbf{z}_i - \mathbf{z}_j\|}$$

extremal system

$$\operatorname{argmax} \det(G_t(X_N))$$

spherical design

$$\int_{\mathbb{S}^2} p(\mathbf{z}) d\mathbf{z} = \frac{4\pi}{N} \sum_{i=1}^N p(\mathbf{z}_i), \quad \forall p \in \mathbb{P}_t$$

minimum cond points

optimal solution of problem:

$$\min \frac{\lambda_{\max}(G_t(X_N))}{\lambda_{\min}(G_t(X_N))}$$

Spherical t -Design

Definition 1 (Delsarte-Goethals-Seidel 1977)

A spherical t -design is a set of N points $X_N = \{\mathbf{z}_1, \dots, \mathbf{z}_N\} \subset S^2$ such that

$$\frac{1}{4\pi} \int_{S^2} p(\mathbf{z}) d\mathbf{z} = \frac{1}{N} \sum_{i=1}^N p(\mathbf{z}_i)$$

for every polynomial $p \in P_t$.

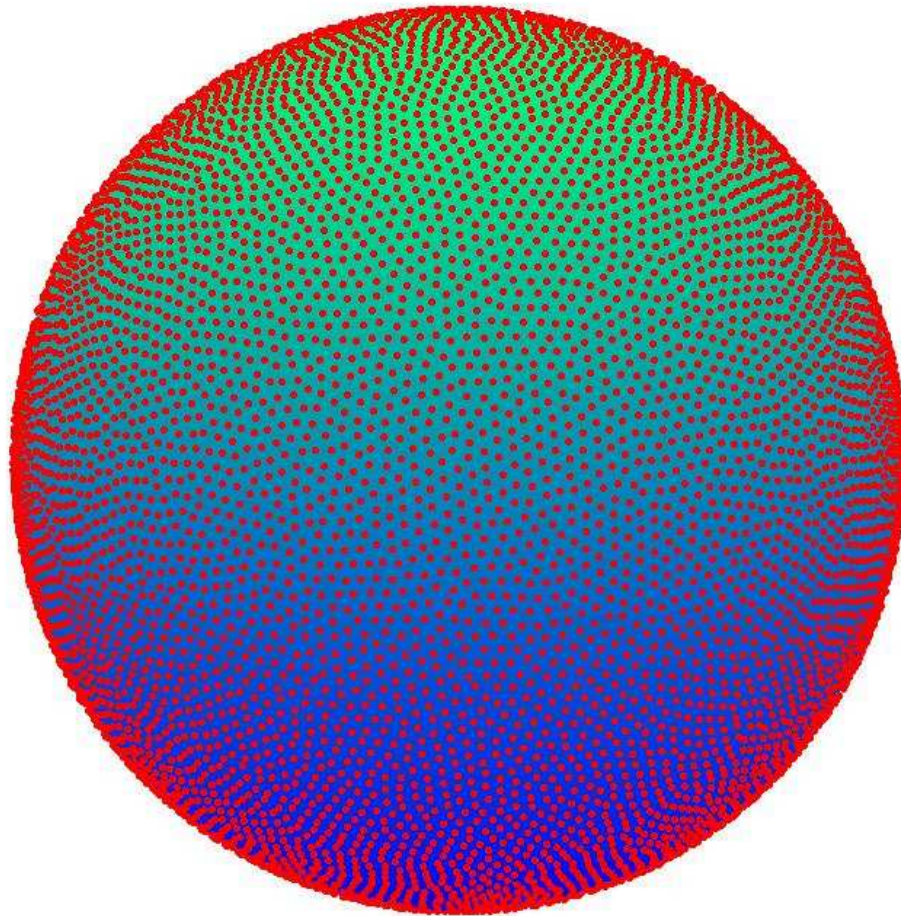
- The average value of $p \in P_t$ on the whole sphere equals the average value of p on the set.
- The equally weighted cubature rule is exact for all $p \in P_t$.

No answer to what is the number of points needed to construct a spherical t -design for any $t \geq 1$? Whether $N = O(t^2)$ as $t \rightarrow \infty$?

Can we guarantee the existence of spherical t -designs with $(t + 1)^2$ points and well-conditioned Gram matrices ?

Spherical 100-design

Chen-Frommer-Lang, to appear in Numer. Math.



Reformulation I: Parametrization

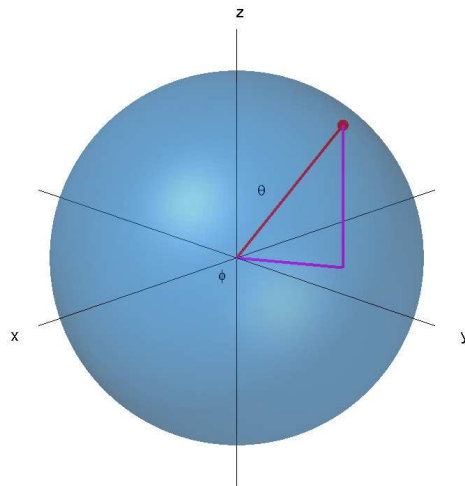
$$N = (t + 1)^2, \quad m = 2N - 3, \quad X_N = \{\mathbf{z}_1, \dots, \mathbf{z}_N\} \subset S^2.$$

Represent $\mathbf{z}_i \in X_N \subset S^2$ using polar coordinates with angles θ_i, ϕ_i .

$$\mathbf{z}_i = [\sin(\theta_i) \cos(\phi_i), \quad \sin(\theta_i) \sin(\phi_i), \quad \cos(\theta_i)]^T, \quad \theta_1 = 0, \phi_1 = \phi_2 = 0$$

Fix \mathbf{z}_1 on the **north pole** and \mathbf{z}_2 on the **zero meridian**.

$$x_\theta = [\theta_2, \dots, \theta_N]^T, \quad x_\phi = [\phi_3, \dots, \phi_N]^T, \quad x = [x_\theta^T, x_\phi^T]^T \in R^m$$



Reformulation II, Gram matrix

Define the **Legendre polynomials** by the recurrence

$$p_0(u) = 1$$

$$p_1(u) = z$$

$$\ell p_\ell(z) = (2\ell - 1)u p_{\ell-1}(u) - (\ell - 1)p_{\ell-2}(u)$$

for $\ell = 2, \dots, t$, $u \in [-1, 1]$.

Define the **Jacobi polynomials**

$$J_t(u) = \sum_{\ell=0}^t (2\ell + 1) p_\ell(u)$$

Define the **Gram matrix** $G(x) \in R^{N \times N}$

$$G_{i,j}(X_N(x)) = J_t(\mathbf{z}_i(x))^T \mathbf{z}_j(x)$$

Euclidean condition number

$V \in R^{\ell \times n}$ with $\ell \geq n$, and $\text{rank}(V) = n$. Let $A = V^T V$

The Euclidean condition number of V is defined by

$$\kappa(V) = \max_{y \neq 0} \frac{\|y\|}{\|Vy\|} \max_{z \neq 0} \frac{\|Vz\|}{\|z\|} = \|V\| \|V^\dagger\| = \sqrt{\kappa(A)} = \frac{\sqrt{\lambda_1(A)}}{\sqrt{\lambda_n(A)}},$$

$V^\dagger = (V^T V)^{-1} V^T$ is the Moore-Penrose generalized inverse of V .

Suppose each entry of $V(x)$ is a continuously differentiable function of $x \in R^m$. Let $A(x) = V(x)^T V(x)$.

We consider

$$\begin{aligned} & \text{minimize} && \kappa(A(x)) \\ & \text{subject to} && x \in \mathcal{X}, \end{aligned} \tag{1}$$

where \mathcal{X} is a convex set in R^m .

A small Vandermonde-like matrix

$$V(x) = \begin{pmatrix} 1 & -x \\ 1 & 0 \\ 1 & x \end{pmatrix}, \quad A(x) = V(x)^T V(x) = \begin{pmatrix} 3 & 0 \\ 0 & 2x^2 \end{pmatrix}$$

$$f(x) = \frac{\lambda_1(A(x))}{\lambda_n(A(x))} = \begin{cases} \frac{3}{2x^2}, & 0.5 \leq x \leq \sqrt{1.5} \\ \frac{2x^2}{3}, & \sqrt{1.5} \leq x \leq 1.5. \end{cases}$$

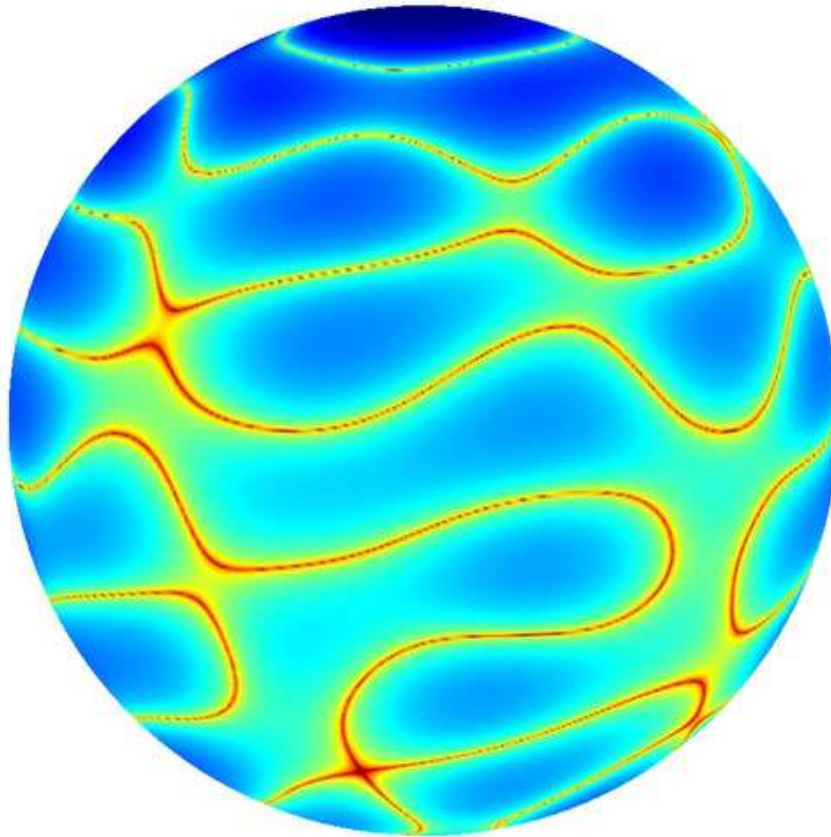
We consider

$$\begin{aligned} & \text{minimize} && f(x) \\ & \text{subject to} && x \in [0.5, 1.5]. \end{aligned}$$

$x^* = \sqrt{1.5}$ is the minimizer. The Clarke generalized gradient at x^* is

$$\partial f(x^*) = \text{conv} \left\{ -2\sqrt{\frac{2}{3}}, \quad 2\sqrt{\frac{2}{3}} \right\} = 2\sqrt{\frac{2}{3}}[-1, 1].$$

An example on the sphere



The condition number for degree $t = 9$, $N = 100$ extremal points. The plot is of the condition number of the Gram matrix G when the point at the north pole is moved to locations over the whole sphere.

Least squares approximation on $[-1, 1]$

Let $\{p_j, j = 0, \dots, n - 1\}$ be a basis for $\mathbb{P}_{n-1}[-1, 1]$, the linear space of polynomials of degree $\leq n - 1$.

For given ℓ distinct real numbers

$$-1 \leq a_1 < a_2 < \dots < a_\ell \leq 1$$

we consider the **weighted Vandermonde-like matrix**

$$V(w, a) = \begin{pmatrix} w_1 p_0(a_1) & w_1 p_1(a_1) & w_1 p_2(a_1) & \dots & w_1 p_{n-1}(a_1) \\ w_2 p_0(a_2) & w_2 p_1(a_2) & w_2 p_2(a_2) & \dots & w_2 p_{n-1}(a_2) \\ \vdots & \vdots & \vdots & & \vdots \\ w_\ell p_0(a_\ell) & w_\ell p_1(a_\ell) & w_\ell p_2(a_\ell) & \dots & w_\ell p_{n-1}(a_\ell) \end{pmatrix}.$$

$$w = (w_1, \dots, w_\ell)^T \quad \text{and} \quad a = (a_1, \dots, a_\ell)^T.$$

Generalized gradient

Let $d(x)$ be the multiplicity of $\lambda_1(A(x))$, and $b(x)$ be the multiplicity of $\lambda_n(A(x))$. Let $A(x)$ admit an eigenvalue decomposition

$$A(x) = U(x)\text{diag}(\lambda(A(x)))U(x)^T$$

with $U(x)^T U(x) = I$. Denote

$$U_\alpha = (u_1(x), \dots, u_{d(x)}(x)), \quad \text{and} \quad U_\beta = (u_{n-b(x)+1}(x), \dots, u_n(x)).$$

Suppose that $\text{rank}(V(x)) = n$. Then f is **Clarke regular** and the **Clarke generalized gradient** $\partial f(x)$ is

$$\{g \in R^m : g_k = \frac{1}{\lambda_n(A(x))} \langle U_\alpha^T A_k(x) U_\alpha, P_\alpha \rangle - \frac{\kappa(A(x))}{\lambda_n(A(x))} \langle U_\beta^T A_k(x) U_\beta, P_\beta \rangle\}$$
$$k = 1, \dots, m$$

where $P_\alpha \in D_{d(x)}^+$, $\text{tr}(P_\alpha) = 1$, $P_\beta \in D_{b(x)}^+$, $\text{tr}(P_\beta) = 1$.

Some properties

1. $f(x) = \kappa(A(x))$ is **strongly semismooth** on \mathcal{X} .
2. Suppose that $V(x)$ is a **linear mapping** of x on \mathcal{X} . Then $\lambda_1(A(x))$ with $A(x) = V(x)^T V(x)$ is a **convex** function on \mathcal{X} .
3. Let B be a fixed $m \times n$ matrix with $m \geq n$ and $\text{rank}(B) = n$. Define

$$h(W) = \kappa(B^T W B), \quad W \in S_m^{++}.$$

Then h is **quasi-convex and strongly pseudo-convex**.

Example: $V(x) = XB$, where $X \in D_m^{++}$ with diagonal elements $x_i, i = 1, \dots, m$, $A(V(x)) = B^T X^T X B = B^T W B$.

Smoothing function I

We introduce the **smoothing function** of the condition number as follows:

$$\tilde{f}(x, \mu) = -\frac{\ln(\sum_{i=1}^n e^{\lambda_i(A(x))/\mu})}{\ln(\sum_{i=1}^n e^{-\lambda_i(A(x))/\mu})}. \quad (2)$$

In numerical computations, we use an equivalent formula

$$\tilde{f}(x, \mu) = \frac{\lambda_1(A(x)) + \mu \ln(\sum_{i=1}^n e^{(\lambda_i(A(x)) - \lambda_1(A(x)))/\mu})}{\lambda_n(A(x)) - \mu \ln(\sum_{i=1}^n e^{(\lambda_n(A(x)) - \lambda_i(A(x)))/\mu})},$$

which is more numerically stable than (2).

Smoothing function II

(i) $\tilde{f}(\cdot, \mu)$ is continuously differentiable for any fixed $\mu > 0$.

(ii) There exists $c > 0$ such that for any $x \in \mathcal{X}$ and $\mu \leq \frac{\lambda_n}{2 \ln n}$

$$0 \leq \tilde{f}(x, \mu) - f(x) \leq c\mu. \quad (3)$$

$$\lim_{x \rightarrow \bar{x}, \mu \downarrow 0} \tilde{f}(x, \mu) = f(\bar{x}).$$

(iii) For any $\bar{x} \in \mathcal{X}$, $\left\{ \lim_{x \rightarrow \bar{x}, \mu \downarrow 0} \nabla_x \tilde{f}(x, \mu) \right\}$ is nonempty and bounded.

$$\left\{ \lim_{x \rightarrow \bar{x}, \mu \downarrow 0} \nabla_x \tilde{f}(x, \mu) \right\} \subset \partial f(\bar{x}).$$

(iv) For any fixed $\mu > 0$, there exists a constant L_μ such that

$$\|\nabla \tilde{f}(x, \mu) - \nabla \tilde{f}(y, \mu)\| \leq L_\mu \|x - y\|. \quad (4)$$

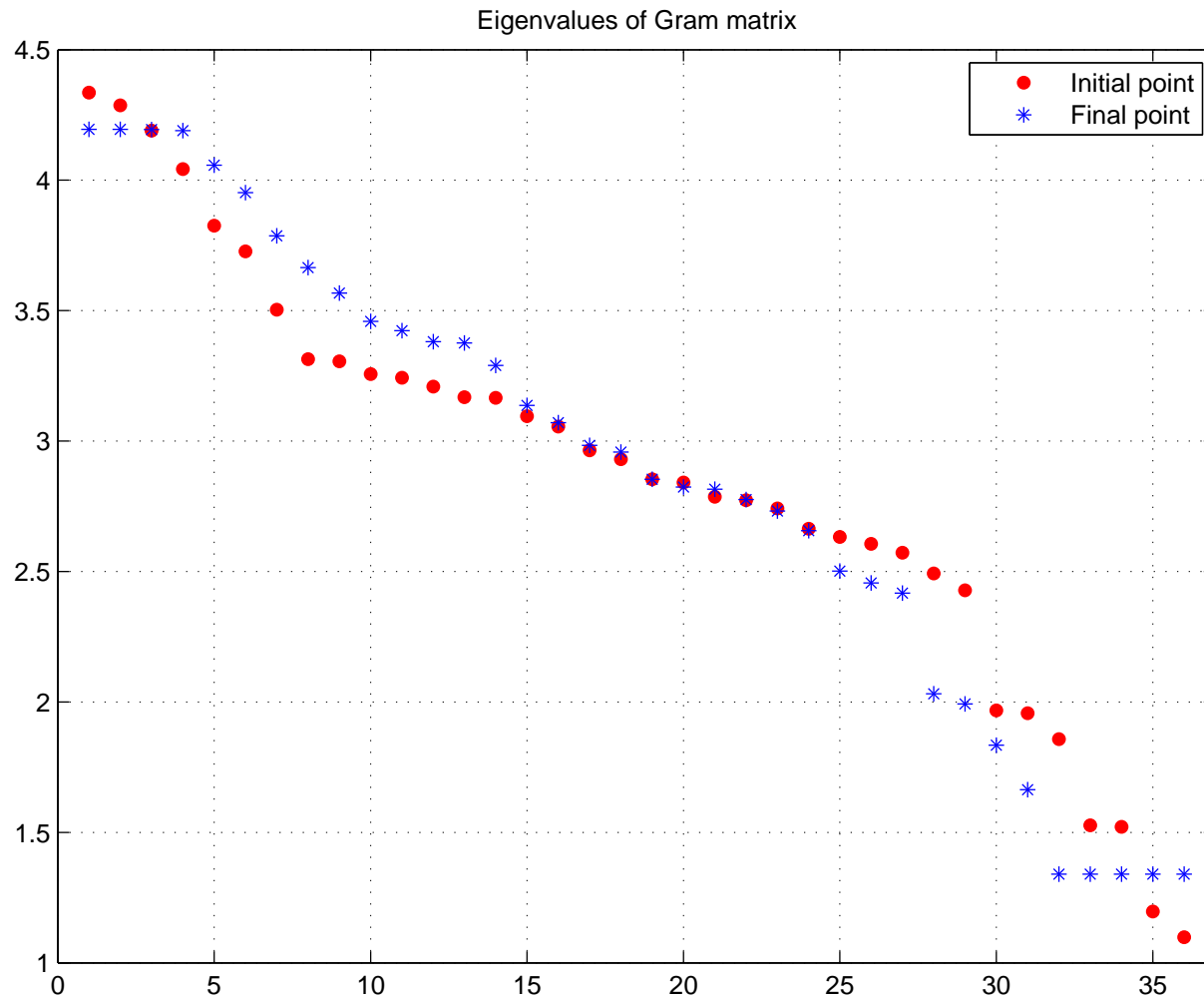
Convergence

Theorem From any starting point $x^0 \in \mathcal{X}$, the sequence $\{x^k\}$ generated by the SPG method is contained in \mathcal{X} and any accumulation point \bar{x} of $\{x^k\}$ is a Clarke stationary point, that is, there is $g \in \partial f(\bar{x})$ such that

$$\langle g, x - \bar{x} \rangle \geq 0, \quad \forall x \in \mathcal{X}.$$

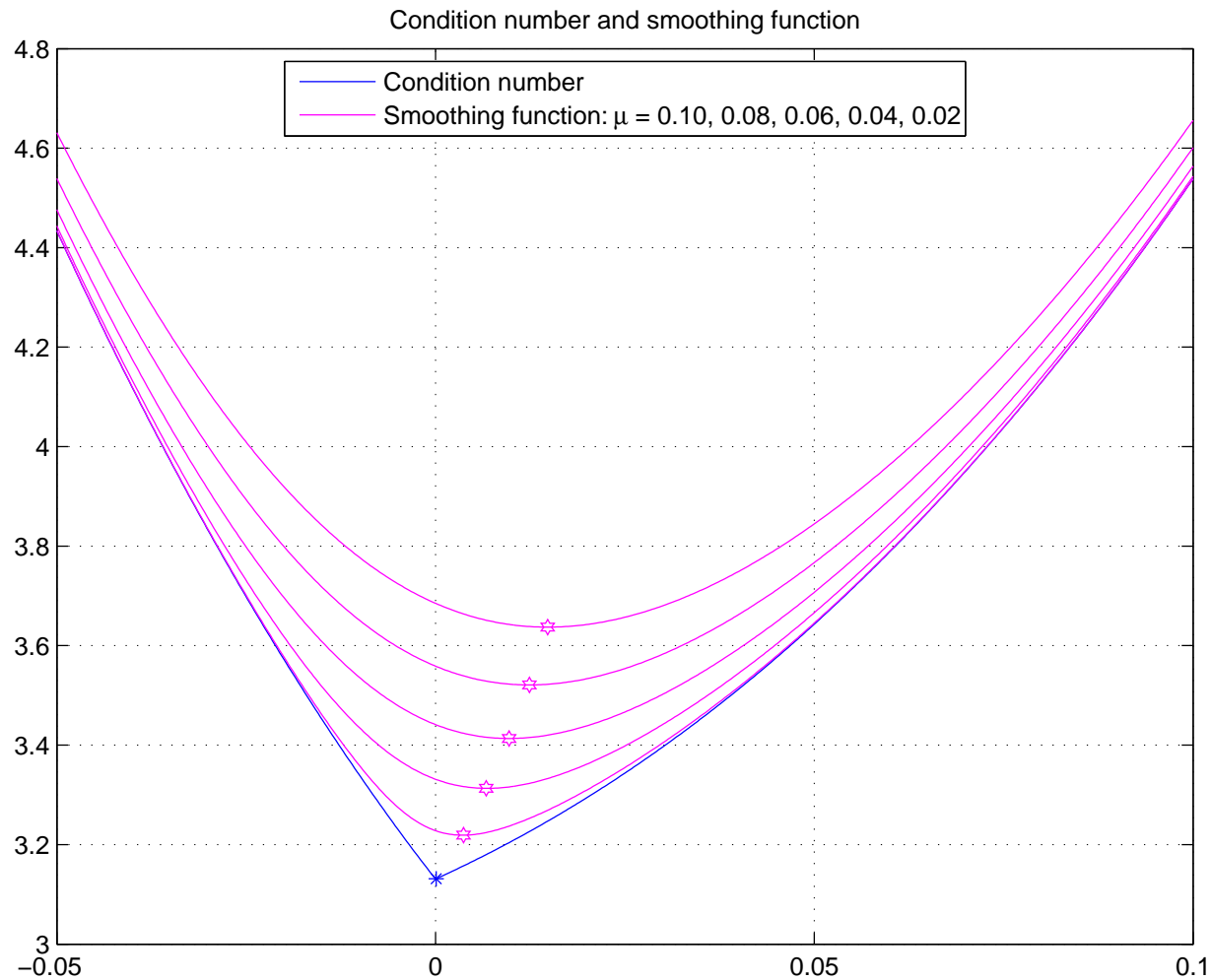
Corollary If the function $f(x)$ is pseudoconvex in a neighborhood $B(\bar{x}) \subset \mathcal{X}$, then the accumulation point is a local optimal solution and if the function $f(x)$ is pseudoconvex on \mathcal{X} , then the accumulation point is a global optimal solution over \mathcal{X} .

Eigenvalues of a Gram matrix on the sphere



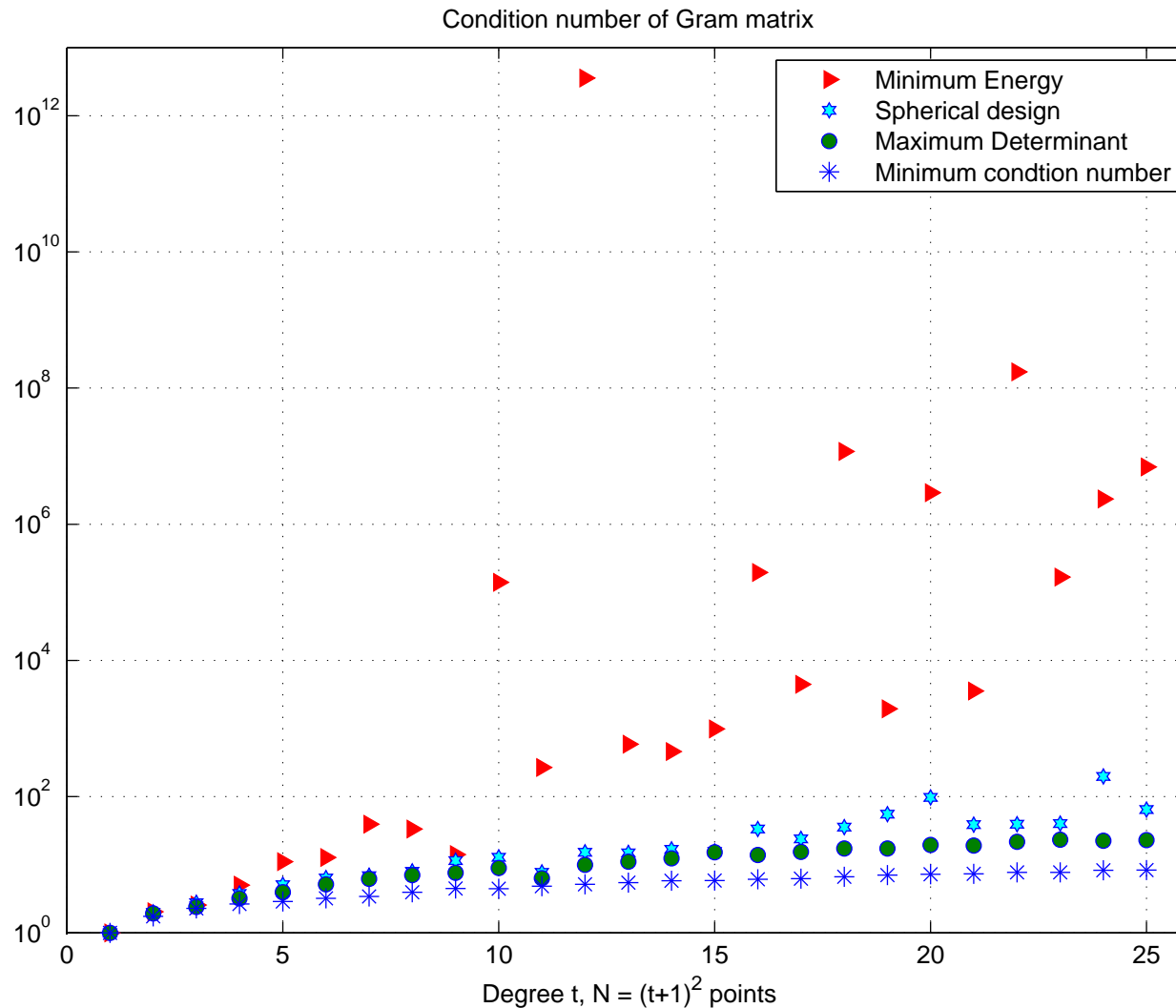
The eigenvalues of 36×36 Gram matrix for degree 5, with 36 points on the sphere.

Smoothing function



For the same Gram matrix, $f(x^* + \alpha \nabla_x \tilde{f}(x^*, 0.0766))$, $\alpha \in [-0.05, 0.1]$.

Condition number on the four sets of points



References

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3. C. An, X. Chen, I.H. Sloan and R.S. Womersley, Well-conditioned spherical designs on 2-sphere, to appear in SIAM J. Numer. Anal.
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Thank you very much

Five sets of points in the interval $[-1, 1]$

equally spaced points $a_i = -1 + \frac{2(i-1)}{\ell-1}, \quad i = 1, \dots, \ell.$

Gauss points $\int_{-1}^1 p(\tau) d\tau = \sum_{i=0}^{n-1} \alpha_i p(a_i), \quad \forall p \in \mathbb{P}_{2n}$

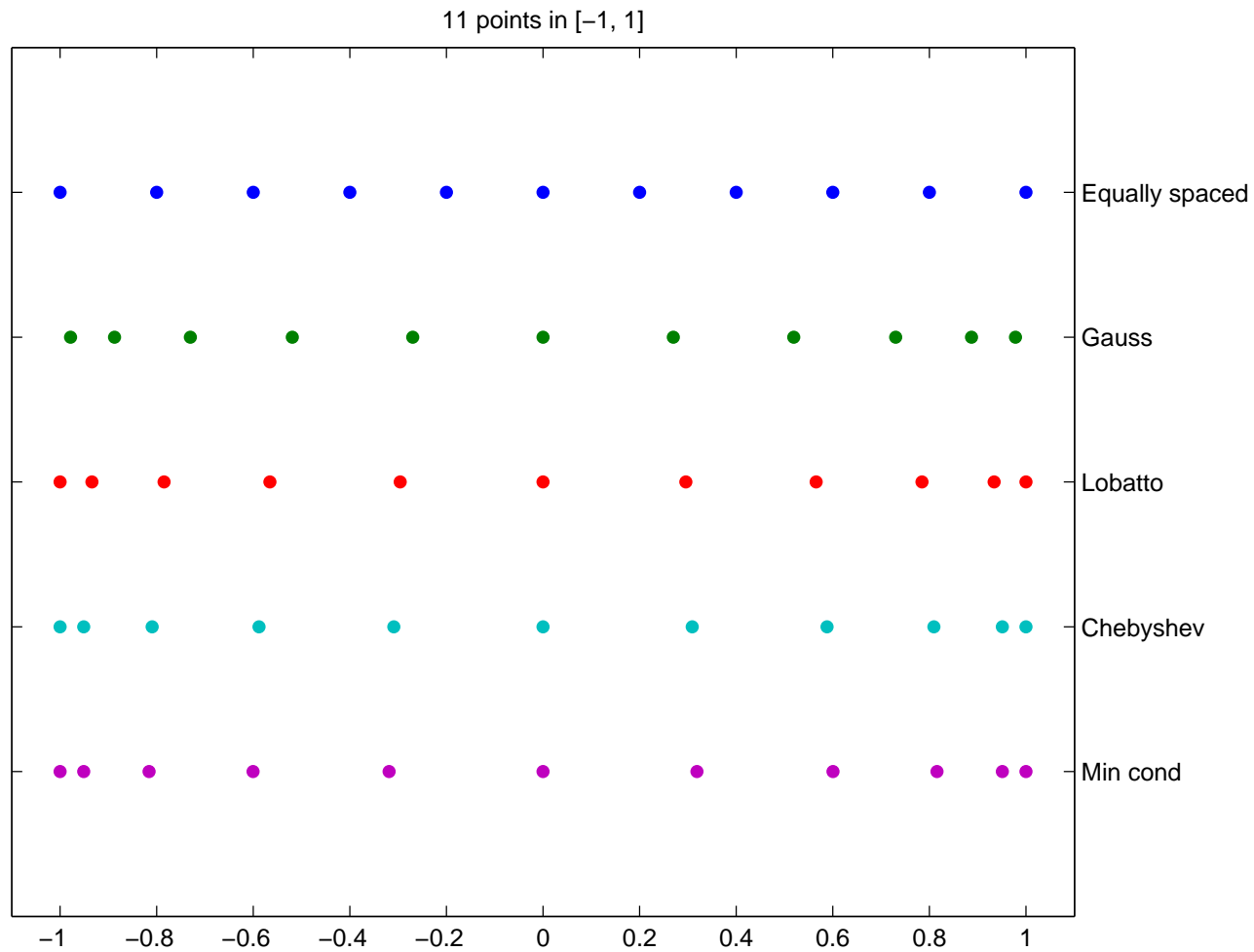
Gauss Lobatto points $a = \operatorname{argmax} \det(V(x)^T V(x)).$

Chebyshev points $a_i = \cos \frac{\pi(2i-1)}{2\ell}, \quad i = 1, \dots, \ell.$

minimum cond points $a = \text{optimal solution of (1)} .$

Here $\alpha_i, i = 0, \dots, n-1$ are the integral values of the Lagrange interpolation polynomials on $[-1, 1]$.

Distribution of points



condition number and determinant

	condition number	determinant
equally spaced points	1.946479e+008	5.755277e-022
Gauss points	1.767123e+007	4.616572e-020
Lobatto points	9.606328e+006	7.968101e-019
Chebyshev points	8.307060e+006	6.310887e-019
min cond points	8.176691e+006	5.826573e-019

Values of the condition number and determinant at equally spaced points, Gauss Points, Gauss Lobatto points, Chebyshev points and minimum cond points