

Lecture 5: Matrix Volume



C. G. J. Jacobi.

Index sets of maximal full-rank submatrices

Given $A \in \mathbb{C}_r^{m \times n}$, we denote

$$\mathcal{I}(A) = \{I \in Q_{r,m} : \text{rank } A[I, *] = r\} ,$$

$$\mathcal{J}(A) = \{J \in Q_{r,n} : \text{rank } A[* , J] = r\} ,$$

$$\mathcal{N}(A) = \{(I, J) : I \in Q_{r,m}, J \in Q_{r,n}, A[I, J] \text{ is nonsingular}\} ,$$

Index sets of maximal full-rank submatrices

Given $A \in \mathbb{C}_r^{m \times n}$, we denote

$$\mathcal{I}(A) = \{I \in Q_{r,m} : \text{rank } A[I, *] = r\} ,$$

$$\mathcal{J}(A) = \{J \in Q_{r,n} : \text{rank } A[* , J] = r\} ,$$

$$\mathcal{N}(A) = \{(I, J) : I \in Q_{r,m}, J \in Q_{r,n}, A[I, J] \text{ is nonsingular}\} ,$$

If

$$A = CR$$

is a **full rank factorization**, then

$$\mathcal{I}(A) = \mathcal{I}(C)$$

$$\mathcal{J}(A) = \mathcal{J}(R)$$

$$\mathcal{N}(A) = \mathcal{I}(A) \times \mathcal{J}(A)$$

Note: $I \in \mathcal{I}(A)$ and $J \in \mathcal{J}(A) \implies A[I, J]$ nonsingular (!?)

The Gramian

The **Gram matrix** of a set $S = \{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_k\} \subset \mathbb{R}^n$ is the $k \times k$ matrix $G(S)$, or $G(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_k)$, of **inner products**

$$G(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_k)[i, j] := \langle \mathbf{x}_i, \mathbf{x}_j \rangle .$$

$\det G(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_k)$ is the **Gramian** of $\{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_k\}$.

If $\{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_k\}$ are columns of $X \in \mathbb{R}^{n \times k}$, then

$$\det G(\mathbf{x}_1, \dots, \mathbf{x}_k) = \det X^T X = \sum_{I \in Q_{k,n}} \det^2 X_{I^*} , \text{ (by C-B)}$$

The Gramian

The **Gram matrix** of a set $S = \{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_k\} \subset \mathbb{R}^n$ is the $k \times k$ matrix $G(S)$, or $G(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_k)$, of **inner products**

$$G(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_k)[i, j] := \langle \mathbf{x}_i, \mathbf{x}_j \rangle .$$

$\det G(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_k)$ is the **Gramian** of $\{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_k\}$.

If $\{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_k\}$ are columns of $X \in \mathbb{R}^{n \times k}$, then

$$\det G(\mathbf{x}_1, \dots, \mathbf{x}_k) = \det X^T X = \sum_{I \in Q_{k,n}} \det^2 X_{I^*} , \text{ (by C-B)}$$

- (a) The set S is **linearly dependent** iff $G(S) = 0$.
- (b) The **volume** of the **parallelepiped** generated by the vectors $\{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_k\}$ is

$$\text{vol } \square \{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_k\} = \sqrt{\det G(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_k)}$$

(in the subspace $\text{span } \{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_k\}$, not in \mathbb{R}^n if $k < n$)

Volume

The **volume** of a matrix $A \in \mathbb{R}_r^{m \times n}$, denoted $\text{vol } A$, is defined as 0 if $r = 0$, and otherwise

$$\text{vol } A := \sqrt{\sum_{(I,J) \in \mathcal{N}(A)} \det^2 A_{IJ}} . \quad (1)$$

Volume

The **volume** of a matrix $A \in \mathbb{R}_r^{m \times n}$, denoted $\text{vol } A$, is defined as 0 if $r = 0$, and otherwise

$$\text{vol } A := \sqrt{\sum_{(I,J) \in \mathcal{N}(A)} \det^2 A_{IJ}}. \quad (1)$$

- (a) If $C \in \mathbb{R}_n^{m \times n}$ (**full row rank**) then $\text{vol } C = \sqrt{\det C^T C}$.
- (b) If $R \in \mathbb{R}_m^{m \times n}$ (**full column rank**) then $\text{vol } R = \sqrt{\det RR^T}$.
- (c) If $A = CR$ is a **full rank factorization** then

$$\text{vol}^2(A) = \sum_{I \in \mathcal{I}(A)} \text{vol}^2(A_{I*}) \quad (2)$$

$$= \sum_{J \in \mathcal{J}(A)} \text{vol}^2(A_{*J}) \quad (3)$$

$$= \text{vol}^2(C) \text{vol}^2(R) \quad (4)$$

Proof. Each nonsingular A_{IJ} is $C_{I*}R_{*J}$, $I \in \mathcal{I}(A)$, $J \in \mathcal{J}(A)$.

Singular values

Theorem. The volume of $A \in \mathbb{R}_r^{m \times n}$ is the product of its singular values $\{\sigma_1, \dots, \sigma_r\}$

$$\text{vol } A = \prod_{\sigma_i \in \sigma(A)} \sigma_i \quad (1)$$

Proof. The SVD of A ,

$$A = U\Sigma V^* = \begin{bmatrix} U_1 & U_2 \end{bmatrix} \begin{bmatrix} \Sigma_{11} & O \\ O & O \end{bmatrix} \begin{bmatrix} V_1^* \\ V_2^* \end{bmatrix}, \quad \Sigma_{11} \in \mathbb{R}_r^{r \times r},$$

is a full rank factorization

$$A = U_1 \Sigma_{11} V_1^*, \quad \text{with } U_1^* U_1 = V_1^* V_1 = I_r.$$

$$\therefore \text{vol } A = \text{vol } \Sigma_{11} = \prod_{\sigma_i \in \sigma(A)} \sigma_i$$

Why "volume"?

Let $\{\mathbf{v}_j : j \in \overline{1, r}\}$ be an o.n. basis of $R(A^*A) = R(A^*)$ such that

$$A^*A\mathbf{v}_j = \sigma_j^2\mathbf{v}_j, \quad j \in \overline{1, r},$$

Then an o.n. basis $\{\mathbf{u}_j : j \in \overline{1, r}\}$ of $R(A)$ is given by

$$A\mathbf{v}_j = \sigma_j\mathbf{u}_j$$

showing that A maps the **unit cube** $\square\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_r\}$ into the **cube** $\square\{\sigma_1\mathbf{u}_1, \sigma_2\mathbf{u}_2, \dots, \sigma_r\mathbf{u}_r\}$ of **volume** $\sigma_1\sigma_2\cdots\sigma_r$. But the singular values are **orthogonally invariant**, and therefore **every** unit cube in $\mathbb{R}(A^*)$ is mapped into a cube of volume $\sigma_1\sigma_2\cdots\sigma_r$.

Given $A \in \mathbb{R}^{m \times n}$, the **volume** of A is an **intrinsic property** of the **linear transformation** in $\mathcal{L}(\mathbb{R}^n, \mathbb{R}^m)$ represented by A .

LSS

Let $A \in \mathbb{R}_n^{m \times n}$, $\mathbf{b} \in \mathbb{R}^m$. Then the LSS of $A\mathbf{x} = \mathbf{b}$ is

$$A^\dagger \mathbf{b} = \sum_{I \in \mathcal{I}(A)} \lambda_I \widehat{A_{I^*}^{-1} \mathbf{b}_I}, \quad \lambda_I = \frac{\det^2 A_{I^*}}{\sum_{K \in \mathcal{I}(A)} \det^2 A_{K^*}}$$

Ex. $n = 1$.

$$\begin{bmatrix} a_1 \\ \vdots \\ a_m \end{bmatrix} x = \begin{bmatrix} b_1 \\ \vdots \\ b_m \end{bmatrix} .$$

The LSS is

$$x = \frac{\sum_{i=1}^m a_i b_i}{\sum_{k=1}^m a_k^2} = \sum_{i=1}^m \left(\frac{a_i^2}{\sum_{k=1}^m a_k^2} \right) a_i^{-1} b_i = \sum_{i=1}^m \lambda_i a_i^{-1} b_i$$

This trivial example proves the general case.

MNLSS in 2 stages

For $A \in \mathbb{R}_r^{m \times n}$, $\mathbf{b} \in \mathbb{R}^m$, the MNLSS of the linear equation

$$A\mathbf{x} = \mathbf{b} ,$$

is the solution of a **two-stage minimization problem**:

Stage 1: minimize $\|A\mathbf{x} - \mathbf{b}\|$ (1)

Stage 2: minimize $\|\mathbf{x}\|$ among all solutions of Stage 1 (2)

MNLSS in 2 stages

For $A \in \mathbb{R}_r^{m \times n}$, $\mathbf{b} \in \mathbb{R}^m$, the MNLSS of the linear equation

$$A\mathbf{x} = \mathbf{b} ,$$

is the solution of a **two-stage minimization problem**:

Stage 1: minimize $\|A\mathbf{x} - \mathbf{b}\|$ (1)

Stage 2: minimize $\|\mathbf{x}\|$ among all solutions of Stage 1 (2)

Let

$$A = CR$$

be a **full rank factorization**. Then:

Stage 1: minimize $\|C\mathbf{y} - \mathbf{b}\|$ (1a)

Stage 2: minimize $\|\mathbf{x}\|$ among all solutions of $R\mathbf{x} = \mathbf{y}$ (2a)

The problem (1a), unlike (1), has a **unique solution** $\mathbf{y} = C^\dagger \mathbf{b}$.

Full rank systems

Lemma.

(a) Let $C \in \mathbb{R}_r^{m \times r}$, $\mathbf{b} \in \mathbb{R}^m$. Then the LSS of

$$C\mathbf{y} = \mathbf{b}, \quad (1)$$

is

$$\mathbf{y} = \sum_{I \in \mathcal{I}(C)} \mu_{I^*} C_{I^*}^{-1} \mathbf{b}_I, \quad \mu_{I^*} = \frac{\text{vol}^2 C_{I^*}}{\text{vol}^2 C}. \quad (2)$$

Full rank systems

Lemma.

(a) Let $C \in \mathbb{R}_r^{m \times r}$, $\mathbf{b} \in \mathbb{R}^m$. Then the LSS of

$$C\mathbf{y} = \mathbf{b}, \quad (1)$$

is

$$\mathbf{y} = \sum_{I \in \mathcal{I}(C)} \mu_{I*} C_{I*}^{-1} \mathbf{b}_I, \quad \mu_{I*} = \frac{\text{vol}^2 C_{I*}}{\text{vol}^2 C}. \quad (2)$$

(b) Let $R \in \mathbb{R}_r^{r \times n}$, $\mathbf{y} \in \mathbb{R}^r$. Then the MNS of

$$R\mathbf{x} = \mathbf{y}, \quad (3)$$

is

$$\mathbf{x} = \sum_{J \in \mathcal{J}(R)} \nu_{*J} \widehat{R_{*J}^{-1}} \mathbf{y}, \quad \nu_{*J} = \frac{\text{vol}^2 R_{*J}}{\text{vol}^2 R}. \quad (4)$$

MNLSS as convex combination

Theorem (Berg). Let $A \in \mathbb{R}_r^{m \times n}$, $\mathbf{b} \in \mathbb{R}^m$. Then the MNLSS of

$$A\mathbf{x} = \mathbf{b}, \quad (1)$$

is the convex combination

$$\mathbf{x} = \sum_{(I,J) \in \mathcal{N}(A)} \lambda_{IJ} \widehat{A_{IJ}^{-1} \mathbf{b}_I}, \quad (2)$$

with weights given by

$$\lambda_{IJ} = \frac{\det^2 A_{IJ}}{\sum_{(K,L) \in \mathcal{N}(A)} \det^2 A_{KL}}, \quad (I, J) \in \mathcal{N}(A). \quad (3)$$

Proof. Let $A = CR$ be a full rank factorization, and use previous lemma and the fact

$$A_{IJ} = C_{I*} R_{*J}, \quad \forall I \in \mathcal{I}(A), J \in \mathcal{J}(A).$$

Weighted LS

Consider next a **weighted** least squares problem

$$\min \|D^{1/2}(A\mathbf{x} - \mathbf{b})\|, \quad D = \text{diag}(d_i), \quad d_i > 0. \quad (1)$$

Theorem (Ben-Tal and Teboulle). The solutions of (1), i.e., the least-squares solutions of $D^{1/2}A\mathbf{x} = D^{1/2}\mathbf{b}$, satisfy the normal equation, $A^T D A \mathbf{x} = A^T D \mathbf{b}$. The MN (weighted) **LSS** is

$$\mathbf{x}(D) = \sum_{(I,J) \in \mathcal{N}(A)} \lambda_{IJ}(D) \widehat{A_{IJ}^{-1} \mathbf{b}_I}, \quad (2)$$

with weights

$$\lambda_{IJ}(D) = \frac{(\prod_{i \in I} d_i) \det^2 A_{IJ}}{\sum_{(I,J) \in \mathcal{N}(A)} (\prod_{i \in K} d_i) \det^2 A_{KL}}. \quad (3)$$

Note: Only the weights λ_{IJ} depend on D .

Bordered matrices

Let $A \in \mathbb{R}_r^{m \times n}$, and let

$$U \in \mathbb{R}^{m \times (m-r)}, \text{ cols } U = \text{o.n. basis } N(A^T),$$

$$V \in \mathbb{R}^{n \times (n-r)}, \text{ cols } V = \text{o.n. basis } N(A).$$

Consider the **bordered matrix**,

$$\mathbf{B}(A) := \begin{bmatrix} A & U \\ V^T & O \end{bmatrix}, \quad (1)$$

and if $m = n$, the **complemented matrix**,

$$\mathbf{C}(A) := A + UV^T. \quad (2)$$

Then

$$\text{vol } A = |\det \mathbf{B}(A)|, \quad (3)$$

$$= |\det \mathbf{C}(A)|. \quad (4)$$

The change-of-variables formula in integration

Theorem (**Jacobi** general n , **Euler** $n = 2$, **Lagrange** $n = 3$)

$$\int_{\mathcal{V}} f(\mathbf{v}) d\mathbf{v} = \int_{\mathcal{U}} (f \circ \phi)(\mathbf{u}) |\det J_{\phi}(\mathbf{u})| d\mathbf{u} \quad (1)$$

The change-of-variables formula in integration

Theorem (**Jacobi** general n , **Euler** $n = 2$, **Lagrange** $n = 3$)

$$\int_{\mathcal{V}} f(\mathbf{v}) d\mathbf{v} = \int_{\mathcal{U}} (f \circ \phi)(\mathbf{u}) |\det J_{\phi}(\mathbf{u})| d\mathbf{u} \quad (1)$$

\mathcal{U}, \mathcal{V} sets in \mathbb{R}^n ,

$\phi : \mathcal{U} \rightarrow \mathcal{V}$ a sufficiently well-behaved function,

f is integrable on \mathcal{V} ,

$d\mathbf{x}$ denotes the **volume element** $|dx_1 \wedge dx_2 \wedge \cdots \wedge dx_n|$,

J_{ϕ} is the **Jacobian matrix** (or **Jacobian**)

$$J_{\phi} := \left(\frac{\partial \phi_i}{\partial u_j} \right), \quad \text{also denoted} \quad \frac{\partial(v_1, v_2, \cdots, v_n)}{\partial(u_1, u_2, \cdots, u_n)}, \quad (2)$$

representing the **derivative** of ϕ .

An **advantage** of (1) is that integration on \mathcal{V} is translated to (perhaps **simpler**) integration on \mathcal{U} .

The change-of-variable formula (cont'd)

$$\int_{\mathcal{V}} f(\mathbf{v}) d\mathbf{v} = \int_{\mathcal{U}} (f \circ \phi)(\mathbf{u}) |\det J_{\phi}(\mathbf{u})| d\mathbf{u} \quad (1)$$

If $\mathcal{U} \subset \mathbb{R}^n$ and $\mathcal{V} \subset \mathbb{R}^m$ with $n > m$, formula (1) cannot be applied.

The change-of-variable formula (cont'd)

$$\int_{\mathcal{V}} f(\mathbf{v}) d\mathbf{v} = \int_{\mathcal{U}} (f \circ \phi)(\mathbf{u}) |\det J_{\phi}(\mathbf{u})| d\mathbf{u} \quad (1)$$

If $\mathcal{U} \subset \mathbb{R}^n$ and $\mathcal{V} \subset \mathbb{R}^m$ with $n > m$, formula (1) cannot be applied.

However, if the Jacobian J_{ϕ} is of **full column rank** throughout \mathcal{U} , we can replace $|\det J_{\phi}|$ by the **volume** $\text{vol } J_{\phi}$ of the **Jacobian** to get

$$\int_{\mathcal{V}} f(\mathbf{v}) d\mathbf{v} = \int_{\mathcal{U}} (f \circ \phi)(\mathbf{u}) \text{vol } J_{\phi}(\mathbf{u}) d\mathbf{u} . \quad (2)$$

Here,

$$\text{vol } J_{\phi} = \sqrt{\det (J_{\phi}^T J_{\phi})} \quad (3)$$

since J_{ϕ} is assumed of full rank, and (2) reduces to (1) if $m = n$.

Example in surface integration

Let \mathcal{S} be a subset of a **surface** in \mathbb{R}^3 represented by

$$z = g(x, y) , \quad (1)$$

and let $f(x, y, z)$ be a function integrable on \mathcal{S} . Let \mathcal{A} be the **projection** of \mathcal{S} on the xy -plane. Then

$$\mathcal{S} = \phi(\mathcal{A}) , \quad \text{or} \quad \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} x \\ y \\ g(x, y) \end{bmatrix} = \phi \begin{bmatrix} x \\ y \end{bmatrix} , \quad \begin{bmatrix} x \\ y \end{bmatrix} \in \mathcal{A} . \quad (2)$$

The Jacobi matrix of ϕ is the 3×2 matrix

$$J_\phi(x, y) = \frac{\partial(x, y, z)}{\partial(x, y)} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ g_x & g_y \end{bmatrix} , \quad (3)$$

where $g_x = \frac{\partial g}{\partial x}$, $g_y = \frac{\partial g}{\partial y}$.

Example (cont'd)

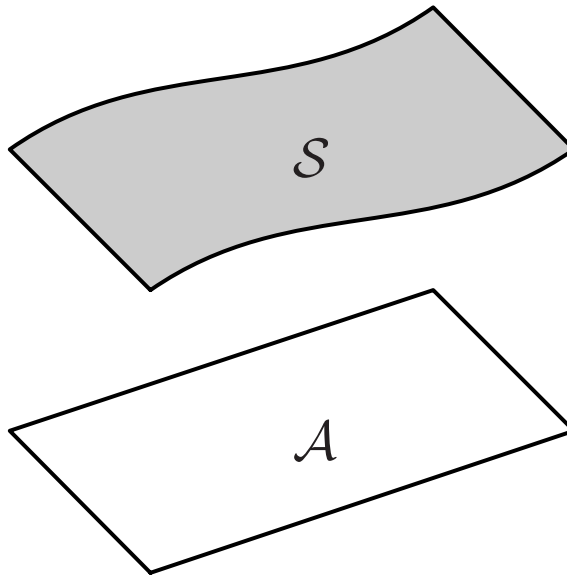
$$\text{vol } J_\phi = \text{vol} \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ g_x & g_y \end{bmatrix} = \sqrt{1 + g_x^2 + g_y^2}. \quad (4)$$

$$\therefore \int_{\mathcal{S}} f(x, y, z) \, ds = \int_{\mathcal{A}} f(x, y, g(x, y)) \sqrt{1 + g_x^2 + g_y^2} \, dx \, dy.$$

Example (cont'd)

$$\text{vol } J_\phi = \text{vol} \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ g_x & g_y \end{bmatrix} = \sqrt{1 + g_x^2 + g_y^2}. \quad (4)$$

$$\therefore \int_S f(x, y, z) \, ds = \int_A f(x, y, g(x, y)) \sqrt{1 + g_x^2 + g_y^2} \, dx \, dy.$$



Cylindrical coordinates

Let \mathcal{S} be a surface in \mathbb{R}^3 , represented by

$$z = z(r, \theta)$$

where r, θ are **cylindrical coordinates**,

$$x = r \cos \theta$$

$$y = r \sin \theta$$

$$z = z(r, \theta)$$

Cylindrical coordinates

Let \mathcal{S} be a surface in \mathbb{R}^3 , represented by

$$z = z(r, \theta)$$

where r, θ are **cylindrical coordinates**,

$$x = r \cos \theta$$

$$y = r \sin \theta$$

$$z = z(r, \theta)$$

The **Jacobi matrix** is

$$J_\phi = \frac{\partial(x, y, z)}{\partial(r, \theta)} = \begin{bmatrix} \cos \theta & -r \sin \theta \\ \sin \theta & r \cos \theta \\ \frac{\partial z}{\partial r} & \frac{\partial z}{\partial \theta} \end{bmatrix}$$

and its **volume** is

$$\text{vol } J_\phi = \sqrt{r^2 + r^2 \left(\frac{\partial z}{\partial r}\right)^2 + \left(\frac{\partial z}{\partial \theta}\right)^2} = r \sqrt{1 + \left(\frac{\partial z}{\partial r}\right)^2 + \frac{1}{r^2} \left(\frac{\partial z}{\partial \theta}\right)^2}$$

Cylindrical coordinates (cont'd)

An integral over a domain $\mathcal{V} \subset \mathcal{S}$ is

$$\int_{\mathcal{V}} f(x, y, z) dV =$$
$$\int_{\mathcal{U}} f(r \cos \theta, r \sin \theta, z(r, \theta)) \text{vol} J_{\phi}(r, \theta) dr d\theta$$

Cylindrical coordinates (cont'd)

An integral over a domain $\mathcal{V} \subset \mathcal{S}$ is

$$\int_{\mathcal{V}} f(x, y, z) dV = \int_{\mathcal{U}} f(r \cos \theta, r \sin \theta, z(r, \theta)) \text{vol} J_{\phi}(r, \theta) dr d\theta$$

If \mathcal{S} is **symmetric** about the z -axis, then

$$\frac{\partial z}{\partial \theta} = 0, \text{ i.e. } z = z(r),$$

and

$$\int_{\mathcal{V}} f(x, y, z) dV = \int_{\mathcal{U}} f(r \cos \theta, r \sin \theta, z(r)) r \sqrt{1 + z'(r)^2} dr d\theta$$

Surface integral in \mathbb{R}^n

Let $\mathbf{x} = (x_i)$, and $\mathbf{x}[\overline{1, k}] := \{(x_i) : i \in \overline{1, k}\}$.

A **surface** \mathcal{S} in \mathbb{R}^n is given by

$$x_n := g(x_1, \dots, x_{n-1}) = g(\mathbf{x}[\overline{1, (n-1)}]) , \quad (1)$$

Let

$\mathcal{V} \subset \mathcal{S}$, f integrable on \mathcal{V} , and it is required to calculate

$$\int_{\mathcal{V}} f(\mathbf{x}) dS \quad (2)$$

\mathcal{U} the **projection** of \mathcal{V} on \mathbb{R}^{n-1} , the space of variables $\mathbf{x}[1:(n-1)]$,

$\phi : \mathcal{U} \rightarrow \mathcal{V}$, the **mapping** given by its components

$$\phi := (\phi_1, \phi_2, \dots, \phi_n) ,$$

$$\phi_i(\mathbf{x}[\overline{1, (n-1)}]) := x_i , \quad i = 1, \dots, n-1$$

$$\phi_n(\mathbf{x}[\overline{1, (n-1)}]) := g(\mathbf{x}[\overline{1, (n-1)}])$$

Surface integral in \mathbb{R}^n (cont'd)

The Jacobi matrix of ϕ is

$$J_\phi = \begin{bmatrix} 1 & 0 & \cdots & 0 & 0 \\ 0 & 1 & \cdots & 0 & 0 \\ 0 & 0 & \ddots & 0 & 0 \\ 0 & 0 & \cdots & 1 & 0 \\ 0 & 0 & \cdots & 0 & 1 \\ \frac{\partial g}{\partial x_1} & \frac{\partial g}{\partial x_2} & \cdots & \frac{\partial g}{\partial x_{n-2}} & \frac{\partial g}{\partial x_{n-1}} \end{bmatrix}$$

and its volume is

$$\text{vol } J_\phi = \sqrt{1 + \sum_{i=1}^{n-1} \left(\frac{\partial g}{\partial x_i} \right)^2}$$

$$\therefore \int_{\mathcal{V}} f(\mathbf{x}) dS = \int_{\mathcal{U}} f(\mathbf{x}[\overline{1, (n-1)}], g(\mathbf{x}[\overline{1, (n-1)}])) \text{vol } J_\phi dx_1 \cdots dx_{n-1}$$

Radon transform

Let $\mathcal{H}_{\xi,p}$ be a hyperplane in \mathbb{R}^n represented by

$$\mathcal{H}_{\xi,p} := \left\{ \mathbf{x} \in \mathbb{R}^n : \sum_{i=1}^n \xi_i x_i = p \right\} = \{ \mathbf{x} : \langle \xi, \mathbf{x} \rangle = p \} \quad (1)$$

where the **normal vector** $\xi = (\xi_1, \dots, \xi_n)$ of $\mathcal{H}_{\xi,p}$ has $\xi_n \neq 0$.

Then $\mathcal{H}_{\xi,p}$ is given as

$$\mathcal{H}_{\xi,p} = \phi(\mathbb{R}^{n-1}) \quad (2)$$

with

$$\begin{aligned} \phi_i(\mathbf{x}[\overline{1, (n-1)}]) &:= x_i, \quad i \in \overline{1, (n-1)} \\ \phi_n(\mathbf{x}[\overline{1, (n-1)}]) &:= \frac{p}{\xi_n} - \sum_{i=1}^{n-1} \frac{\xi_i}{\xi_n} x_i \end{aligned} \quad (3)$$

Radon transform (cont'd)

The volume of J_ϕ is here

$$\text{vol } J_\phi = \sqrt{1 + \sum_{i=1}^{n-1} \left(\frac{\xi_i}{\xi_n}\right)^2} = \frac{\|\boldsymbol{\xi}\|}{|\xi_n|} \quad (4)$$

The **Radon transform** $(\mathbf{R}f)(\boldsymbol{\xi}, p)$ of a function $f : \mathbb{R}^n \rightarrow \mathbb{R}$ is its integral over the hyperplane $\mathcal{H}_{\boldsymbol{\xi}, p}$,

$$(\mathbf{R}f)(\boldsymbol{\xi}, p) := \int_{\{\mathbf{x}: \langle \boldsymbol{\xi}, \mathbf{x} \rangle = p\}} f(\mathbf{x}) \, d\mathbf{x} . \quad (5)$$

The Radon transform can be computed as an integral in \mathbb{R}^{n-1}

$$(\mathbf{R}f)(\boldsymbol{\xi}, p) = \frac{\|\boldsymbol{\xi}\|}{|\xi_n|} \int_{\mathbb{R}^{n-1}} f \left(\mathbf{x}[1, (n-1)], \frac{p}{\xi_n} - \sum_{i=1}^{n-1} \frac{\xi_i}{\xi_n} x_i \right) dx_1 \cdots dx_{n-1} \quad (6)$$

Integrals over \mathbb{R}^n

Consider an integral over \mathbb{R}^n ,

$$\int_{\mathbb{R}^n} f(\mathbf{x}) d\mathbf{x} = \int_{\mathbb{R}^n} f(x_1, x_2, \dots, x_n) dx_1 \cdots dx_n \quad (1)$$

Since \mathbb{R}^n is a **union** of (parallel) hyperplanes,

$$\mathbb{R}^n = \bigcup_{p=-\infty}^{\infty} \{\mathbf{x} : \langle \boldsymbol{\xi}, \mathbf{x} \rangle = p\}, \quad \text{where } \boldsymbol{\xi} \neq \mathbf{0}, \quad (2)$$

Compute (1) iteratively: an integral over \mathbb{R}^{n-1} (**Radon transform**), followed by an integral on \mathbb{R} ,

$$\int_{\mathbb{R}^n} f(\mathbf{x}) d\mathbf{x} = \int_{-\infty}^{\infty} \frac{dp}{\|\boldsymbol{\xi}\|} (\mathbf{R}f)(\boldsymbol{\xi}, p) \quad (3)$$

Integrals over \mathbb{R}^n (cont'd)

$$\int_{\mathbb{R}^n} f(\mathbf{x}) d\mathbf{x} = \int_{-\infty}^{\infty} \frac{dp}{\|\boldsymbol{\xi}\|} (\mathbf{R}f)(\boldsymbol{\xi}, p) \quad (3)$$

Here $dp/\|\boldsymbol{\xi}\|$ is the **differential of the distance** along $\boldsymbol{\xi}$ (i.e. dp times the distance between the parallel hyperplanes $\mathcal{H}_{\boldsymbol{\xi},p}$ and $\mathcal{H}_{\boldsymbol{\xi},p+1}$).

Then by the result for the Radon transform,

$$\int_{\mathbb{R}^n} f(\mathbf{x}) d\mathbf{x} = \frac{1}{|\xi_n|} \int_{-\infty}^{\infty} \left(\int_{\mathbb{R}^{n-1}} f \left(\mathbf{x}_{[1, (n-1)]}, \frac{p}{\xi_n} - \sum_{i=1}^{n-1} \frac{\xi_i}{\xi_n} x_i \right) dx_1 \cdots dx_{n-1} \right) dp. \quad (4)$$

Application in Probability

Let $\mathbf{X} = (\mathbf{X}_1, \dots, \mathbf{X}_n)$ have **joint density** $f_{\mathbf{X}}(x_1, \dots, x_n)$ and let

$$y = h(x_1, \dots, x_n) \quad (1)$$

where $h : \mathbb{R}^n \rightarrow \mathbb{R}$ is sufficiently well-behaved, in particular

$$\frac{\partial h}{\partial x_n} \neq 0,$$

and (1) can be solved for x_n ,

$$x_n = h^{-1}(y|x_1, \dots, x_{n-1}) \quad (2)$$

with x_1, \dots, x_{n-1} as parameters.

It is required to find the **density** $f_{\mathbf{Y}}(y)$ of \mathbf{Y} .

Method 1

Change variables from $\{x_1, \dots, x_n\}$ to $\{x_1, \dots, x_{n-1}, y\}$, and use the fact

$$\det \left(\frac{\partial(x_1, \dots, x_n)}{\partial(x_1, \dots, x_{n-1}, y)} \right) = \frac{\partial h^{-1}}{\partial y} \quad (3)$$

to write the density of $\mathbf{Y} = h(\mathbf{X}_1, \dots, \mathbf{X}_n)$ as

$$f_{\mathbf{Y}}(y) = \int_{\mathbb{R}^{n-1}} f_{\mathbf{X}}(x_1, \dots, x_{n-1}, h^{-1}(y|x_1, \dots, x_{n-1})) \left| \frac{\partial h^{-1}}{\partial y} \right| dx_1 \cdots dx_{n-1} \quad (4)$$

Method 2

Let $\mathcal{V}(y)$ be the surface given by (1), represented as

$$\begin{pmatrix} x_1 \\ \vdots \\ x_{n-1} \\ x_n \end{pmatrix} = \begin{pmatrix} x_1 \\ \vdots \\ x_{n-1} \\ h^{-1}(y|x_1, \dots, x_{n-1}) \end{pmatrix} = \phi \begin{pmatrix} x_1 \\ \vdots \\ x_{n-1} \end{pmatrix} \quad (5)$$

Then the surface integral of $f_{\mathbf{X}}$ over $\mathcal{V}(y)$ is

$$\int_{\mathcal{V}(y)} f_{\mathbf{X}} = \quad (6)$$
$$\int_{\mathbb{R}^{n-1}} f_{\mathbf{X}}(x_1, \dots, x_{n-1}, h^{-1}(y|x_1, \dots, x_{n-1}))$$
$$\sqrt{1 + \sum_{i=1}^{n-1} \left(\frac{\partial h^{-1}}{\partial x_i} \right)^2} dx_1 \cdots dx_{n-1}$$

The density of $Y = h(\mathbf{X}_1, \dots, \mathbf{X}_n)$

Theorem. If the ratio

$$\frac{\frac{\partial h^{-1}}{\partial y}}{\sqrt{1 + \sum_{i=1}^{n-1} \left(\frac{\partial h^{-1}}{\partial x_i}\right)^2}}$$
 does not depend on x_1, \dots, x_{n-1} , (\mathbf{X})

then

$$f_{\mathbf{Y}}(y) = \frac{\left| \frac{\partial h^{-1}}{\partial y} \right|}{\sqrt{1 + \sum_{i=1}^{n-1} \left(\frac{\partial h^{-1}}{\partial x_i}\right)^2}} \int_{\mathcal{V}(y)} f_{\mathbf{X}}$$

Proof. Compare (4) and (6).

Hyperplanes

Condition **(X)** holds for hyperplanes. Let

$$y = h(x_1, \dots, x_n) := \sum_{i=1}^n \xi_i x_i$$

where $\boldsymbol{\xi} = (\xi_1, \dots, \xi_n)$ is a given vector with $\xi_n \neq 0$. Then (2) is

$$x_n = h^{-1}(y | x_1, \dots, x_{n-1}) := \frac{y}{\xi_n} - \sum_{i=1}^{n-1} \frac{\xi_i}{\xi_n} x_i$$

$$\text{and } \frac{\partial h^{-1}}{\partial y} = \frac{1}{\xi_n},$$

$$\sqrt{1 + \sum_{i=1}^{n-1} \left(\frac{\partial h^{-1}}{\partial x_i} \right)^2} = \sqrt{1 + \sum_{i=1}^{n-1} \left(\frac{\xi_i}{\xi_n} \right)^2} = \frac{\|\boldsymbol{\xi}\|}{|\xi_n|}$$

Hyperplanes (cont'd)

Corollary. Let $\mathbf{X} = (\mathbf{X}_1, \mathbf{X}_2, \dots, \mathbf{X}_n)$ be random variables with joint density $f_{\mathbf{X}}(x_1, x_2, \dots, x_n)$, and let $\mathbf{0} \neq \boldsymbol{\xi} \in \mathbb{R}^n$.

The random variable

$$\mathbf{Y} := \sum_{i=1}^n \xi_i \mathbf{X}_i$$

has the density

$$f_{\mathbf{Y}}(y) = \frac{(\mathbf{R}f_{\mathbf{X}})(\boldsymbol{\xi}, y)}{\|\boldsymbol{\xi}\|}.$$

where $(\mathbf{R}f_{\mathbf{X}})(\boldsymbol{\xi}, y)$ is the **Radon transform** of $f_{\mathbf{X}}$,

$$\begin{aligned} & (\mathbf{R}f_{\mathbf{X}})(\boldsymbol{\xi}, y) = \\ & \frac{\|\boldsymbol{\xi}\|}{|\xi_n|} \int_{\mathbb{R}^{n-1}} f_{\mathbf{X}} \left(x_1, \dots, x_{n-1}, \frac{y}{\xi_n} - \sum_{i=1}^{n-1} \frac{\xi_i}{\xi_n} x_i \right) dx_1 dx_2 \cdots dx_{n-1}. \end{aligned}$$

Spheres

Condition **(X)** holds for spheres. Let

$$y = h(x_1, \dots, x_n) := \sum_{i=1}^n x_i^2$$

which has two solutions for x_n , representing the upper and lower hemispheres,

$$x_n = h^{-1}(y | x_1, \dots, x_{n-1}) := \pm \sqrt{y - \sum_{i=1}^{n-1} x_i^2}$$

$$\text{with } \frac{\partial h^{-1}}{\partial y} = \pm \frac{1}{2 \sqrt{y - \sum_{i=1}^{n-1} x_i^2}},$$

$$\sqrt{1 + \sum_{i=1}^{n-1} \left(\frac{\partial h^{-1}}{\partial x_i} \right)^2} = \frac{\sqrt{y}}{\sqrt{y - \sum_{i=1}^{n-1} x_i^2}}$$

Spheres (cont'd)

Corollary. Let $\mathbf{X} = (\mathbf{X}_1, \dots, \mathbf{X}_n)$ have joint density $f_{\mathbf{X}}(x_1, \dots, x_n)$. The density of

$$Y = \sum_{i=1}^n \mathbf{X}_i^2$$

is

$$f_{\mathbf{Y}}(y) = \frac{1}{2\sqrt{y}} \int_{\mathcal{S}_n(\sqrt{y})} f_{\mathbf{X}}$$

where the integral is over the sphere $\mathcal{S}_n(\sqrt{y})$ of radius \sqrt{y} ,

Spheres (cont'd)

computed as an integral over the ball $\mathcal{B}_{n-1}(\sqrt{y})$,

$$\int_{\mathcal{S}_n(\sqrt{y})} f_{\mathbf{X}} = \int_{\mathcal{B}_{n-1}(\sqrt{y})} \left[f_{\mathbf{X}} \left(x_1, \dots, x_{n-1}, \sqrt{y - \sum_{i=1}^{n-1} x_i^2} \right) + f_{\mathbf{X}} \left(x_1, \dots, x_{n-1}, -\sqrt{y - \sum_{i=1}^{n-1} x_i^2} \right) \right] \frac{\sqrt{y} \, dx_1 \cdots dx_{n-1}}{\sqrt{y - \sum_{i=1}^{n-1} x_i^2}}$$