

3 Applications

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3a Constrained optimization

One of the most brilliant and well-known achievements of differential calculus is the collection of recipes it provides for finding the extrema of functions. . . . Frequently a situation that is more complicated and from the practical point of view even more interesting arises, in which one seeks an extremum of a function under certain constraints . . . ¹

Let $Z \subset \mathbb{R}^n$ be a set, $f : Z \rightarrow \mathbb{R}$ a function, and $x_0 \in Z$ a point. We say that x_0 is a local maximum point of f on Z , if $f(x) \leq f(x_0)$ for all $x \in Z$ near x_0 . (A local minimum point is defined similarly.)

In particular, if $Z = g^{-1}(\{0\}) = \{x : g(x) = 0\}$ for a given $g : \mathbb{R}^n \rightarrow \mathbb{R}^m$, a local maximum point of f on Z is called a local maximum point of f subject to the constraint $g(\cdot) = 0$. That is, subject to $g_1(\cdot) = \cdots = g_m(\cdot) = 0$ where $(g_1(x), \dots, g_m(x)) = g(x)$. “Extremum” means either maximum or minimum, of course.

3a1 Theorem. Assume that $x_0 \in \mathbb{R}^n$, $1 \leq m \leq n-1$, functions $f, g_1, \dots, g_m : \mathbb{R}^n \rightarrow \mathbb{R}$ are continuously differentiable near x_0 , $g_1(x_0) = \cdots = g_m(x_0) = 0$, and the vectors $\nabla g_1(x_0), \dots, \nabla g_m(x_0)$ are linearly independent. If x_0 is a local constrained extremum point of f subject to $g_1(\cdot) = \cdots = g_m(\cdot) = 0$, then there exist $\lambda_1, \dots, \lambda_m \in \mathbb{R}$ such that

$$\nabla f(x_0) = \lambda_1 \nabla g_1(x_0) + \cdots + \lambda_m \nabla g_m(x_0).$$

¹Quoted from: Zorich, Sect. 8.7.3a, p. 527.

The numbers $\lambda_1, \dots, \lambda_m$ are called *Lagrange multipliers*.

A physicist could say: in equilibrium, the driving force is neutralized by constraints reaction forces.

In practice, seeking local constrained extrema of f on $Z = g^{-1}(\{0\})$ one solves (that is, finds *all* solutions of) a system of $m + n$ equations

$$\begin{aligned} g_1(x) = \dots = g_m(x) = 0, & \quad (m \text{ equations}) \\ \nabla f(x) = \lambda_1 \nabla g_1(x) + \dots + \lambda_m \nabla g_m(x) & \quad (n \text{ equations}) \end{aligned}$$

for $m + n$ variables

$$\begin{aligned} \lambda_1, \dots, \lambda_m, & \quad (m \text{ variables}) \\ x. & \quad (n \text{ variables}) \end{aligned}$$

For each solution $(\lambda_1, \dots, \lambda_m, x)$ one ignores $\lambda_1, \dots, \lambda_m$ and checks $f(x)$.¹

In addition, one checks $f(x)$ for all points x that violate the conditions of 3a1; that is, $\nabla g_1(x), \dots, \nabla g_m(x)$ are linearly dependent, or f, g_1, \dots, g_m fail to be continuously differentiable near x .

If the set Z is not compact, one checks *all* relevant limits of f .

If all that is feasible (which is not guaranteed!), one finally obtains the infimum and supremum of f on Z .

More formally: $\sup_{x \in Z} f(x) = \lim_k f(x_k) \in (-\infty, +\infty]$ for some $x_1, x_2, \dots \in Z$. Choosing a subsequence we ensure either $x_k \rightarrow x$ for some $x \in \overline{Z}$ or $|x_k| \rightarrow \infty$. In the case $x \in Z$ the point x must violate conditions of 3a1. That is enough if Z is compact. Otherwise, if Z is bounded and not closed, the case $x \in \overline{Z} \setminus Z$ must be examined. And if Z is unbounded, the case $|x_k| \rightarrow \infty$ must be examined.

In order to prove Th. 3a1 we first generalize Th. 2c3 as follows (recall 2a9).

3a2 Theorem. Let $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$ be continuously differentiable near 0, $f(0) = 0$, and $(Df)_0 = A : \mathbb{R}^n \rightarrow \mathbb{R}^m$ be onto. Then f is open at 0.

Proof. We take an m -dimensional subspace $E \subset \mathbb{R}^n$ such that $A|_E$ is an invertible mapping from E onto \mathbb{R}^m (this is possible, as explained in Sect. 2a, Item “linear algebra”). Then $(D(f|_E))_0 = A|_E$ is invertible; by Th. 2b1,² $f|_E$ is a local diffeomorphism, and therefore,³ open at 0. It follows that f is open at 0. \square

¹Being ignored in this framework, $(\lambda_1, \dots, \lambda_m)$ are of interest in another framework, see Sect. 3e.

²Choosing a basis in E we turn it to a copy of \mathbb{R}^m . Or, alternatively, E may be chosen to be spanned by some m out of the n standard basis vectors of \mathbb{R}^n .

³Use 2a7(a), as in the proof of 2c3.

Proof of Theorem 3a1. WLOG, the extremum is maximum, $x_0 = 0$ and $f(0) = 0$. Assume the contrary: $\nabla f(0)$ is not a linear combination of $\nabla g_1(0), \dots, \nabla g_m(0)$. Then vectors $\nabla g_1(0), \dots, \nabla g_m(0), \nabla f(0)$ are linearly independent. These vectors being the rows of $(D\varphi)_0$, where $\varphi(x) = (g_1(x), \dots, g_m(x), f(x))$, we see that $(D\varphi)_0 : \mathbb{R}^n \rightarrow \mathbb{R}^{m+1}$ is onto.¹ By Th. 3a2, φ is open at 0.

We take a neighborhood $U \subset \mathbb{R}^n$ of 0 such that $f(x) \leq f(x_0)$ for all $x \in U \cap Z$ (where $Z = g^{-1}(\{0\})$), note that $\varphi(U)$ is a neighborhood of 0 in \mathbb{R}^{m+1} , and therefore $\varphi(U)$ contains $(0, \dots, 0, \varepsilon)$ for $\varepsilon > 0$ small enough. That is, $\varphi(x) = (0, \dots, 0, \varepsilon)$ for some $x \in U$. Then $x \in Z$ and $f(x) > f(0)$, which is a contradiction. \square

Theorem 3a1, formulated in terms of gradients, involves a Euclidean metric on \mathbb{R}^n . However, it is easy to reformulate it for vector spaces (with no given metric), to be invariant under arbitrary change of basis (not just orthonormal), as follows.

Assume that V is an n -dimensional vector space, $x_0 \in V$, $1 \leq m \leq n-1$, functions $f, g_1, \dots, g_m : V \rightarrow \mathbb{R}$ are continuously differentiable near x_0 , $g_1(x_0) = \dots = g_m(x_0) = 0$, and the linear functions $(Dg_1)_{x_0}, \dots, (Dg_m)_{x_0} : V \rightarrow \mathbb{R}$ are linearly independent. If x_0 is a local constrained extremum point of f subject to $g_1(\cdot) = \dots = g_m(\cdot) = 0$, then there exist $\lambda_1, \dots, \lambda_m \in \mathbb{R}$ such that

$$(Df)_{x_0} = \lambda_1(Dg_1)_{x_0} + \dots + \lambda_m(Dg_m)_{x_0}.$$

3b Example: arithmetic, geometric, harmonic, and more general means

Here is an isoperimetric inequality for triangles Δ on the plane:

$$\text{area}(\Delta) \leq \frac{1}{12\sqrt{3}} (\text{perimeter}(\Delta))^2,$$

and equality is attained for equilateral triangles and only for them. In other words, among all triangles with the given perimeter, the equilateral one has the largest area.²

¹Recall Sect. 2a, Item “linear algebra”.

²Generally, $\text{area}(G) \leq \frac{1}{4\pi} (\text{perimeter}(G))^2$ for any G on the plane, and equality is attained for disks only. This is a famous deep fact. But I do not give an exact formulation (nor a proof, of course).

The proof is based on Heron's formula for the area A of a triangle whose side lengths are x, y, z (and perimeter $L = x + y + z$):

$$A^2 = \frac{L}{2} \left(\frac{L}{2} - x \right) \left(\frac{L}{2} - y \right) \left(\frac{L}{2} - z \right).$$

The sum of the three positive¹ numbers $\frac{L}{2} - x, \frac{L}{2} - y, \frac{L}{2} - z$ is fixed (equal to $\frac{3L}{2} - L = \frac{L}{2}$); their product is claimed to be maximal when these numbers are equal (to $L/6$), and then $A^2 = \frac{L}{2} \left(\frac{L}{6} \right)^3 = \frac{L^4}{2^4 \cdot 3^3}$; $A = \frac{L^2}{2^2 \cdot 3\sqrt{3}}$.

More generally, $\max\{x_1 \dots x_n : x_1, \dots, x_n \geq 0, x_1 + \dots + x_n = c\}$ is reached for $x_1 = \dots = x_n = c/n$ and is equal to $(c/n)^n$. Equivalently, $\max\{(x_1 \dots x_n)^{1/n} : x_1, \dots, x_n \geq 0, (x_1 + \dots + x_n)/n = c\}$ is reached for $x_1 = \dots = x_n = c$ and is equal to c , which is the well-known inequality for geometric mean and arithmetic mean,

$$(3b1) \quad (x_1 \dots x_n)^{1/n} \leq \frac{1}{n}(x_1 + \dots + x_n) \quad \text{for } n = 1, 2, \dots \text{ and } x_1, \dots, x_n \geq 0.$$

It follows easily from concavity of the logarithm: the set $A = \{(x, y) : x \in (0, \infty), y \leq \ln x\}$ is convex, therefore the convex combination $(\frac{1}{n}(x_1 + \dots + x_n), \frac{1}{n}(\ln x_1 + \dots + \ln x_n))$ of points $(x_1, \ln x_1), \dots, (x_n, \ln x_n) \in A$ belongs to A , which gives (3b1). And still, it is worth to exercise Lagrange multipliers.

3b2 Exercise. Prove (3b1) via Lagrange multipliers.

By the way, the harmonic mean h defined by $\frac{1}{h} = \frac{1}{n} \left(\frac{1}{x_1} + \dots + \frac{1}{x_n} \right)$ satisfies $h \leq (x_1 \dots x_n)^{1/n}$; just apply (3b1) to $\frac{1}{x_1}, \dots, \frac{1}{x_n}$.

More generally, the Hölder mean (called also power mean) with exponent $p \in (-\infty, 0) \cup (0, \infty)$ is

$$M_p(x_1, \dots, x_n) = \left(\frac{x_1^p + \dots + x_n^p}{n} \right)^{1/p} \quad \text{for } x_1, \dots, x_n > 0.$$

In particular, M_1 is the arithmetic mean and M_{-1} is the harmonic mean. For $p \rightarrow 0$ L'Hôpital's rule gives

$$\begin{aligned} \ln \lim_{p \rightarrow 0} M_p(x_1, \dots, x_n) &= \lim_{p \rightarrow 0} \frac{1}{p} \ln \frac{x_1^p + \dots + x_n^p}{n} = \\ &= \lim_{p \rightarrow 0} \frac{x_1^p \ln x_1 + \dots + x_n^p \ln x_n}{x_1^p + \dots + x_n^p} = \frac{\ln x_1 + \dots + \ln x_n}{n} = \ln(x_1 \dots x_n)^{1/n}; \end{aligned}$$

¹ $\frac{L}{2} - x = \frac{x+y+z}{2} - x = \frac{y+z-x}{2} > 0$ by the triangle inequality.

accordingly, one defines

$$M_0(x_1, \dots, x_n) = (x_1 \dots x_n)^{1/n},$$

and observes that $M_{-1}(x_1, \dots, x_n) \leq M_0(x_1, \dots, x_n) \leq M_1(x_1, \dots, x_n)$. For $p \rightarrow +\infty$ we have

$$\frac{1}{n} \max(x_1^p, \dots, x_n^p) \leq \frac{x_1^p + \dots + x_n^p}{n} \leq \max(x_1^p, \dots, x_n^p),$$

therefore $M_p(x_1, \dots, x_n) \rightarrow \max(x_1, \dots, x_n)$; one writes

$$M_{+\infty}(x_1, \dots, x_n) = \max(x_1, \dots, x_n); \quad M_{-\infty}(x_1, \dots, x_n) = \min(x_1, \dots, x_n)$$

(the latter being similar to the former) and observes that $M_{-\infty}(x_1, \dots, x_n) \leq M_{-1}(x_1, \dots, x_n) \leq M_0(x_1, \dots, x_n) \leq M_1(x_1, \dots, x_n) \leq M_{+\infty}(x_1, \dots, x_n)$. That is interesting! Maybe $M_p \leq M_q$ whenever $p \leq q$?

We treat M_p as a function on $(0, \infty)^n \subset \mathbb{R}^n$ and calculate its gradient ∇M_p , or rather, the direction of the vector ∇M_p ; indeed, we only need to know when two vectors $\nabla M_p, \nabla M_q$ are linearly dependent, that is, collinear (denote it \parallel). We have $\nabla M_p \parallel \nabla M_p^p \parallel \nabla(nM_p^p) \parallel (x_1^{p-1}, \dots, x_n^{p-1})$ for $p \neq 0$; however, this result holds for $p = 0$ as well, since $\nabla M_0 \parallel \nabla \ln M_0 \parallel (x_1^{-1}, \dots, x_n^{-1})$. Thus, $\nabla M_p, \nabla M_q$ are collinear if and only if $\frac{x_1^{q-1}}{x_1^{p-1}} = \dots = \frac{x_n^{q-1}}{x_n^{p-1}}$, that is, $x_1^{q-p} = \dots = x_n^{q-p}$, or just $x_1 = \dots = x_n$. In this case, evidently, $M_p = M_q$. Does it prove that $M_p \leq M_q$ always? Not yet. Functions M_p, M_q are continuously differentiable on the open set $G = (0, \infty)^n$, and on the set $Z_p = \{x \in G : M_p(x) = 1\}$ ¹ the conditions of 3a1 are violated at one point $(1, \dots, 1)$ only. This could not happen on a compact Z_p ! Surely Z_p is not compact, and we must examine $\overline{Z_p} \setminus Z_p$ and/or ∞ .

CASE 1: $0 < p < q < \infty$. The set Z_p is bounded, since $\max(x_1, \dots, x_n) \leq (x_1^p + \dots + x_n^p)^{1/p} = n^{1/p} M_p(x_1, \dots, x_n) = n^{1/p}$, but not closed.² Functions M_p, M_q are continuous on $\overline{G} = [0, \infty)^n$. Maybe the (global) minimum of M_q on $\overline{Z_p} = \{x \in \overline{G} : M_p(x) = 1\}$ is reached at some $x \in \overline{Z_p} \setminus Z_p$? In this case at least one coordinate of x vanishes. We use induction in n . For $n = 1$, $M_p(x) = x = M_q(x)$. Having $M_p \leq M_q$ in dimension $n - 1$ we get (assuming

¹No need to consider $M_p(x) = c$, since $M_p(\lambda x) = \lambda M_p(x)$ for all $\lambda \in (0, \infty)$ and all p , thus $\frac{M_q(\lambda x)}{M_p(\lambda x)}$ does not depend on λ .

²For example, the point $(n^{1/p}, 0, \dots, 0)$ belongs to $\overline{Z_p} \setminus Z_p$.

$x_n = 0$)

$$\begin{aligned} \frac{M_q(x)}{M_p(x)} &= \frac{\left(\frac{1}{n}(x_1^q + \cdots + x_{n-1}^q + 0^q)\right)^{1/q}}{\left(\frac{1}{n}(x_1^p + \cdots + x_{n-1}^p + 0^p)\right)^{1/p}} = \\ &= \left(\frac{n}{n-1}\right)^{\frac{1}{p}-\frac{1}{q}} \frac{\left(\frac{1}{n-1}(x_1^q + \cdots + x_{n-1}^q)\right)^{1/q}}{\left(\frac{1}{n-1}(x_1^p + \cdots + x_{n-1}^p)\right)^{1/p}} \geq \left(\frac{n}{n-1}\right)^{\frac{1}{p}-\frac{1}{q}} > 1, \end{aligned}$$

therefore $M_q > M_p$ on $\bar{Z}_p \setminus Z_p$.

CASE 2: $0 = p < q < \infty$. Follows from Case 1 via the limiting procedure $p \rightarrow 0+$.

CASE 3: $-\infty < p < q < 0$. Follows from Case 1 applied to $1/x_1, \dots, 1/x_n$, since

$$\begin{aligned} 1/M_{-p}(x_1^{-1}, \dots, x_n^{-1}) &= \left(\frac{x_1^p + \cdots + x_n^p}{n}\right)^{1/p} = M_p(x_1, \dots, x_n); \\ M_p(x_1, \dots, x_n) &= 1/M_{-p}(x_1^{-1}, \dots, x_n^{-1}) \leq 1/M_{-q}(x_1^{-1}, \dots, x_n^{-1}) = M_q(x_1, \dots, x_n). \end{aligned}$$

CASE 4: $-\infty < p < q = 0$. Follows from Case 3 via the limiting procedure $q \rightarrow 0-$.

CASE 5: $-\infty < p < 0 < q < \infty$. Follows from Cases 2 and 4: $M_p \leq M_0 \leq M_q$.

So, $M_p \leq M_q$ whenever $p \leq q$.

Some practical advice.

The system of $m + n$ equations proposed in Sect. 3a is only one way of finding local constrained extrema. Not necessarily the simplest way.

No need to find ∇f when $f(\cdot) = \varphi(g(\cdot))$; just find ∇g and note that ∇f is collinear to ∇g .

In many cases there are alternatives to the Lagrange method. For example, we could replace $\inf\{M_q(x) : M_p(x) = 1\}$ with $\inf\left\{\frac{M_q(x)}{M_p(x)} : M_1(x) = 1\right\}$, substitute $x_n = n - (x_1 + \cdots + x_{n-1})$ and optimize in x_1, \dots, x_{n-1} without constraints. Alternatively we could use convexity of the function $t \mapsto t^{q/p}$, that is, convexity of the set $A = \{(t, u) : t \in (0, \infty), u \geq t^{q/p}\}$. The convex combination $\left(\frac{1}{n}(x_1^p + \cdots + x_n^p), \frac{1}{n}(x_1^q + \cdots + x_n^q)\right)$ of points $(x_1^p, x_1^q), \dots, (x_n^p, x_n^q) \in A$ belongs to A , which gives $\left(\frac{1}{n}(x_1^p + \cdots + x_n^p)\right)^{q/p} \leq \frac{1}{n}(x_1^q + \cdots + x_n^q)$, that is, $M_p \leq M_q$. Moreover, the same applies to *weighted* mean

$$M_{p,w}(x) = (x_1^p w_1 + \cdots + x_n^p w_n)^{1/p}$$

for given $w_1, \dots, w_n \geq 0$ satisfying $w_1 + \dots + w_n = 1$. In particular, $M_{1,w}(x) \leq M_{p,w}(x)$ for $p \geq 1$, that is, $x_1 w_1 + \dots + x_n w_n \leq (x_1^p w_1 + \dots + x_n^p w_n)^{1/p}$. Substituting $x_i = a_i b_i^{-q/p}$ and $w_i = b_i^q$ where q is such that $\frac{1}{p} + \frac{1}{q} = 1$ we have $\sum_i a_i b_i^{-q/p} b_i^q \leq (\sum_i a_i^p b_i^{-q} b_i^q)^{1/p}$, that is, $\sum_i a_i b_i \leq (\sum_i a_i^p)^{1/p}$ provided that $\sum_i b_i^q = 1$. This leads easily to the *Hölder's inequality*

$$\left| \sum_i x_i y_i \right| \leq \left(\sum_i |x_i|^p \right)^{1/p} \left(\sum_i |y_i|^q \right)^{1/q}$$

for $p, q \in (1, \infty)$, $\frac{1}{p} + \frac{1}{q} = 1$, and arbitrary $x_i, y_i \in \mathbb{R}$. The right-hand side may be rewritten as $nM_p(|x|)M_q(|y|)$, admitting $p, q \in [1, \infty]$. Note the special cases $p = q = 2$ and $p = 1, q = \infty$.

However, the shown way to this inequality is rather tricky.

3b3 Exercise. Given $a_1, \dots, a_n > 0$, maximize $a_1 x_1 + \dots + a_n x_n$ on $\{x \in [0, \infty)^n : x_1^p + \dots + x_n^p = 1\}$ using the Lagrange method.¹ Deduce Hölder's inequality.

Hölder's inequality persists in the case of countably many variables x_i and y_i . If two series $\sum |x_i|^p$ and $\sum |y_i|^q$ converge (and $\frac{1}{p} + \frac{1}{q} = 1$), then the series $\sum x_i y_i$ also converges (and the inequality holds).

3b4 Exercise. Given $a, b, c, k > 0$, find the maximum of the function $f(x, y, z) = x^a y^b z^c$ where $x, y, z \in [0, \infty)$ and $x^k + y^k + z^k = 1$.

3b5 Exercise. Find the maximum of y over all points $(x, y) \in \mathbb{R}^2$ that satisfy the equation $x^2 + xy + y^2 = 27$.

3c Example: Three points on a spheroid

We consider an ellipsoid of revolution (in other words, spheroid)

$$x^2 + y^2 + \alpha z^2 = 1$$

for some $\alpha \in (0, 1) \cup (1, \infty)$, and three points P, Q, R on this surface. We want to maximize $|PQ|^2 + |QR|^2 + |RP|^2$.

We'll see that the maximum is reached when P, Q, R are situated either in the horizontal plane $z = 0$ or the vertical plane $y = 0$ (or another vertical plane through the origin; they all are equivalent due to symmetry). Thus, the three-dimensional problem boils down to a pair of two-dimensional problems (not to be solved here).

¹Hint: induction in n is needed again.

We introduce 9 coordinates,

$$P = (x_1, y_1, z_1), \quad Q = (x_2, y_2, z_2), \quad R = (x_3, y_3, z_3)$$

and 4 functions $f, g_1, g_2, g_3 : \mathbb{R}^9 \rightarrow \mathbb{R}$ of these coordinates,

$$\begin{aligned} f(x_1, \dots, z_3) &= (x_1 - x_2)^2 + (y_1 - y_2)^2 + (z_1 - z_2)^2 \\ &\quad + (x_2 - x_3)^2 + (y_2 - y_3)^2 + (z_2 - z_3)^2 \\ &\quad + (x_3 - x_1)^2 + (y_3 - y_1)^2 + (z_3 - z_1)^2; \\ g_1(x_1, \dots, z_3) &= x_1^2 + y_1^2 + \alpha z_1^2 - 1, \\ g_2(x_1, \dots, z_3) &= x_2^2 + y_2^2 + \alpha z_2^2 - 1, \\ g_3(x_1, \dots, z_3) &= x_3^2 + y_3^2 + \alpha z_3^2 - 1. \end{aligned}$$

We use the approach of Sect. 3a with $n = 9$, $m = 3$. The functions f, g_1, g_2, g_3 are continuously differentiable on \mathbb{R}^9 . The set $Z = Z_{g_1, g_2, g_3} \subset \mathbb{R}^9$ is compact. The gradients of g_1, g_2, g_3 do not vanish on Z (check it) and are linearly independent (and moreover, orthogonal).

We introduce Lagrange multipliers $\lambda_1, \lambda_2, \lambda_3$ corresponding to g_1, g_2, g_3 and consider a system of $m + n = 12$ equations for 12 unknowns. The first three equations are

$$x_1^2 + y_1^2 + \alpha z_1^2 = 1, \quad x_2^2 + y_2^2 + \alpha z_2^2 = 1, \quad x_3^2 + y_3^2 + \alpha z_3^2 = 1.$$

Now, the partial derivatives. We have

$$\frac{\partial f}{\partial x_1} = 2(x_1 - x_2) - 2(x_3 - x_1) = 4x_1 - 2x_2 - 2x_3,$$

which is convenient to write as $6x_1 - 2(x_1 + x_2 + x_3)$; similarly,

$$\begin{aligned} \frac{\partial f}{\partial x_k} &= 6x_k - 2(x_1 + x_2 + x_3), \\ \frac{\partial f}{\partial y_k} &= 6y_k - 2(y_1 + y_2 + y_3), \\ \frac{\partial f}{\partial z_k} &= 6z_k - 2(z_1 + z_2 + z_3) \end{aligned}$$

for $k = 1, 2, 3$. Also,

$$\frac{\partial g_k}{\partial x_k} = 2x_k, \quad \frac{\partial g_k}{\partial y_k} = 2y_k, \quad \frac{\partial g_k}{\partial z_k} = 2\alpha z_k;$$

other partial derivatives vanish. We get 9 more equations:

$$\begin{aligned} 6x_k - 2(x_1 + x_2 + x_3) &= \lambda_k \cdot 2x_k, \\ 6y_k - 2(y_1 + y_2 + y_3) &= \lambda_k \cdot 2y_k, \\ 6z_k - 2(z_1 + z_2 + z_3) &= \lambda_k \cdot 2\alpha z_k \end{aligned}$$

for $k = 1, 2, 3$. That is,

$$\begin{aligned} (3 - \lambda_k)x_k &= x_1 + x_2 + x_3, \\ (3 - \lambda_k)y_k &= y_1 + y_2 + y_3, \\ (3 - \alpha\lambda_k)z_k &= z_1 + z_2 + z_3. \end{aligned}$$

We note that

$$(x_1 + x_2 + x_3)y_k = (3 - \lambda_k)x_k y_k = (y_1 + y_2 + y_3)x_k$$

for $k = 1, 2, 3$.

CASE 1: $x_1 + x_2 + x_3 \neq 0$ or $y_1 + y_2 + y_3 \neq 0$.

Then P, Q, R are situated on the vertical plane $\{(x, y, z) : (x_1 + x_2 + x_3)y = (y_1 + y_2 + y_3)x\}$.

CASE 2: $x_1 + x_2 + x_3 = y_1 + y_2 + y_3 = 0$ and $(\lambda_1, \lambda_2, \lambda_3) \neq (3, 3, 3)$.

If $\lambda_1 \neq 3$ then $x_1 = y_1 = 0$; the three vectors $(x_1, y_1), (x_2, y_2), (x_3, y_3) \in \mathbb{R}^2$ (of zero sum!) are collinear; therefore P, Q, R are situated on a vertical plane (again). The same holds if $\lambda_2 \neq 3$ or $\lambda_3 \neq 3$.

CASE 3: $x_1 + x_2 + x_3 = y_1 + y_2 + y_3 = 0$ and $\lambda_1 = \lambda_2 = \lambda_3 = 3$.

Then $z_1 = z_2 = z_3 = \frac{z_1 + z_2 + z_3}{3 - 3\alpha}$ (since $\alpha \neq 0$), therefore $z_1 = z_2 = z_3 = 0$; P, Q, R are situated on the horizontal plane $\{(x, y, z) : z = 0\}$.

Another practical advice.

If Lagrange method does not solve a problem to the end, it may still give a useful information. Combine it with other methods as needed.

3c1 Exercise. ¹

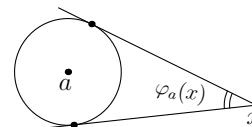
Let $a, b \in \mathbb{R}^n$ be linearly independent, $|a| = 5$, $|b| = 10$.

Functions φ_a, φ_b on the sphere $S_1(0) = \{x : |x| = 1\} \subset \mathbb{R}^n$

are defined as follows: $\varphi_a(x)$ is the angular diameter of the sphere $S_1(a) = \{y : |y - a| = 1\}$ viewed from x ;

similarly, $\varphi_b(x)$ is the angular diameter of $S_1(b)$ from x .

Prove that every point of local extremum of the function $\varphi_a + \varphi_b$ on $S_1(0)$ is some linear combination of a, b .²



¹Exam of 26.01.14, Question 2.

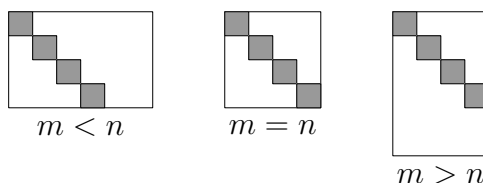
²Hint: show that $\sin \frac{1}{2}\varphi_a(x) = 1/|x - a|$; use the gradient.

3d Example: Singular value decomposition

3d1 Proposition. Every linear operator from one finite-dimensional Euclidean vector space to another sends some orthonormal basis of the first space into an orthogonal system in the second space.

This is called the Singular Value Decomposition.¹ It may be reformulated as follows.

3d2 Proposition. Every linear operator from an n -dimensional Euclidean vector space to an m -dimensional Euclidean vector space has a diagonal $m \times n$ matrix in some pair of orthonormal bases.



In particular, this holds for every linear operator $\mathbb{R}^n \rightarrow \mathbb{R}^n$. It does not mean that every matrix is diagonalizable! Two bases give much more freedom than one basis.

Do you think this is unrelated to constrained optimization? Wait a little. Prop. 3d1 will be derived from Prop. 3d3 below.

3d3 Proposition. Every finite-dimensional vector space endowed with two Euclidean metrics contains a basis orthonormal in the first metric and orthogonal in the second metric.

Proof. Let an n -dimensional vector space V be endowed with two Euclidean metrics. It means, two norms $|\cdot|$ and $|\cdot|_1$ corresponding to two inner products $\langle \cdot, \cdot \rangle$ and $\langle \cdot, \cdot \rangle_1$ by $|x|^2 = \langle x, x \rangle$ and $|x|_1^2 = \langle x, x \rangle_1$. We denote by E the Euclidean space $(V, |\cdot|)$ and define a mapping $A : E \rightarrow E$ by

$$\forall x, y \in E \quad \langle x, y \rangle_1 = \langle Ax, y \rangle;$$

it is well-defined, since the linear form $\langle x, \cdot \rangle_1$, as every linear form, is $\langle a, \cdot \rangle$ for some $a \in E$. It is easy to see that A is a linear operator, symmetric in the sense that

$$\forall x, y \in E \quad \langle Ax, y \rangle = \langle x, Ay \rangle.$$

¹See: Todd Will, "Introduction to the Singular Value Decomposition", <http://websites.uwlax.edu/twill/svd/> Quote:

The Singular Value Decomposition (SVD) is a topic rarely reached in undergraduate linear algebra courses and often skipped over in graduate courses.

Consequently relatively few mathematicians are familiar with what M.I.T. Professor Gilbert Strang calls "absolutely a high point of linear algebra."

We want to maximize $|\cdot|_1^2$ on the sphere $S = \{x \in E : |x| = 1\}$. We have¹

$$\nabla|x|^2 = 2x, \quad \nabla|x|_1^2 = 2Ax$$

by 1d1(a), or just by a very simple calculation:

$$\begin{aligned} |x+h|^2 &= |x|^2 + \langle x, h \rangle + \langle h, x \rangle + |h|^2 = |x|^2 + 2\langle x, h \rangle + o(|h|), \\ |x+h|_1^2 &= |x|_1^2 + \langle x, h \rangle_1 + \langle h, x \rangle_1 + |h|_1^2 = |x|_1^2 + 2\langle Ax, h \rangle + o(|h|). \end{aligned}$$

These two gradients are collinear if and only if $\exists \lambda \ Ax = \lambda x$; it means, x is an eigenvector of A , and λ is the eigenvalue. Now we could use well-known results of linear algebra, but here is the analytic way.

By compactness, $|\cdot|_1^2$ reaches its maximum on S ; by Theorem 3a1, a maximizer is an eigenvector. Existence of an eigenvector is thus proved. Denote it by e_n , and the eigenvalue by λ_n .

If $x \perp e_n$ then $Ax \perp e_n$ due to symmetry of A : $\langle Ax, e_n \rangle = \langle x, Ae_n \rangle = \langle x, \lambda_n e_n \rangle = \lambda_n \langle x, e_n \rangle = 0$. We consider a hyperplane (that is, $(n-1)$ -dimensional subspace)

$$E_{n-1} = \{x \in E : x \perp e_n\}$$

and the restricted operator

$$A_{n-1} : E_{n-1} \rightarrow E_{n-1}, \quad A_{n-1}x = Ax \text{ for } x \in E_{n-1}.$$

The Euclidean space E_{n-1} is endowed with two Euclidean metrics $|\cdot|$ and $|\cdot|_1$ (restricted to E_{n-1}), and $\langle x, y \rangle_1 = \langle A_{n-1}x, y \rangle$ for $x, y \in E_{n-1}$.

Now we use induction in n . The case $n=1$ is trivial. The claim for $n-1$ applied to E_{n-1} gives a basis (e_1, \dots, e_{n-1}) of E_{n-1} orthonormal in $|\cdot|$ and orthogonal in $|\cdot|_1$. Thus, $(e_1, \dots, e_{n-1}, e_n)$ is a basis of E . We normalize e_n to $|e_n| = 1$; now this basis is orthonormal in $|\cdot|$. It is also orthogonal in $|\cdot|_1$, since $\langle e_k, e_n \rangle_1 = \langle Ae_k, e_n \rangle = 0$ for $k = 1, \dots, n-1$. \square

3d4 Remark. Positivity of the quadratic form $x \mapsto |x|_1^2 = \langle x, x \rangle_1$ was not used. The same holds for arbitrary quadratic form on a Euclidean space. (In contrast, positivity of $|\cdot|^2$ was used.)

Proof of Prop. 3d1. We have two Euclidean spaces E, E_2 and a linear operator $T : E \rightarrow E_2$. First, assume in addition that T is one-to-one. Then T induces a second Euclidean metric on E :

$$|x|_1 = |Tx|; \quad \langle x, y \rangle_1 = \langle Tx, Ty \rangle$$

¹All gradients are taken in $E = (V, |\cdot|)$, not $(V, |\cdot|_1)$!

(of course, $|Tx|$ is the norm in E_2). Prop. 3d3 gives an orthonormal basis (e_1, \dots, e_n) of E , orthogonal in the second metric: $\langle e_k, e_l \rangle_1 = 0$ for $k \neq l$. That is, $\langle Te_k, Te_l \rangle = 0$, which shows that (Te_1, \dots, Te_n) is an orthogonal system in E_2 .

If T is not one-to-one, the same argument applies due to Remark 3d4.¹ \square

Prop. 3d2 follows immediately, and gives a diagonal matrix. Its diagonal elements can be made ≥ 0 (changing signs of basis vectors as needed) and decreasing (renumbering basis vectors as needed); this way one gets the so-called *singular values* of the given operator T . They depend on T only, not on the choice of the pair of bases,^{2,3} and are the square roots of the eigenvalues of the operator $A = T^*T$. The highest singular value is the operator norm $\|T\|$ of T (think, why). The lowest singular value (if not 0) is $1/\|T^{-1}\|$.

3e Sensitivity of optimum to parameters

When using a mathematical model one often bothers about sensitivity⁴ of the result (the output of the model) to the assumptions (the input). Here is one of such questions.⁵

What happens if the restrictions $g_1(x) = \dots = g_m(x) = 0$ are replaced with $g_1(x) = c_1, \dots, g_m(x) = c_m$?

Assume that the system of $m + n$ equations

$$\begin{aligned} g_1(x) = c_1, \dots, g_m(x) = c_m, & \quad (m \text{ equations}) \\ \nabla f(x) = \lambda_1 \nabla g_1(x) + \dots + \lambda_m \nabla g_m(x) & \quad (n \text{ equations}) \end{aligned}$$

for $(\lambda, x) \in \mathbb{R}^m \times \mathbb{R}^n$ has a solution $(\lambda(c), x(c))$ for all $c \in \mathbb{R}^m$ near 0, and the mapping $c \mapsto x(c)$ is differentiable at 0. Then, by the chain rule,

$$\left. \frac{\partial}{\partial c_k} \right|_{c=0} f(x(c)) = \left\langle \nabla f(x(0)), \left. \frac{\partial}{\partial c_k} \right|_{c=0} x(c) \right\rangle \quad \text{for } k = 1, \dots, m.$$

On the other hand,

$$\nabla f(x(0)) = \lambda_1(0) \nabla g_1(x(0)) + \dots + \lambda_m(0) \nabla g_m(x(0))$$

¹Alternatively, define $|x|_1^2 = |Tx|^2 + |x|^2$, $\langle x, y \rangle_1 = \langle Tx, Ty \rangle + \langle x, y \rangle$.

²The only freedom in this choice (in addition to sign change and renumbering) is, rotation within each eigenspace of dimension > 1 (if any).

³On the space of operators, the *Schatten norm* is $\|T\|_p = (|s_1|^p + \dots + |s_n|^p)^{1/p}$ where s_1, \dots, s_n are the singular values of T (and $1 \leq p \leq \infty$).

⁴Closely related ideas: stability, robustness; uncertainty; elasticity, ...

⁵A more general one: $g_1(x, c_1) = 0, \dots, g_m(x, c_m) = 0$.

and

$$\left\langle \nabla g_1(x(0)), \frac{\partial}{\partial c_k} \Big|_{c=0} x(c) \right\rangle = \frac{\partial}{\partial c_k} \Big|_{c=0} g_1(x(c)) = \begin{cases} 1, & \text{if } k = 1, \\ 0, & \text{otherwise} \end{cases}$$

(since $g_1(x(c)) = c_1$). The same holds for g_2, \dots, g_m . Therefore

$$\frac{\partial}{\partial c_k} \Big|_{c=0} f(x(c)) = \lambda_k(0).$$

It means that $\lambda_k = \lambda_k(0)$ is the sensitivity of the critical value to the level c_k of the constraint $g_k(x) = c_k$. That is,

$$f(x(c)) = f(x(0)) + \lambda_1(0)c_1 + \dots + \lambda_m(0)c_m + o(|c|).$$

Does it mean that

$$(3e1) \quad \sup_{Z_c} f = \sup_{Z_0} f + \lambda_1(0)c_1 + \dots + \lambda_m(0)c_m + o(|c|)$$

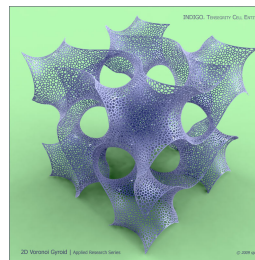
where $Z_c = \{x : g_1(x) = c_1, \dots, g_m(x) = c_m\}$? Not necessarily, for several reasons (possible non-compactness, non-differentiability, greater or equal value at another critical point when $c = 0$). But if $\sup_{Z_c} f = f(x(c))$ for all c near 0 then (3e1) holds.¹

3f Manifolds in \mathbb{R}^n

Everyone knows what a curve is, until he has studied enough mathematics. . .

Felix Klein²

Image: (CC) Jonathan Johanson,
<http://cliptic.wordpress.com>



By a manifold (to be defined soon) we mean a differential k -dimensional submanifold of \mathbb{R}^n , of class C^1 , without boundary.³ It is also called “ k -dimensional smooth surface in \mathbb{R}^n ” or “ k -dimensional submanifold on \mathbb{R}^n ”,⁴ or “smooth manifold in \mathbb{R}^n ”⁵ etc.

¹See also Sect. 13.2 in book: J. Cooper, “Working analysis”, Elsevier 2005.

²Quoted from: Hubbard, Sect. 3.1 “Manifolds”.

³Generally, “smooth” means “as many times differentiable as is relevant to the problem at hand. . . (Some authors use “smooth” to mean C^∞ : “infinitely many times differentiable”. For our purposes this is overkill.)” Hubbard, Sect. 3.1, p. 293–294.

⁴Zorich Sect. 8.7.1.

⁵Hubbard Sect. 3.1.

Several equivalent definitions of a manifold are used: via equations;¹ via diffeomorphisms;² via graphs of mappings;³ and via parametrizations (so-called charts, to be treated in Analysis-4).

3f1 Theorem. The following conditions on a set $M \subset \mathbb{R}^n$, a point $x_0 \in M$ and a number $k \in \{1, 2, \dots, n-1\}$ are equivalent:

(a) there exists a mapping $f : \mathbb{R}^n \rightarrow \mathbb{R}^{n-k}$, continuously differentiable near x_0 , such that $(Df)_{x_0} = A : \mathbb{R}^n \rightarrow \mathbb{R}^{n-k}$ is onto, and

$$x \in M \iff f(x) = f(x_0) \quad \text{for all } x \text{ near } x_0;$$

(b) there exists a local diffeomorphism φ near x_0 such that

$$x \in M \iff \varphi(x) \in \mathbb{R}^k \times \{0_{n-k}\} \quad \text{for all } x \text{ near } x_0;$$

(c) there exists a permutation (i_1, \dots, i_n) of $\{1, \dots, n\}$ and a mapping $g : \mathbb{R}^k \rightarrow \mathbb{R}^{n-k}$, continuously differentiable near $(x_{0,i_1}, \dots, x_{0,i_k})$, such that

$$x \in M \iff g(x_{i_1}, \dots, x_{i_k}) = (x_{i_{k+1}}, \dots, x_{i_n}) \quad \text{for all } x \text{ near } x_0.$$

Proof. First, WLOG, $x_0 = 0$ (as usual).

Second, the three conditions are insensitive to permutations of the n coordinates of x .⁴ Indeed, in (a) we may change the order of arguments of f as needed; in (b) we may change the order of arguments of φ as needed; and in (c) we may change the permutation (i_1, \dots, i_n) as needed.

(a) \implies (c): WLOG, $f(0) = 0$ and $A = (B|C)$ with $B = \mathbb{R}^k \rightarrow \mathbb{R}^{n-k}$, $C : \mathbb{R}^{n-k} \rightarrow \mathbb{R}^{n-k}$, C invertible (using the fact that $\text{rank } A = n-k$). Theorem 2b3 (for n and $n-k$ in place of n and m) gives $g : \mathbb{R}^k \rightarrow \mathbb{R}^{n-k}$ such that $g(x_1, \dots, x_k) = (x_{k+1}, \dots, x_n) \iff f(x_1, \dots, x_n) = 0 \iff x \in M$, which gives (c) for $(i_1, \dots, i_n) = (1, \dots, n)$.

(c) \implies (b): WLOG, $(i_1, \dots, i_n) = (1, \dots, n)$. Similarly to the proof of 2b3 \implies 2b1 (in Sect. 2a) we define φ by $\varphi(u, v) = (u, g(u) - v)$ for $u \in \mathbb{R}^k$ and $v \in \mathbb{R}^{n-k}$; then $\varphi(u, v) \in \mathbb{R}^k \times \{0_{n-k}\} \iff \varphi(u, v) = (u, 0) \iff g(u) = v \iff x \in M$.

(b) \implies (a): we define $f(x) = (y_{k+1}, \dots, y_n)$ whenever $\varphi(x) = (y_1, \dots, y_n)$; then $f(0) = 0$ and $f(x) = 0 \iff \varphi(x) \in \mathbb{R}^k \times \{0_{n-k}\} \iff x \in M$. \square

3f2 Definition. A nonempty set $M \subset \mathbb{R}^n$ is a k -dimensional *manifold*, if the equivalent conditions 3f1(a,b,c) hold for every $x_0 \in M$.

¹Fleming; also Hubbard, Th. 3.1.10.

²Lang, Zorich.

³Hubbard.

⁴I mean, coordinates of x , not of $f(x)$ or $\varphi(x)$.

We may say that M is a k -manifold near x_0 when 3f1(a,b,c) hold for M , x_0 and k . Accordingly, M is a k -manifold when it is a k -manifold near every point (of M).

3f3 Exercise. Let $\varphi : \mathbb{R}^n \rightarrow \mathbb{R}^n$ be a diffeomorphism, and $M \subset \mathbb{R}^n$.

(a) If M is a k -manifold near x_0 , then its image $\varphi(M)$ is a k -manifold near $\varphi(x_0)$;

(b) M is a k -manifold if and only if $\varphi(M)$ is a k -manifold.

Prove it.

This applies, in particular, to shifts, rotations, and all invertible affine transformations of \mathbb{R}^n .

3f4 Exercise. Let $M_1, M_2 \subset \mathbb{R}^n$ be k -dimensional manifolds, and $M = M_1 \cup M_2$.

(a) If $\overline{M_1} \cap M_2 = \emptyset$ and $M_1 \cap \overline{M_2} = \emptyset$, then M is a k -dimensional manifold. Prove it.

(b) It can happen that $M_1 \cap M_2 = \emptyset$ but M is not a k -dimensional manifold. Give a counterexample.

3f5 Exercise. Let $0 < m < n$, and $g_1, \dots, g_m \in C^1(\mathbb{R}^n \rightarrow \mathbb{R})$ be such that the vectors $\nabla g_1(x), \dots, \nabla g_m(x)$ are linearly independent for every $x \in M$ where $M = \{x : g_1(x) = \dots = g_m(x) = 0\}$. Then M is a $(n - m)$ -dimensional manifold.

Prove it.

3f6 Exercise. Which of the following subsets of \mathbb{R}^2 are 1-dimensional manifolds? Prove your answers, both affirmative and negative.

- * $M_1 = \mathbb{R} \times \{0\}$;
- * $M_2 = [0, 1] \times \{0\}$;
- * $M_3 = (0, 1) \times \{0\}$;
- * $M_4 = \{(0, 0)\}$;
- * $M_5 = \mathbb{R} \times \{0, 1\}$;
- * $M_6 = \mathbb{R} \times \mathbb{Z}$;
- * $M_7 = \mathbb{R} \times \{1, \frac{1}{2}, \frac{1}{3}, \dots\}$;
- * $M_8 = M_7 \cup M_1$.

3f7 Example. The sphere $S = \{x \in \mathbb{R}^n : |x| = 1\}$ is a $(n - 1)$ -dimensional manifold (by 3f5 for $m = 1$ and $g(x) = |x|^2 - 1$).

Alternatively, we may prove that S is a manifold around just one point, say, $e_1 = (1, 0, \dots, 0)$, and then use rotation invariance: $U(S) = S$ for every

linear isometry $U : \mathbb{R}^n \rightarrow \mathbb{R}^n$, and each $x \in S$ is Ue_1 for some U ;¹ use 3f3(a). Near e_1 the equality $x_1 = \sqrt{1 - x_2^2 - \cdots - x_n^2}$ gives 3f1(c).

3f8 Example.² Consider the set M of all 3×3 matrices A of the form

$$A = \begin{pmatrix} a^2 & ab & ac \\ ba & b^2 & bc \\ ca & cb & c^2 \end{pmatrix} \quad \text{for } a, b, c \in \mathbb{R}, \quad a^2 + b^2 + c^2 = 1.$$

These are orthogonal projections to one-dimensional subspaces of \mathbb{R}^3 , that is, straight lines through the origin. Note that each line contains two points of the sphere $S = \{(a, b, c) \in \mathbb{R}^3 : a^2 + b^2 + c^2 = 1\}$, which gives a 2-to-1 mapping $S \rightarrow M$. We treat M as a subset of the six-dimensional space of all symmetric 3×3 matrices.

The set M is invariant under transformations $A \mapsto UAU^{-1}$ where U runs over all orthogonal matrices (linear isometries); these are linear transformations of the six-dimensional space of matrices. If A corresponds to $x = (a, b, c)$ then UAU^{-1} corresponds to Ux . For arbitrary $A, B \in M$ there exists U such that $UAU^{-1} = B$ (“transitive action”).

Thus, M looks the same around all its points (“homogeneous space”). In order to prove that M is a 2-manifold (in \mathbb{R}^6) it is sufficient to prove this near a single point of M , say,

$$A_1 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} \in M,$$

that corresponds to $(a, b, c) = (1, 0, 0)$ (but also $(-1, 0, 0)$, of course). For $(a, b, c) \rightarrow (1, 0, 0)$ we have in the linear approximation

$$\begin{pmatrix} a^2 & ab & ac \\ ba & b^2 & bc \\ ca & cb & c^2 \end{pmatrix} \approx \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} + \begin{pmatrix} 0 & b & c \\ b & 0 & 0 \\ c & 0 & 0 \end{pmatrix}$$

(think, why). Thus, in the linear approximation all elements of A are functions of two of them. Returning to the nonlinear situation we want to express a^2, b^2, c^2 and bc in terms of ab and ac (locally, for (a, b, c) near $(1, 0, 0)$). We

¹Since x is the first vector of some orthogonal basis.

²The projective plane in disguise.

have

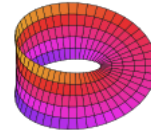
$$\begin{aligned}(ab)^2 + (ac)^2 &= a^2(b^2 + c^2) = a^2(1 - a^2); \\ a^2 &= \frac{1}{2} + \sqrt{\frac{1}{4} - (ab)^2 - (ac)^2}; \\ b^2 &= \frac{(ab)^2}{\frac{1}{2} + \sqrt{\dots}}; \quad c^2 = \frac{(ac)^2}{\frac{1}{2} + \sqrt{\dots}}; \quad bc = \frac{(ab)(ac)}{\frac{1}{2} + \sqrt{\dots}};\end{aligned}$$

thus, M is a 2-manifold near A_1 according to 3f1(c).¹

Interestingly, the part of M that corresponds to a spherical zone (symmetrical, around the equator), say $a^2 + b^2 + c^2 = 1$, $|c| < 1/2$, is homeomorphic to the Möbius strip² (without the edge),

$$M = \{h(s, \theta) : s \in (-1, 1), \theta \in [0, 2\pi]\},$$

$$h(s, \theta) = \begin{pmatrix} (R+rs \cos \frac{\theta}{2}) \cos \theta \\ (R+rs \cos \frac{\theta}{2}) \sin \theta \\ rs \sin \frac{\theta}{2} \end{pmatrix},$$



for given $R > r > 0$. You see, a straight segment on the x, z plane rotates by $\theta/2$ (around the y axis) and at the same time it rotates (in the three dimensions) by θ around the z axis.

A point $h(s, \theta)$ of the Möbius strip corresponds to the point

$$\left(\sqrt{1 - \frac{1}{4}s^2} \cos \frac{1}{2}\theta, \sqrt{1 - \frac{1}{4}s^2} \sin \frac{1}{2}\theta, \frac{1}{2}s \right)$$

on the sphere S , and the corresponding point of M . (Think, what happens for $\theta = 2\pi$.)

The rest of M is homeomorphic to a disk (not two disks), and this disk is glued to the Möbius strip in a way unthinkable in three dimensions.³

¹It is easy to check that, locally, every matrix that satisfies these equations belongs to M .

²Images from Wikipedia, "Möbius strip".

³Dimension 6 can be reduced to dimension 4 by taking only $(a^2 - b^2, ab, ac, bc)$, see "Real projective plane" in Wikipedia.

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